

UNIVERSIDAD NACIONAL AUTÓNOMA DE MÉXICO PROGRAMA DE MAESTRÍA Y DOCTORADO EN CIENCIAS MÉDICAS, ODONTOLÓGICAS Y DE LA SALUD FACULTAD DE MEDICINA CAMPO DE CONOCIMIENTO: CIENCIAS MÉDICAS

BIOMARCADORES AVANZADOS DE RESONANCIA MAGNÉTICA COMO FACTORES PRONÓSTICOS PARA LA SOBREVIDA DE PACIENTES CON GLIOBLASTOMA MULTIFORME.

TESIS QUE PARA OPTAR POR EL GRADO DE DOCTOR EN CIENCIAS

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RESUMEN

La evaluación por imagen de los tumores cerebrales implica la obtención de secuencias convencionales que convencionales consisten en el T1 con y sin gadolinio (medio de contraste), el T2, la difusión y la secuencia Flair. Entre los años 2006 y 2009, diversas publicaciones en revistas especializadas de neurología e imagenología, presentaron la propuesta de incluir secuencias para "análisis avanzado" en el estudio de los tumores cerebrales.

Estas secuencias están representadas por: la secuencia de difusión y su post proceso para medir el *Coeficiente de Difusión Aparente* (ADC); la *perfusión cerebral*, que permite cuantificar *el volumen y el flujo sanguíneo cerebral (CBV, CBF)*; la *espectroscopia*, que permite cuantificar diferentes metabolitos en el tumor y zona peritumoral, y dar evidencia de infiltración en áreas que todavía no presentan realce al contraste. La ultima técnica de análisis avanzado es la secuencia *tensor de difusión (DTI)*, que mide la orientación tridimensional del movimiento de las moléculas de agua en el tejido nervioso y con ello, crea mapas cuantitativos y permite medir diferentes biomarcadores.

Esta tesis presenta el protocolo de investigación realizado para el análisis avanzado de secuencias de resonancia magnética en el estudio de sobrevida de pacientes con glioblastoma multiforme; dos artículos originales en proceso de revisión que representan los avances realizados hasta el momento de presentación y tres artículos originales publicados en revistas indexadas como parte de los productos del proyecto de investigación.

ABSTRACT

Imaging evaluation of brain tumors implies the acquisition of magnetic resonance imaging (MRI) sequences known as conventional; they are represented by the T1-weighted before and after gadolinium administration, T2-weighted, diffusion and Flair. Between 2006 and 2009, several publications in indexed journals in the fields of neurology and imaging proposed the inclusion of sequences for the "Advanced Analysis" of brain tumors.

This group of images comprises the sequences of diffusion, and its image postprocessing to measure the *apparent diffusion coefficient* (ADC); the *perfusion weighted imaging* (PWI), which allows the calculation of cerebral blood flow and cerebral blood volume; the *spectroscopy* (MRS) which quantifies several brain metabolites in the tumor and peritumoral regions given evidence of infiltrated areas of brain tissue, still not showing enhance after contrast administration. The last advance sequence if the diffusion tensor imaging (DTI), that measures the 3D movement of water molecules in brain tissue; with this sequence it is possible to create quantitative maps that derive up to eleven different biomarkers.

In this thesis, we present the original research protocol proposed at the National Institute of Neurology and Neurosurgery for the study of survival in patients with glioblastoma multiforme using advanced biomarkers of MRI; the full text of two manuscripts currently under review as part of the advances of the project and three full-text, published, original articles as evidence of the products of this research project.

Introducción

Los tumores gliales de alto grado (glioblastoma multiforme) representan el tipo más común de tumor cerebral maligno alcanzando el 40-50% de las neoplasias cerebrales primarias. Los esquemas actuales para el tratamiento de estos tumores implica la combinación de cirugía con radioterapia/quimioterapia adyuvante. Sin embargo, el pronóstico global para estos tumores es pobre todavía a pesar de los tratamientos agresivos, especialmente en pacientes con gliomas de alto grado (IV) cuyo promedio de sobrevida después de la cirugía es de solamente 8-9 meses en México y en hospitales de países desarrollados alcanza una media de sobrevida de hasta 1.4 meses. El pronóstico de los pacientes varía significativamente dependiendo de la histología y del tratamiento empleado. En la actualidad existen publicaciones discordantes, ya que aún con el mismo diagnóstico patológico y tratamientos equivalentes, algunos pacientes tienen un resultado relativamente favorable mientras que otros tienen un pronóstico muy pobre.

Entre los años 2006 y 2009, diversas publicaciones en revistas especializadas de neurología e imagenología, presentaron la propuesta de complementar con secuencias para "análisis avanzado" las imágenes convencionales de resonancia magnética que hasta la fecha se adquieren en el estudio de tumores cerebrales, las secuencias convencionales consisten en el T1 con y sin medio de contraste, el T2 con gadolinio, la difusión y la secuencia Flair.

El tejido tumoral que muestra reactividad al medio de contraste (realce) se le conoce como tejido tumoral viable, y las regiones de tejido cerebral que muestran una señal hiperintensa en las secuencias T2 y Flair, se les conoce como áreas de edema vasogenico, que debido al comportamiento histológico de este tipo de tumores, no descarta tener infiltración microscópica peritumoral.

Las secuencias de análisis avanzado están representadas por el post proceso de la difusión para medir la velocidad de las moléculas de agua en el área de interés, lo que se conoce como <u>Coeficiente de</u> <u>Difusión Aparente</u> (ADC); la <u>perfusión cerebral</u>, técnica con la que puede cuantificarse el volumen y el flujo sanguíneo cerebral, el primero de ellos, al ser cuantificado en el área de tumor se le conoce como volumen tumoral relativo, y es el biomarcador utilizado para evaluar la respuesta a los nuevos agentes quimioterapéuticos; la siguiente técnica la representa la <u>espectroscopia</u>, en la que se pueden cuantificar metabolitos en el área tumoral y peritumoral, lo que puede dar evidencia de infiltración tumoral aun en áreas que no presentan realce al contraste. La ultima técnica de análisis avanzado esta representada por la secuencia llamada <u>tensor de difusión</u>, con la cual, es posible medir la orientación tridimensional del movimiento de las moléculas en los axones y demás componentes del tejido nervioso y con ello, realizar no solo la representación tridimensional de tractos cerebrales, sino también la creación de mapas cuantitativos, que sometidos a análisis estadístico multivariado, pueden descubrir nuevos comportamientos en el tejido cerebral infiltrado por tumor, para este proyecto, glioblastoma multiforme, que pueden apoyar el diagnostico de la sobrevida en los pacientes.

Es importante comprender los factores que contribuyen positiva o negativamente al pronóstico de los pacientes con lesiones cerebrales gliales de alto grado. Sin embargo, pocos estudios a nivel internacional han analizado los factores pronósticos de estas lesiones con biomarcadores avanzados de resonancia magnética, no conocemos un estudio similar que haya sido desarrollado en México hasta la fecha.

Con este proyecto se pretende demostrar la utilidad clínica de nuevos biomarcadores derivados de las secuencias avanzadas de resonancia magnética. Como parte del proyecto, deberemos conocer previamente el comportamiento del tejido cerebral normal al ser evaluado con estos nuevos biomarcadores.



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PROTOCOLO DE INVESTIGACIÓN CLÍNICA

No.: <u>84/11</u> Octubre 16 de 2012.

DEPARTAMENTO QUE PROPONE: Neuroquímica

TITULO DEL PROTOCOLO:

Biomarcadores avanzados de resonancia magnética como factores pronósticos para la sobrevida de pacientes con glioblastoma multiforme.

(3ª modificación para comité científico del INNN)

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Glosario de términos médicos relacionados al protocolo

Anisotropía	Es la propiedad general de la materia según la cual determinadas propiedades físicas, tales como: elasticidad, temperatura, conductividad, velocidad de propagación de la luz, etc. varían según la dirección en que son examinadas. Algo anisótropo podrá presentar diferentes características según la dirección. La (opuesta de isotropía)
Biopsia	procedimiento en el que se extraen muestras de tejido (con una aguja o durante una cirugía) para examinarlas con un microscopio y determinar si existen células cancerosas o anormales.
Difusión	es el proceso de movimiento molecular.
Escala Karnofsky	también llamada KPS, es la forma típica de medir la capacidad de los pacientes con cáncer de realizar tareas rutinarias. Los puntajes de la escala de rendimiento de Karnofsky oscilan entre 0 y 100. Un puntaje más alto significa que el paciente tiene mejor capacidad de realizar las actividades cotidianas. La KPS se puede usar para determinar el pronóstico del paciente, medir los cambios en la capacidad del paciente para funcionar o decidir si un paciente puede ser incluido en un estudio clínico.
Gamma Knife	tipo de radiocirugía que usa rayos gamma para tratar el cáncer de cerebro.
Glioma	(Del griego glia, liga). Tumor desarrollado a expensas del tejido nervioso, de consistencia blanda, cuya estructura se asemeja a la de la neuroglia.
Glioblastoma	Variedad de glioma formada por tejido nervioso en estado embrionario y de evolución maligna. Al lado del glioblastoma multiforme (Bailey y Cushing) o heteromorfo (Del Río Hortega) o espongioblastoma multiforme (Globus y Strauss) de células monstruosas y poliformes, se describe el glioblastoma isomorfo (Del Río Hortega) formado por células indiferenciadas.
Infiltración	acumulación de una sustancia en un tejido en el que no se encuentra en condiciones fisiológicas. Acumulación de elementos celulares en la reacción inflamatoria o de elementos tumorales.
IRM/MRI	imágenes por resonancia magnética (su sigla en inglés es) - procedimiento de diagnóstico que utiliza una combinación de imanes grandes, radiofrecuencias y una computadora para producir imágenes detalladas de los órganos y las estructuras internas del cuerpo.
radiocirugía	tipo de tratamiento de radiología terapéutica que usa haces de radiación muy focalizados para tratar cáncer y otras lesiones en un tratamiento de sesión única.
Tomografía	de las palabras griegas "cortar o seccionar" (tomos) y "escribir" (graphein), en medicina nuclear, es un método para separar la interferencia del área de interés mediante la imagen de una sección transversal del objeto.
Tomografía computarizada	(También llamada CT o CAT.) - procedimiento de diagnóstico por imágenes que utiliza una combinación de rayos X y tecnología de computadoras para producir imágenes transversales (a menudo llamadas "rebanadas"), tanto horizontales como verticales, del cuerpo. Una CT muestra imágenes detalladas de cualquier parte del cuerpo, incluyendo los huesos, los músculos, la grasa y los órganos. La CT es más detallada que los rayos X generales.

1) ANTECEDENTES

Epidemiología del glioblastoma multiforme

Los tumores gliales de alto grado (glioblastoma multiforme) representan el tipo más común de tumor cerebral maligno y representan el 40-50% de las neoplasias cerebrales primarias en adultos (1). En las últimas tres décadas, tanto en Estados Unidos de Norteamérica (EUA) como en otros países industrializados se ha observado un incremento en la mortalidad, secundario a una mayor incidencia de tumores cerebrales primarios.(2, 3) En EU, la incidencia de cáncer del cerebro se ha incrementado 1.2% por año desde 1973 y la mortalidad se ha incrementado 0.7% por año. Esto ha sido más evidente en los pacientes mayores de 60 años, en que la incidencia se incrementa 2.5% por año desde 1980. En este mismo grupo de pacientes se observa una mayor frecuencia de tumores agresivos de origen glial, en particular el glioblastoma multiforme y el astrocitoma anaplásico.(2-4) Varios autores consideran que este incremento es artificial y arguyen que la incidencia se ha incrementado debido al considerable avance en la técnica de neuroimagen ocurrido desde los años 70s, que permite un diagnóstico preciso y evita los diagnósticos erróneos, así como a un incremento en la densidad de población.

Aunque la incidencia de los tumores cerebrales es de solamente 17,000 casos por año, estos representan un problema médico considerable, ya que hasta un 60% de estos son malignos.(5) En EUA se reportan en promedio aproximadamente 12,000 nuevos casos por año, ubicándolo como el tumor intra-axial cerebral más frecuente.(6)

Los gliomas de alto grado representan la segunda causa de mortalidad por cáncer en adultos jóvenes, menores de 35 años, son la cuarta causa de muerte en menores de 54 años, y la principal causa de perdida de la capacidad productiva en esta población.(5) La edad media de diagnóstico son los 65 años con un promedio de sobrevida de 10 meses, la sobrevida media a 5 años desde el tiempo del diagnóstico es alrededor del 3.3% solamente. Esta cifra es aun más baja, de 3.5 meses para mayores de 65 años. Aproximadamente 2% de los pacientes son mayores de 65 años y solo 15% de ellos son menores de 45 años, de estos alrededor del 10 sobrevive más de 2 años.(6)

En México poco se sabe sobre la epidemiología de los tumores del SN, ya que los datos publicados a la fecha sobre los diversos tipos de gliomas rara vez mencionan el grado de malignidad de los mismos, además al ser efectuados en hospitales del Distrito Federal (DF) que por sus características de población concentran a pacientes adultos o niños (locales o foráneos) y no a ambos, por lo que dejan de ser representativos de la población general.(7) En el Instituto Nacional de Neurología y Neurocirugía (INNN) de la ciudad de México, el glioblastoma multiforme (GBM) representa el 9% de todas las

neoplasias intracraneales y el 28% de los gliomas.(8) Estadísticas aun no publicadas del departamento de Radioneurocirugía del INNN refieren la incidencia del GBM en hasta 104 casos al año con una frecuencia de ingresos por casos nuevos de 1-2 casos por semana. Alrededor del 30% de los casos se refieren en menores de 45 años de edad, lo cual contrasta con el 15% que se menciona en la literatura mundial. La sobrevida media al año del diagnóstico en el INNN es del 10% aproximadamente y a 18 meses del 5% aproximadamente.(9)

Clasificación de la OMS para los astrocitomas
Grado I-OMS: Astrocitoma pilocítico
Grado II-OMS: Astrocitoma difuso (de bajo grado)
Grado III-OMS: Astrocitoma anaplásico
Grado IV-OMS: Glioblastoma multiforme

Los tres tipos histopatológicos más frecuentes son el glioblastoma multiforme, el astrocitoma anaplásico y el oligodendroglioma anaplásico. El glioblastoma multiforme constituye la forma más agresiva de los astrocitomas (tumor grado IV-OMS).

Clasificación de los tumores cerebrales

Determinación de la malignidad según las características histológicas del tumor.

Tumores de evolución lenta (Bajo grado)

Tumores de grado I benignos, de crecimiento lento y circunscriptos. Tumores de grado II De crecimiento lento, pero con límites imprecisos, o de extensión.

Tumores de evolución rápida (alto grado)

Tumores de grado III Tumores anaplásicos, su evolución es más rápida. Tumores de grado IV Tumores malignos, muestran signos histológicos de crecimiento muy rápido en todas las regiones examinadas.

CLASIFICACIÓN HISTOLÓGICA DE LOS TUMORES DEL SISTEMA NERVIOSO

Tumores del tejido glial	Tumores de las <u>meninges</u>	Tumores de células germinales	Tumores de la región selar	Otros
Tumores astrocíticos Astrocitoma				
Glioblastoma multiforme				
Tumores oligodendrogliales				т
Oligodendroglioma		Germinoma		metastásicos
Tumores ependimales	Meningioma	Carcinoma	Adenoma pituitario	Tumores de las vainas nerviosas
Ependimoma	Hemangiopericitoma	embrional	Carcinoma pituitario	Schwannoma
Tumores del plexo coroides	Tumor melanocítico Hemangioblastoma	Teratoma Tumor del seno	Craneofaringioma	Neurofibroma
Papiloma		endodérmico		Linfoma primario del SNC
Carcinoma				
Tumores embrionarios				
Meduloblastoma				
Tumores del parénquina pineal				

Diagnóstico Clínico

Los tumores cerebrales causan síntomas variados. En general, se distinguen las manifestaciones derivadas de la <u>hipertensión intracraneal</u>, y los síntomas secundarios a la expansión tumoral, estos últimos denominados *signos focales*, que dependen de la estructura anatómica afectada.

- Síntomas de hipertensión intracraneal
 - o Visión doble
 - o Dolor en una extremidad
 - o <u>Cefalea</u>
 - o <u>Vómitos</u>
 - Edema de pupila y alteraciones visuales
 - Trastornos del comportamiento (irritabilidad, labilidad emocional, fallos en el discernimiento, alteraciones de la <u>memoria</u>, falta de iniciativa, indiferencia a las costumbres sociales.)
- Síndromes focales: Son manifestaciones que orientan la localización de la lesión.

La presentación más común de los astrocitomas malignos es con déficit neurológico (68%) y de éstos la afección de la vía piramidal la más común (45%) también pueden presentarse con cefalea (54%) y crisis convulsivas (26%). Los síntomas anteriormente mencionados son debido a afección indirecta de dichas vías pues son atribuidos principalmente al edema vasogénico y a la inflamación perilesional producida. Existe información que indica que los síntomas clínicos y la historia natural de la enfermedad tienen un curso más agresivo en pacientes de edad avanzada que en los menores de 45 años pero esto aún está por comprobarse.(10)

Diagnóstico por Imagen del GBM

El diagnóstico se realiza mediante exámenes imagenológicos como la <u>tomografía axial computarizada</u> (<u>TAC</u>) o la <u>resonancia magnética</u> (RM), las cuales permiten conocer la localización y el tamaño del tumor y además sugerir la naturaleza del mismo, pero es la <u>biopsia</u> la que indica el tipo exacto de tumor. Antes de la tomografía (TC) muchos de esos pacientes eran catalogados erróneamente como portadores de enfermedades neurodegenerativas o cerebrovasculares.(3, 4)

Evaluación convencional con RM: Volumetría 3D y edema perilesional

La evaluación convencional es aquella proporcionada por las secuencia T1 post gadolinio para delinear la enfermedad macroscópica (volumen tumoral) (11), y el edema perilesional en cuantificado en la secuencia T2 (12).

El uso de la volumetría cerebral mediante resonancia magnética ha sido reportado en patologías que pertenecen a diferentes especialidades médicas, por ejemplo, la esclerosis múltiple ⁽¹³⁾, alcoholismo (13), Alzheimer y deterioro cognitivo leve (14), esquizofrenia (14-16), el trastorno por déficit de atención con hiperactividad (17-19), síndrome de Tourette (17), el autismo (20-22), el síndrome de Appert (23), el síndrome de Down (24), el síndrome de Rett (25), retardo del crecimiento intrauterino (26), y trastornos de conducta (27).

En el caso de los tumores cerebrales, la importancia de la volumetría está basada en el conocimiento de que el área de realce por el ácido de gadolinio-diethylenetriaminepentaacetico (Gd-DTPA) generalmente se asume que corresponde a la masa principal de tejido de tumor activo, sin embargo una evaluación global también debería considerarse en aquellos casos con un componente quístico o necrótico. La imagen anatómica que se usa en RM es el método más extensamente aceptado para determinar el blanco para el tratamiento en la radioterapia y la cirugía (28), ver figura 1.



Figura 1. Ejemplo de la aplicación de la volumetría para cuantificación de un tumor cerebral en secuencia T1 postgadolinio en los planos axial (A) y coronal (B). Se observa en ambos casos una imagen representativa del volumen en cada plano, el tumor presenta formas irregulares, que impiden la aplicación de formulas para cálculo del volumen (por ejemplo la fórmula del volumen para la esfera o la elipse no se aplican a este caso).

En el INNN, la evaluación convencional de GBM incluye: localización, volumen tumoral en secuencia T1 post gadolinio, volumen tumoral incluido el edema perilesional en T2, y el índice de edema tumoral (volumen tumoral en secuencia T1/volumen tumoral incluido el edema en secuencia T2).

Utilización del Análisis Avanzado por Resonancia Magnética en la evaluación de GBM

El concepto de "análisis avanzado" es de reciente aplicación en la evaluación con RM de tumores cerebrales, fue publicado por primera vez por investigadores del departamento de radiología de "University of Pennsylvania School of Medicine" en octubre de 2006,(29) en una revista especializada en imagenología (*Radiographics*) y al año siguiente, 2007, una publicación en una revista clínica (*Neurologic Clinics*) del departamento de radiología del Brigham and Women's Hospital de Harvard Medical School en Boston, Estados Unidos, presentó también una revisión de la aplicación clínica de estas técnicas avanzadas.(30)

Aunque en la actualidad exista mas de una docena de variables que se pueden cuantificar con RM, las secuencias de RM consideradas actualmente en el "análisis avanzado" para la evaluación diagnóstica, progresión tumoral y/o respuesta a los diferentes tratamientos son:

- ✓ Difusión,
- ✓ Perfusión,
- Tensor de Difusión, y
- ✓ Espectroscopia

A continuación se presenta una breve descripción de estas técnicas de RM.

Biomarcadores de la Espectroscopia por Resonancia Magnética

La espectroscopia por resonancia magnética (ERM) es una técnica que nos permite cuantificar diversos metabolitos e interpretarlos de manera cualitativa y/o cuantitativa (31). El objetivo principal de la ERM es brindar una herramienta para mejorar la sensibilidad y especificidad de una gran variedad de patologías, actualmente es utilizada principalmente en el sistema nervioso central y puede ser utilizada en otras regiones anatómicas como son próstata, glándula mamaria, hígado y tumores de partes blandas.

La ERM más utilizada es la del núcleo del hidrógeno. La técnica se basa en las diferencias mínimas que se encuentran en las frecuencias de resonancia de los núcleos de hidrógeno dependiendo de su entorno, a este fenómeno se le denomina "desplazamiento químico" (32, 33).

Cada núcleo de ¹H otorga una amplia gama de información ya que de manera suimultánea son detectados varios metabolitos y neurotransmisores; la localización o resonancia de los compuestos más relevantes en partes por millón (ppm) se resumen en la siguiente tabla (34):

Metabolito	Pico (ppm)	Descripción
N-Acetilaspartato (NAA)	2.02	Marcador de densidad e integridad neuronal y viabilidad axonal.
Colina (Cho) 3.22		Componente del metabolismo de fosfolípidos de membrana celular, refleja proliferación celular.
Creatina (Cr)	3.02 y 3.94	Utilizado como referencia, debido a que es el metabolito más estable y constante en sus valores.
Mioinositol (ml)	3.56	Marcador glial en los astrocitos, producto de la degradación de mielina, refleja volumen celular.
Lactato (Lac)	1.33 y 4.1	Indicador de glicolisis anaeróbica
Lípidos	0.8 - 1.3	Indican necrosis y/o disrupción de la vaina de mielina.
Alanina	1.49	Aminoácido no esencial presente en el cerebro.

A pesar del éxito de la ERM clínica y la disponibilidad de la ERM en la mayoría de los equipos de resonancia magnética en los hospitales, esta técnica no es utilizada tan ampliamente como podría serlo (35, 36).

Puesto que las neoplasias primarias son propensas a infiltrar el tejido cerebral circundante, el examen de áreas fuera de la zona de realce del tumor han probado ser promisorias (37), estas mediciones son obtenidas con espectroscopia.

El índice de cho/NAA mayor a 1.0 en el tejido cerebral residual o circundante a una lesión tumoral ya conocida, es el biomarcador de espectroscopia mas comúnmente aceptado y ha mostrado una exactitud de hasta el 100% para representar el tejido tumoral (37).

Varios puntos de corte han sido sugeridos para este biomarcador, aunque no hay todavia un consenso. Por ejemplo, algunos estudios han reportado un indice de Cho/NAA de más de 1.90 con un tiempo de eco largo (sensibilidad 83%; especificidad 88%) como punto de corte para lesiones de alto grado (38).

Otros biomarcadores de espectroscopia que han sido propuestos para lesiones gliales incluyen el porcentaje de lípidos, el índice de lactato/creatina, colina/creatina, myoinositol/Naa y N-acetilaspartato/creatina (39, 40). La presencia de lípidos y/o lactato en gliomas no tratados sugiere la presencia de un tumor grado IV (41).

La figura 2 muestra la utilidad de la ERM, en el diagnóstico del GBM.



Fig. 2 - Espectroscopia multivoxel, A) mapa de color del indice de Cho/NAA en una muestra de cuatro voxels, B) espectros que muestran la elevación del índice de Cho/NAA en diferentes regiones: en el interior de la zona con realce del tumor (8.0), en el sitio de edema peritumoral (4.69) dando evidencia de un patron de infiltración peritumoral. El estudio histopatologico reporto una lesion glial grado IV (glioblastoma). (Imagenes procesadas en la Unidad de RM del hospital Médica Sur).

Importancia de la Medición de Lipidos y Lactato en el Tejido Tumoral

La presencia de lípidos ha sido correlacionada con la formación de necrosis donde la muerte celular resulta en el rompimiento de membranas.(42) Este aumento en la concentración de lípidos ha sido observado en tumores de alto grado, particularmente en el glioblastoma multiforme.(43)

Una publicacion reciente demostro aumento en la concentración de lipidos casi exclusivamente en los gliomas de alto grado, y muy raramente presente en gliomas de bajo grado. Estas regiones han coincidido ademas con zonas que muestran aumento de volumen cerebral relativo, demostrado también con RM.(44)

La metodología para cuantificar la concentración de grasa con espectroscopia en el higado, se ha demostrado que puede sustituir sustituir a la biopsia hepática para cuantificar la grasa.(45) Esta misma metodología de cuantificacion se utilizara en el tejido cerebral esperando una alta correlacion con la cuantificacion que se realice en la pieza quirurgica utilizando metodos de morfometria por el departamento de neuroquimica del INNN.

El lactato, producto final de la glucolisis no oxidativa ha sido correlacionado con niveles de oxigeno bajos o hipoxia en el tejido tumoral.(46) Los gliomas se caracterizan por presentar extensas zonas de hippoxia y necrosis.(47) La hipoxia es un promotor poderoso de la angiogenesis y de la infiltracion tumoral atraves de la activacion de genes criticos para estas funciones.(48-51) La angiogenesis condicionada por la hipoxia condiciona un circulo vicioso degeneracion de microvasculatura inmadura y anormal que no puede funcionar adecuadamente y condiciona mas hipoxia tumoral, resultando en necrosis.(52)

La aplicación clinica de la medición de lactato y lipidos esta orientada por tanto a la diferenciación del glioma en bajo y alto grado. Esta es una información metabólica que va más alla de la imagen anatómica.

Coeficiente de Difusión Aparente obtenido con secuenia de difusión

La difusion, es un método de adquisición de imágenes que evalúa el movimiento microscópico de las moléculas de agua. Normalmente, este movimiento es desorganizado (o aleatorio) y se le denomina "movimiento browniano" (40).

Para generar una imagen de difusión, el equipo de resonancia aplica un gradiente magnético en una determinada dirección espacial que sirve para realinear las moléculas de agua estacionarias, haciendo que éstas emitan una señal. Así tras la aplicación de gradientes sensibles a la difusión, el agua normalmente en movimiento se evidencia como una pérdida en la intensidad de señal, y se observa un incremento en la intensidad de señal en aquellas regiones en las que las moléculas de agua permanecen estacionarias (53, 54).

Para cuantificar el grado de movimiento del agua, es necesario procesar las imágenes de difusión y obtener mapas del denominado "coeficiente de difusión aparente" (CDA) (31), definido también como la longitud del camino promedio de difusión de agua dentro de cada voxel, y está determinado por las barreras del tejido a la difusión en una escala de aproximadamente 10 mm.

El CDA es inversamente proporcional a la densidad celular, presumiblemente a causa de la tortuosidad del espacio intersticial y la limitación resultante en el movimiento del agua (55-57). La figura 3 muestra un ejemplo de difusión y CDA en las mediciones del tálamo derecho (las imágenes en RM se presentan invertidas, es decir, de frente al radiólogo) de un paciente con un tumor primario del sistema nervioso.



Fig. 3 – Mapas de difusión (A) y del coeficiente de diffusion aparente (B) que muestra una lesion intraaxial con un efecto de masa significativo y que se encuentra ademas rodeada de edema. Se observan dos pequeños ROIs (regions de interes) una colocada en el sitio del tumor y otra colocada en el sitio correspondiente del cerebro contralateral. El analisis cuantitativo mosto un CDA elevado en el sitio del tumor, lo que correspondió con un glioblastoma multiforme. *(Imagenes procesadas en la Unidad de RM del hospital Médica Sur).*

Hay una correlación negativa entre el CDA mínimo y la celularidad tumoral que ha sido verificado en histología en una amplia gama de tumors incluyendo a gliomas de alto y bajo grado, linfoma, meduloblastoma, meningioma y metástasis (57-61).

El valor promedio de CDA en sustancia blanca de tejido cerebral normal ha sido reportado en $0.764\pm0.063~(10^{-3})~mm^2/s$ (rango, 0.653-0.855) y para el tejido tumoral en GBM en $0.963\pm0.119~(10^{-3})~mm^2/s$ (rango 0.768-1.135).(62)

En los gliomas, se han reportado valores de CDA en por debajo de un rango de 1.7 a 2.5 para distinguir gliomas de alto y bajo grado (41, 63). Aunque el valor del CDA de gliomas de alto grado se ha demostrado por debajo de los de gliomas de bajo grado, hay un traslape sustancial (64); por lo tanto los mapas de CDA únicos, son insuficientes para predecir el tipo y grado de neoplasias gliales (65).

El CDA y la celularidad tumoral varían ampliamente dependiendo del grado del tumor (especialmente de alto grado) y por hallazgos relacionados (presencia en la lesión de necrosis, hemorragia y calcificaciones) (66). Esta heterogeneidad del CDA puede limitar la utilidad de la difusión como un

subrogado de la histopatología pero deja la posibilidad que el CDA pueda ayudar a sub estratificar los tumores dentro de los grados ya conocidos (40).

La difusión podría ser muy útil en el seguimiento de la respuesta al tratamiento y recurrencia subsecuente en pacientes individuales debido a que la radiación que es citotóxica, y la quimioterapia reducen ambas la celularidad tumoral y por lo tanto aumentan el CDA dentro de un área definida del tumor (67, 68).

Anisotropia fraccional cuantificada con secuencia Tensor de Difusión

La secuencia tensor difusión (DTI) permite la medición de la restricción de la difusión del agua en el tejido cerebral para producir imágenes de tractos nerviosos en lugar de utilizar los datos solamente para valorar el color o contraste de los pixeles de una imagen axial; esta secuencia permite la medición cuantitativa del movimiento de las moléculas en un espacio de 3 dimensiones.(69) Inicialmente un índice simple fue calculado de la imágenes ponderadas en difusión,(70) conocido como, este es ADC obtenido de una imagen con dirección perpendicular al movimiento del agua. El ADC desafortunadamente depende de la dirección escogida para realizar las mediciones.(71)

La anisotropía fraccional (FA) es aceptado como una medida de la direccionalidad de la difusión del agua (forma del tensor de difusión en cada voxel) del cerebro.(72) Esta FA puede apoyar a determinar el grado de infiltración de sustancia Blanca por el tumor, y provee evidencia de la degeneración de tractos distales a los sitios con tumor (degeneración Walleriana), esto cambia el pronostico de un paciente que va a someterse a una resección extensa de un tumor, con lo que se produce un daño mínimo o nulo al paciente al observar que las fibras tienen ya un daño irreversible y no hay déficits adicionales que puedan surgir.(73)

La fracción de anisotropía se basa en la varianza normalizada de los valores centrales de los vectores principales de la difusión (correspondiente a los eigenvectores del tensor), su valor varía entre 0 (difusión isotrópica) y 1 (anisotropía infinita).(74) La difusión es anisotrópica en los tractos de sustancia blanca, debido a que las membranas axonales y las vainas de mielina representan barreras al movimiento d las moléculas de agua en direcciones no paralelas a su propia orientación. Una FA reducida (difusión del agua paralela a los tractos axonales) es indicativo de degeneración axonal y una difusión aumentada (movimiento del agua perpendicular a los tractos axonales) es asociado con cambios en el contenido de agua, en disrupción y rompimiento parcial de la citoarquitectura tisular, o bien de procesos de desmielinización.(75) Basados en reportes previos, valores de FA > 0.2 han sido considerados como indicativos de algún grado de orientación estructural dentro de los axones mielinizados.(76)

Utilizando los valores de FA del tensor de difusión y con un post procesamiento de imágenes, es posible obtener una reconstrucción tridimensional (3D) de los tractos cerebrales, un procedimiento llamado *tractografía*. Se han designado 3 colores básicos que pueden decir al observador cómo las fibras están

orientadas en un sistema de coordenadas 3D: esto se denomina "mapa anisotrópico". El software codifica los colores de esta manera: el rojo indica las direcciones en el eje X: de derecha a izquierda o de izquierda a derecha, el verde indica las direcciones en el eje Y: posterior a anterior o de anterior a posterior, azul indica las direcciones en el eje Z: dirección de pies a cabeza o viceversa, ver figura 4. La tractografía es útil para proporcionar orientación en los procedimientos neuroquirúrgicos (77), tales como biopsias y tratamientos quirúrgicos, mediante la representación previa a la operación de tractos importantes de sustancia blanca, la visualización de la orientación de los tractos de fibras y su relación con el tumor puede reducir el riesgo de lesiones durante la realización de procedimientos quirúrgicos y biopsias.



Fig. 4 – Plano axial que muestra un mapa direccional de anisotropia (A) en un paciente con un glioma de bajo grado. El mapa de fracción de anisotropia muestra una reducción cuantitativa del traco geniculocalcarino adyacente al ventrículo izquierdo (B). Se observa una tractografía (C) que muestra la disrupción de alguna fibras del tracto mencionado (flecha roja). (Imágenes procesadas en la Unidad de RM del hospital Médica Sur).

Cuando un tumor esta presente y condiciona desplazamiento del tejido cerebral, edema peritumoral o infiltración, los tractos de sustancia Blanca pueden caracterizarse de la siguiente manera: desplazados, si mantienen su anisotropia normal relativa al tracto colateral correspondiente pero estan situados en una localizacion anormal con alteración en el mapa de color; edematoso, si mantiene su anisotropia normal y orientación pero demuestra alta señal en la secuencia T2; infiltrado si tienen anisotropia reducida pero pueden aun identificarse en el mapa; e interrumpidas, si la anisotropia esta marcadamente reducida asi que el tracto no puede ser identificado en su orientación (77).



Figura 5. Ejemplo de tractografía fusionada con imágenes de RM funcional BOLD (imagen color naranja señalada por flechas). Se observa la activación de la zona motora correspondiente a la pierna derecha (origen en el hemisferio izquierdo) la cual se encuentra desplazada en sentido posterior y lateral debido a la presencia de un tumor (letra T), que también condiciona desplazamiento de la vía piramidal ipsilateral (cabezas de flecha). Se visualizan en sentido anterior otros tractos correspondientes a la circunvolución del cíngulo y fascículo longitudinal inferior. (Imágenes procesadas en la Unidad de RM del hospital Médica Sur).

Volumen Cerebral Relativo en perfusión cerebral

La perfusión cerebral con administración dinámica de contraste ofrece información hemodinámica cerebral que complementa la información anatómica del estudio convencional de RM (31). La perfusión cerebral de contraste explota los cambios de señal que acompañan al paso de un agente de contraste paramagnético a través del sistema vascular cerebral y puede usarse para obtener información del volumen y del flujo sanguíneo (78). Así, puede ser una técnica de investigación para evaluar lesiones tumorales intracraneales y poder correlacionar sus resultados con los de estudios histopatológicos.

El radiotrazador que se utiliza como contraste puede ser endógeno (agua arterial) o exógeno (gadolinio) y puede ser difusible o no. El gadolinio se usa más frecuentemente en imagen de perfusión porque se acorta el tiempo de obtención de imágenes (tiempo de relajación) y requiere herramientas del procesado de imágenes muy accesibles.

Para medir el flujo sanguíneo cerebral se adquieren múltiples imágenes de varios planos del parénquima cerebral en intervalos de aproximadamente 1s, antes, durante y después de la inyección del agente de contraste por vía endovenosa (79).

El cambio de señal en cada voxel se utiliza para calcular el volumen sanguíneo cerebral relativo (rCBV, por sus siglas en inglés) de ese voxel, que luego se puede mostrar como un mapa de colores o un gráfico de la variación de intensidad de la señal en una zona determinada en el tiempo (curva tiempointensidad), ver figura 4. Las imágenes potenciadas por perfusión permiten visualizar la angiogénesis, proceso esencial para el crecimiento neoplásico (40). En pacientes que reciben terapias anticancerosas, la perfusión RM puede ser un método no invasivo para ver cambios en el volumen sanguíneo cerebral relativo durante el tratamiento y puede usarse como monitorización de la eficacia terapéutica (80).

La angiogénesis inducida por un tumor resulta en alteración estructural de los vasos sanguíneos que tienden a ser mas permeables y por tanto tienen parámetros de permeabilidad aumentada en las imágenes de perfusión (40).

Las neoplasias primarias como los gliomas tienen un volumen tumoral relativo (rTBV) que tiende a aumentar con el grado histológico. Este grado, se correlaciona con el volumen sanguíneo (81, 82).

Se ha publicado un punto de corte para el rTBV de 1.75 sugerido como umbral para distinguir neoplasias de alto y bajo grado, donde el rTBV es el cociente del volumen sanguíneo máximo comparado con una región de interés en sustancia blanca del hemisferio contralateral.



Fig. 6. Perfusión cerebral en un paciente con un glioma de alto grado. A, imagen de referencia en T2 que muestra la localización de un ROI en el area sólida de un tumor. B, mapa de perfusión que muestra el rTBV elevado en 3.07, compatible con lesión glial de alto grado. C, Curva de perfusión que muestra la región de interés en color rosa, comparada con el tejido cerebral normal. El tumor carece de barrera hematoencefálica por lo que hay fuga de contraste durante la inyección del bolo de contraste. (Imágenes procesadas en la Unidad de RM del hospital Médica Sur).

Evidencia del mejor biomarcador por RM

Algunos estudios han comparado el desempeño diagnóstico de los biomarcadores de RM en la "gradación" de tumores cerebrales; estos resultados muestran cual método de RM permite una detección mas temprana, lo que puede utilizarse en la evaluación de la respuesta a radioterapia. De acuerdo a estas publicaciones, los biomarcadores de espectroscopia detectarían de forma mas temprana los tumores (83), seguidos por la perfusión cerebral (84) y la etapa final corresponde al realce con gadolinio (85). La Figura 7, muestra esta relación de biomarcadores avanzados de RM.



Fig. 7. Utilidad de las secuencias avanzadas de RM en el diagnóstico de tumores cerebrales.

Evolución clínica de los tumores gliales

Los GBM, son los tumores cerebrales primarios más frecuentes y agresivos. Causan globalmente un 2% de las muertes por cáncer. Estos tumores poseen una alta tasa de recidiva local tras tratamiento quirúrgico, progresando localmente, lo que finalmente termina causando la muerte del paciente. Se diseminan principalmente a través de la sustancia blanca o por vía líquido cefalorraquídeo. A diferencia de otras neoplasias malignas avanzadas, éstas rara vez reportan metástasis a distancia.(86)

Sobrevida en GBM

A pesar de terapias estándar como la cirugía, radioterapia y quimioterapia y a pesar del considerable esfuerzo que se realiza actualmente para optimizar estos tratamientos, la sobrevida de los pacientes con gliomas de alto grado tiene rangos de 8-9 meses a 2-3 años. Tradicionalmente se admitía que presentaba una supervivencia media, a los dos años, de aproximadamente 10 por ciento. El tratamiento con un abordaje multidisciplinario incluye cirugía, radioterapia y quimioterapia. Con uno de estos

protocolos de tratamiento, se ha alcanzado recientemente una supervivencia del 25% y aún más en los casos en los que se consiguió una excisión completa. Aunque el pronóstico es malo se han conseguido avances en los últimos años.(87) Lo ideal es que en el tratamiento participen especialistas en neurocirugía, neurología y oncología. El avance terapéutico que más reciente que ha impactado en el tratamiento de esta enfermedad sin duda es el uso de Temozolamida, un nuevo agente alquilante que proporciona una sobrevida media de hasta 14.83 meses comparado con 14.67 meses en el grupo que recibe radioterapia solamente o quimioterapia convencional.(88, 89)

Tratamiento del GBM en el INNN

El tratamiento de quimio/radioterapia utilizado en el INNN para pacientes con GBM esta protocolizado de la manera siguiente: Radioterapia estereotáctica fraccionada 60 Gy a la periferia, 30 sesiones de 2 Gy, delineando la enfermedad macroscópica en secuencia T1 con contraste, el edema perilesional en T2 y un márgen de 20 mm para conformar el PTV (planning Treatment Volume). Se les indica prednisona 50 mg iniciales y en reducción progresiva a lo largo de un mes, iniciándolo el día de fin de tratamiento. Quimioterapia: Temozolamida calculada a 75 mg/m²/d los días de radioterapia, y al terminar calculada a 150-200 mg/m²/d por 5 días por ciclos de 28 días. Cinco ciclos en total. En caso de que el paciente no pueda pagar el medicamento se les administra Carboplatino-Vincristina después de terminar la radioterapia.(90)

Definición de Respuesta a Tratamiento

Una respuesta positiva al tratamiento será asignada a aquellos pacientes que sobrepasen la media del análisis de sobrevida y cumplen con los criterios del análisis de partición recursiva (RPA) el cual en su ultima revisión, utiliza tres grupos pronósticos para GBM considerando: edad, estado de desempeño, extensión de la resección, y función neurológica.(91)

La media del analisis de sobrevida que será de 14.2 meses para los pacientes del INNN tratados con temozolamida e incluidos en el estudio.(92)

Adicionalmente, en el INNN, se valora la respuesta al tratamiento dada por la imagen, donde una buena respuesta es el no crecimiento del tumor en las variables que componen el índice de edema tumoral. (90)

2) PLANTEAMIENTO DEL PROBLEMA

Planteamiento del problema

Aunque la evaluación con RM del GBM constituye la herramienta fundamental con la que el neurocirujano realiza la planeación de la resección quirúrgica, la planeación de la radioterapia y la

evaluación de la respuesta al tratamiento. No existe un consenso internacional de cual es el biomarcador de RM que ofrece el mejor valor pronostico de respuesta al tratamiento.

En el depto. de radioneurocirugía del INNN se utilizan 3 parámetros numéricos de la RM convencional para valorar la respuesta: volumen tumoral en T1, volumen tumoral en T2, y el índice de edema tumoral.

La literatura especializada de imágenes de RM sugiere que al menos 8 biomarcadores avanzados de RM podrían ser efectivos en el pronostico: índice de Colina/Naa, % de grasa en el tejido tumoral, índice de lactato/creatina, índice de mioinositol/creatina, índice de Naa/creatina, volumen sanguíneo cerebral relativo, coeficiente de difusión aparente y anisotropia fraccional.

El contar con biomarcadores de sobrevida institucionales permitirá diseñar una mejor herramienta pronostica que permita una terapia personalizada a las necesidades del paciente.

Pregunta de Investigación

¿Cuál es el biomarcador de resonancia magnética con mejor valor pronóstico para el calculo de sobrevida en pacientes del INNN con glioblastoma multiforme?

3) HIPÓTESIS

HIPÓTESIS

Ha: Hipótesis alterna.

Las mediciones de espectroscopia muestran un mejor valor pronóstico para el calculo de sobrevida comparados con los de difusión, perfusión, y tensor de difusión en pacientes con glioblastoma multiforme.

Ho: Hipótesis nula.

No existe diferencia significativa en los valores de difusión, perfusión, espectroscopia y tensor de difusión para el calculo de la sobrevida de pacientes del INNN con glioblastoma multiforme.

Hipótesis de trabajo

Las mediciones de metabolitos de espectroscopia, presentar mejor valor pronostico para el calculo de sobrevida de pacientes con GBM, comparados con los de perfusión y tensor de difusión.

4) OBJETIVOS

Objetivos

Objetivo General

Identificar el biomarcador de RM que muestre el mejor valor pronóstico de sobrevida en pacientes con glioblastoma multiforme.

Objetivos Específicos

- Identificar el grado de asociación (coeficiente de correlación) y realizar una evaluación completa de pruebas diagnósticas para cada uno de los biomarcadores de RM avanzada mas comúnmente reportados en la literatura: coeficiente de difusión aparente, volumen cerebral relativo, índice de colina/N-acetilaspartato y fracción de anisotropía.
- Evaluar otros biomarcadores adquiridos con las secuencias avanzadas de RM: volumetría 3D, lípidos, lactato, mioinositol, índice de infiltración tumoral;(72) y en las convencionales: volumen tumoral en secuencias T1 y T2 e índice de edema tumoral.
- Realizar análisis multivariado, de regresión múltiple y logística para definir modelos pronósticos con las variables disponibles.

Objetivos Secundarios

 Determinar la correlación entre los diferentes biomarcadores y los parámetros de evaluación clínica (v.gr. escala de Karnofsky).

5) JUSTIFICACIÓN

Justificación

Magnitud del problema

El GBM es el tumor que causa la mayor parte de las consultas y es el principal motivo de intervención quirúrgica en los pacientes derechohabientes del INNN.(93, 94) El GBM afecta no solamente a sujetos en la 5^ª y 6^ª década de la vida principalmente, en el INNN el 30% de los pacientes con GBM son menores de 45 años de edad lo que representa población económicamente activa.(93)

El tumor es rápidamente progresivo con una historia natural que lleva al fallecimiento del paciente al año del diagnóstico en el 90% de los casos. A pesar de los avances mundiales en investigación no se ha modificado significativamente la sobrevida tras 4 décadas de investigación.(93)

Viabilidad

La investigación es viable, evaluando pacientes del centro de referencia más grande del país para esta enfermedad; el INNN proporciona los requerimientos tecnológicos, la infraestructura física y personal capacitado para tratar a estos pacientes. El protocolo propone la colaboración de 5 departamentos del instituto: neuroradiología, neuroquímica, neuropatología, neurocirugía y neuroradiocirugía.

Vulnerabilidad

El protocolo no escapa a la posibilidad que eventualmente pudiera ser modificado en el flujograma de pacientes si el seguimiento clínico considera una evaluación menos estrecha de los pacientes.

Trascendencia

Este sería el primer estudio de esta naturaleza en Mestizo-Mexicana, esta acotación es importante ya que la mayoría de estudios previos han sido desarrollados principalmente en población caucásica.(8) Este es un protocolo de investigación propuesto por un radiólogo, alumno de doctorado en el INNN, y con la revisión del jefe del Depto. De Radioneurocirugía y del jefe de neuroradiologia. Los resultados del proyecto podrán apoyar el seguimiento de pacientes del INNN y para evaluar la respuesta a la radioterapia.

Año	Autor	Titulo	Ptes.	Marcadores
2010	Senft C	 Influence of imri-guidance on the extent of resection and survival of patients with glioblastoma multiforme 	43	T1 post Gd y MRS
2010	Tugcu B	2. Efficacy of clinical prognostic factors on survival in patients with glioblastoma	50	Realce en T1 post Gd
2010	Allahdini F	3. Evaluating the prognostic factors effective on the outcome of patients with glioblastoma multiformis: Does maximal resection of the tumor lengthen the median survival?	35	Volumetria en T1 post Gd
2010	Yamasaki F	4. Glioblastoma treated with postoperative radio- chemotherapy: Prognostic value of apparent diffusion coefficient at MR imaging	33	ADC, apparent difussion coefficient
2009	Wang CH	 Prognostic significance of growth kinetics in newly diagnosed glioblastomas revealed by combining serial imaging with a novel biomathematical model 	32	Volumetria en T1 post Gd y edema en T2
2008	Hirai T	6. Prognostic value of perfusion mr imaging of high- grade astrocytomas: Long-term follow-up study	31	rCBV, volumen cerebral relativo
2007	Murakami R	7. Prognostic value of pretreatment quantitative diffusion-weighted MR imaging	79	ADC, apparent difussion coefficient
2006	Pallud J	8. Prognostic value of initial magnetic resonance imaging growth rates for world health organization grade ii gliomas	143	Volumetria en T1 post Gd
2005	Pope WB	9. MR imaging correlates of survival in patients with high-grade gliomas	110	Volumetria en T1 post Gd y edema en T2
2004	Kurimoto M	10. Impact of neuronavigation and image-guided extensive resection for adult patients with supratentorial malignant astrocytomas: A single- institution retrospective study	76	T1 con Gd en el postoperatorio temprano
2004	Li X	11. Identification of mri and 1h mrsi parameters that may predict survival for patients with malignant gliomas	39	T1 post Gd y MRS

Revisión de PubMed sobre biomarcadores pronósticos en RM para Glioblastoma (hasta 2010)

Estos artículos muestran evidencia de la falta de evaluacion conjunta de biomarcadores de RM, los cuales son de rápida adquisición y bajo costo.

6) METODOLOGÍA

a) Diseño

Estudio prospectivo longitudinal de seguimiento de pacientes con diagnostico de GBM.

b) Población y muestra
 Población blanco Pacientes Mestizo-Mexicanos de cualquier género y edad con evidencia de GBM.

Población elegible

Pacientes portadores de GBM que ingresen para tratamiento al Instituto Nacional de Neurología y Neurocirugía de la ciudad de México.

Población estudiada

Pacientes con GBM, demostrado por biopsia, que cumplan con los criterios de inclusión y acepten firmar el consentimiento informado.

Tamaño de la muestra

El calculo del tamaño de muestra se realizó utilizando el software G*Power v3.1.3 (Franz Faul, Edgar Erdfelder, Albert-Georg Lang, and Axel Buchner, 2006, 2009).(95)

Se escogio el método de regresión linear múltiple, y tomando en cuenta los 4 biomarcadores de básicos principales: coeficiente de difusión aparente, volumen cerebral relativo, índice de Cho/NAA y Fraccion de anisotropia, se obtuvo el siguiente output:

F tests - Linear multiple regression: Fixed model, R² deviation from zero

Analysis:	A priori: Compute required sample size			
Input:	Effect size f ² =	0.35	5	
	a err prob	=	0.05	
	Power (1- β err prob)		=	0.95
	Number of predictors	;	=	4
Output:	Noncentrality parame	eter λ	=	20.6500000
·	Critical F	=	2.542	9175
	Numerator df	=	4	
	Denominator df		=	54
	Total sample size		=	59
	Actual power	=	0.953	6733

El tamaño de la muestra obtenido es de 59 pacientes.

Criterios de selección del estudio

Criterios de inclusión

1. adultos mayores de 18 años,

- 2. de cualquier género,
- 3. con diagnóstico por biopsia esterotáxica de glioblastoma multiforme (astrocitoma IV de la OMS).
- 4. Consentimiento informado debidamente firmado por el paciente o familiar responsable.

Criterios de exclusión

- 1. Contraindicaciones para recibir radiocirugía o quimioterapia,
- 2. Haber recibido tratamiento anterior,
- 3. Pacientes con algún tipo de patología o tratamiento previo que implique algun tipo de inmunosupresión.
- 4. Cualquier contraindicación para un estudio de RM.

Criterios de eliminación

- 1. Presencia de tumor cerebral con otra estirpe histológica.
- 2. Rechazo del consentimiento informado
- 3. Registro de variables incompleto y

Estudio de RM no valorable por artificios durante la adquisición del estudio o el postproceso.

c) Variables

Definición de variables y unidades de medición

Variable	Definición Conceptual (C) y Operacional (Op)	Instrumento	Tipo de variable	Codificación
Edad	C: Tiempo transcurrido en años desde el nacimiento. Op: cálculo de los años. basados en acta de nacimiento.	Historia Clínica	Continua	Años
Género	C: aspecto masculino o femenino de la persona. Op: fenotipo de la persona correspondiente a masculino o femenino.	Historia Clínica	Categórica Nominal	0: masculino 1: femenino
Puntuacion Karnofsky	C: evalua el estado de desempeño. Op: califica del 100 a 0, donde 100 es perfecta salud y 0 es muerte.	Historia Clínica	Categórica Nominal	100 a 0
Grado de resección	C: presencia de tumor después de la cirugía. Op: califica el resultado postquirurgica de la reseccion tumoral.	Historia Clínica	Dicotomica	0: sin tumor 1: residual
Respuesta al tratamiento	C: control tumoral. Op: pacientes que sobrepasen la media de sobrevida para pacientes del INNN con GBM.	Historia clinica	Dicotomica	No. días
% de grasa tumoral	C: Concentracion de grasa en el tejido. Op: medición del porcentaje de	Espectroscopia	Continua	Porcentaje (%)

	grasa.			
Indice de Colina /creatina	C: medición indirecta del contenido de lípidos en membranas celulares, marcador de proliferación. Op: de dividir las concentraciones absolutas de colina entre creatina	Espectroscopia	Continua	adimensional
C: medición indirecta del metabolismo anaerobio a nivel celular, marcador de hipoxia celular. /creatina Op: de dividir las concentraciones		Espectroscopia	Continua	adimensional
Indice de Naa /creatina /creatina /creatina		Espectroscopia	Continua	adimensional
Indice de Mioinositol /Naa	C: medición indirecta de la actividad de fibroblastos en SN, marcador de gliosis. Op: de dividir las concentraciones absolutas de mioinositol entre Naa	Espectroscopia	Continua	adimensional
Coeficiente de Difusión Aparente	C: medición del movimiento de las moleculas de agua en la celula, marcador de edema citotóxico. Op: medición del área de desplazamiento de los protones en los átomos de hidrogeno de las moléculas de agua por unidad de tiempo.	Secuencia de Difusión	Continua	mm²/seg
Anisotropia Fraccional	C: medicion de la direccional de las moleculas de agua en tractos cerebrales definidos. Op: medicion del 0 al 1 (minimo- maximo) de la direccionalidad especifica de c/ tracto cerebral.	Secuencia tensor de difusión	Continua	adimensional
Volumen tumoral relativo	C: perfusión sanguinea tumoral por unidad de tiempo. Op: comparación relativa del volumen sanguineo cerebral entre tejido cerebro sano y tejido solido infiltrado con GBM.	Secuencia de perfusión	Continua	Porcentaje (%)
Volumen tumoral post gadolinio	C: volumen de tumor cerebral viable.	Secuencia T1/SPGR post gadolinio	Continua	mm ³

	Op: medicion volumétrica del tejido tumoral que realza con contraste.			
Volumen de edema cerebral	C: volumen del edema cerebral tumoral. Op: medicion volumétrica del edema vasogénico por fuera del tejido viable (que realza con contraste), causado por el GBM.	Secuencia T2 sin gadolinio	Continua	mm ³
Indice de edema tumoral	C: relación entre volumen tumoral versus el edema que causa un tumor cerebral. Op: cociente de dividir el volumen tumoral que realza con contraste entre el volumen tumoral con su edema vasogénico asociado.	T1 post gadolinio / T2	Continua	mm ³

d) Análisis Estadístico

Todas las evaluaciones estadísticas se llevaran a cabo utilizando el software SPSS versión 17.0 (SPSS Inc., Chicago, IL). La significancia estadística se considerará con 1 valor de p < 0.05 (2 colas). La presentación de los datos se realizará siguiendo las guías para publicaciones de la Asociación Psicológica Americana.(96, 97)

Comparación de medias aritméticas y tamaño del efecto

Los valores de serán evaluados para distribución normal con la prueba de Kolmogorov-Smirnov y con la prueba de Levene statistics para verificar la presencia de homogeneidad de varianzas. La demostración de una distribución normal en las variables permitirá la aplicación de pruebas paramétricas, en el caso contrario se aplicarán por lo tanto pruebas no para métricas en la evaluación.

Se realizará análisis para determinar las frecuencias de las variables de estudio, así como medidas de tendencia central y dispersión para determinar el comportamiento normal o sesgado de la población.

Se realizará análisis bivariado mediante la prueba de Chi 2 o exacta de Fisher para variables categóricas y prueba de t de *student o* Mann–Whitney para variables numéricas.

La medias aritméticas serán comparadas con pruebas t de Student para muestras independientes. Para calcular el tamaño del efecto de cada variable, los valores de t y los grados de libertad (df) serán utilizados para calcular el valor de la "*d*" de Cohen's *d* y el tamaño del efecto "r"; utilizaremos la definición de un tamaño de efecto pequeño con valores de (d = 0.2), medio (d = 0.5), y grande (d = 0.8).(98) *Correlación*.

El grado de asociación entre las variables será valorado por los coeficientes de correlación de Pearson o Spearman según corresponda. Los valores del coeficiente de correlación "r" cercanos o por arriba a 0.75 indicarán 1 correlación buena o excelente. Para esta prueba, se considerarán datos significativos cuando p < 0.05 (una cola).

Análisis de sobrevida.

Se realizará prueba de Cox para análisis de supervivencia y regresión logística para determinar la influencia de variables.

El análisis de regresión de Cox(99) será realizado para determinar un modelo matemático que permita identificar cuales son los mejores biomarcadores pronósticos. Antes de comenzar el análisis de regresión, en caso de obtenerse un resultado no paramétrico de alguna de las variables dependientes se aplicará transformación logarítmica a las variables con el propósito de lograr distribución normal, una vez más la significancia estadística para esta prueba será considerada en p < 0.05.

Evaluación de Pruebas Diagnósticas con Curvas ROC

Se utilizarán curvas para características del operador receptor, con el propósito de evaluar el desempeño diagnóstico de los parámetros.(100) El área bajo cada una de las curvas ROC (AUROC, por sus siglas en inglés), serán interpretadas como el valor promedio de la sensibilidad para todos los valores posibles de especificidad, esta medida será considerada como el representante del desempeño global para cada variable;(101, 102) se obtendrán también los valores para error estándar, valor de *p* y los intervalos de confianza del 95% Cl. La respuesta al tratamiento será considerado el estándar de oro para el análisis de las curvas ROC. Se realizara el cálculo de los "puntos de corte" para cada variable y representarán los puntos previos y subsecuentes que rodean el umbral óptimo para maximizar la sensibilidad y la especificidad.

La precisión de cada marcador se ha determinado utilizando sistema tradicional para clasificar la puntuación: 0.90-1 = excelente (A), 0.80-0.90 = buena (B), 0.70-0.80 = regular (C), 0.60-0.70 = pobre (D), y 0.50-0.60 = fallida (F) (103). Los valores de sensibilidad, especificidad, valor predictivo positivo, valor predictivo negativo, y razones de verosimilitud positivas y negativas así como la precisión, serán también calculadas en este análisis. La significa es estadística para este análisis de curvas ROC se considera con *p* < 0.05 basado en la utilización de la prueba exacta de Fisher.

7) CONSIDERACIONES ÉTICAS

Esta investigación tiene un riesgo mayor al mínimo de la Ley General de Salud en Materia de Investigación para la Salud debido a que los sujetos serán expuestos a estudios de resonancia magnética funcional. El aspecto clínico del estudio se considera de investigación con riesgo mínimo tratándose de procedimientos comunes en exámenes físicos o psicológicos de diagnósticos o tratamiento rutinarios. Así mismo se apega a las pautas establecidas en la Declaración de Helsinski de 1983, en todos los casos los participantes del estudio firmarán un consentimiento informado.

El estudio garantiza los aspectos bioéticos inherentes a los estudios de investigación clínica como son confidencialidad de la información obtenida, principio de autonomía de los sujetos participantes, principio

de beneficencia y consentimiento informado, sin tener conflicto de intereses y permitiendo el acceso a documentos fuente en caso de ser solicitado por autoridades normativas.

A los pacientes se les comentará el tipo de estudio, la intención del mismo y se les solicitará autorización para ingresar al mismo, así como se especificaran detalles respecto al procedimieto de imagen y a la toma y el uso de muestras tanto de sangre periférica. Debido a que los pacientes con GBM pueden tener afección de las funciones mentales se explicara a detalle el protocolo al familiar responsable a fin de que no quede duda del objetivo, destino y uso de las muestras que se analizarán.

8) CONSIDERACIONES FINANCIERAS

Este protocolo no tiene implicaciones financieras adicionales.

Los análisis de la imagen representados por los post-procesos computacionales de la imagen, serán realizados sin costo adicional por el investigador principal Dr. Ernesto Roldan

Los pacientes incluidos, cubrirán el costo que corresponda a su tratamiento y consultas de acuerdo a su clasificación socioeconómica en el INNN.

a) Aporte Financiero

Sin aporte financiero.

b) Recursos con los que se cuenta

Recursos Humanos

- 1. Neurocirujano con Posgrado en radiocirugia, a cargo del ingreso y manejo de los pacientes en el departamento de radioneurocirugia en el INNN.
- 2. Tecnico radiologo, a cargo de realizar los ajustes al software y obtencion de secuencias, en el INNN.
- 3. Radiólogo a cargo del post procesamiento del analisis avanzado de tumores (ciego a pacientes), *el alumno del doctorado en ciencias medicas INNN proponente del protocolo, Dr. Ernesto Roldan.*
- 4. Patólogo, cualquier patólogo asignado a los procedimientos de diagnostico de las biopsias de cerebro, en INNN.
- 5. Neuroquimico, Dr. Camilo Rios, para procesos de morfometria y cuantificacion total de lípidos, en INNN.

-				
	Técnica	Método	Lugar	Responsable
1.	Inclusión de	Consulta	Departamento de	Personal Depto.
	pacientes y Tx		Radioneurocirugía	Radioneurocirugía
	Radiocirugía		INNN	_
2.	Procesamiento de la biopsia	Tinción convencional con hematoxilina y eosina	Departamento de patología, INNN	Personal Depto. Patología
3.	Cuantificación de	Morfometria	Departamento de	Alumno del Doctorado
	grasa en laminillas		Neuroquímica, INNN	en Ciencias Médicas
4.	Cuantificación de	RM con espectroscopia	Depto. Neuroradiologia	Alumno del Doctorado
	grasa no invasiva		del INNN	en Ciencias Médicas
5.	Análisis avanzado de	difusión, perfusión,	Depto. Neuroradiologia	Alumno del Doctorado
	tumores	espectroscopia, volumetría	del INNN	en Ciencias Médicas

Personal a cargo de las evaluaciones
Materiales

- Resonador 3T General Electric, modelo HDX Excite.
- Estación de trabajo para postproceso de espectroscopía, secuencias doble eco y secuencias T2 con saturación grasa.
- Computadora portátil
- Paquete estadístico SPSS v17.0
- Material y equipo del laboratorio para procesamiento de biopsia cerebral. Departamento de Patología, INNN.
- Cuantificación de la Grasa en las Biopsias Tenidas con Hematoxilina Eosina To Departamento de Neuroquímica, INNN.
- Software de GE para postproceso de imágenes medicas.
 - c) Recursos a solicitar

Ninguno

d) Análisis del costo por paciente

Sin costo adicional para el paciente.

9) CRONOGRAMA DE ACTIVIDADES

Nuevo Programa de Doctorado en Ciencias Médicas UNAM (8 semestres 2011-2014)

Actividad	1	2	3	4	5	6	7	8
Redacción de protocolo y presentación a Comité de Ética								
Inclusión de pacientes								
Captura de datos								
Análisis								
Resultados								
Envío de Artículos a Publicación								
Carta del editor, resultados de la revisión								
Presentación final de resultados ante el Comité de Evaluación del Posgrado de la UNAM								

Actividades programadas de acuerdo a la duración del programa

Procedimientos

Reclutamiento de pacientes

Se invitara a los pacientes con diagnóstico de glioblastoma multiforme, que hayan ingresado para tratamiento a departamento de radioneurocirugía del INNN y que debido a su condición clínica sean candidatos a tratamiento. Los pacientes deberán cubrir los criterios de inclusión y firmar un consentimiento informado. La muestra se empezara, a reclutar en el período en cuento sea aprobado el protocolo (se espera el mes de septiembre de 2012) y hasta alcanza el numero planeado de pacientes.

Evaluación clínica

Los pacientes a su ingreso contaran con una historia clínica completa que incluye los datos de identificación, y antecedentes familiares para enfermedades atópicas y alérgicas.

Durante la consulta de ingreso a radioneurocirugía se obtendrán *variables demográficas* consistentes en edad, sexo, peso, talla e índice de masa corporal (definido como peso/talla²).

Se realizará una historia clínica completa con exploración física general y neurológica, con documentación del uso de esteroides antiinflamatorios, tipo y dosis, enfermedades inmunosupresoras, antecedentes recientes de infección.

Escala de Karnofsky

De manera ya protocolizada se evalua la KPSS (Karnofsky Performance Status Scale) a los pacientes en el periodo preoperatorio, postoperatorio inmediato dentro de las primeras 24 horas de la cirugía y al mes de realizada ésta, y posteriormente cada 3 meses de ser posible hasta el fallecimiento.

Postprocesamiento y analisis de las imágenes de RM

Después de la evaluación clínica en el Departamento de Radiocirugía, el paciente será invitado a participar en el protocolo por el Dr. Sergio Moreno. La interpretación de las imágenes de Resonancia Magnética serán realizada por el departamento de neuroradiologia del INNN con el reporte radiológico que corresponda a cada paciente, adicionalmente el Dr. Ernesto Roldan podrá realizar el post procesamiento de algunas secuencias con motivo del análisis cuantitativo de las secuencias propuestas en su protocolo de doctorado.

El Dr. Moreno explicara a los pacientes los hallazgos por imagen y las rutas del tratamiento multidisciplinario que se realizara en el INNN con el medico tratante que le haya sido asignado.

Todos los pacientes deberán contar con RM preoperatoria con especial énfasis al T2 y al T1 con gadolinio.

La evaluación convencional de las imágenes incluye:

- ✓ Volumen tumoral en T1 post gadolinio.
- ✓ Volumen tumoral incluyendo el edema perilesional en T2.
- ✓ Indice de edema tumoral (volumen en T1 / volumen en T2)

El análisis avanzado del tumor, tratara de evaluar:

- ✓ Índices de Colina/Cr, Naa/Cr, Lactato/Cr y % de grasa tumoral con Espectroscopia.
- ✓ Volumen cerebral relativo (máximo) del tejido tumoral solido con Perfusión (rCBV).
- ✓ Coeficiente de difusión aparente con secuencia de Difusión (ADC).
- ✓ Valor de anisotropía fraccional del tejido tumoral con secuencia tensor difusión (FA).

La RM de cerebro se podrá realizar en los equipos de 1.5 y 3.0T HDxt GE Signa scanner (General Electric Healthcare, Milwaukee, WI) de la unidad de resonancia del INNN. Las imágenes serán adquiridas con una bobina para cráneo de 8 canales. Esta consideración es particualrmente importante, considerando que existen periodos durante el año que alguno de los equipos se encuentra en mantenimiento (fuera de servicio), por lo que los pacientes podrán ser evaluados en el magneto disponible sin retrasar la programación de cirugía y tratamiento de los pacientes.

El protocolo de adquisición de las imágenes se encuentra detallado en el apéndice No. 2 para ambos magnetos.

Cronograma para seguimiento de los pacientes con RM

Los pacientes serán evaluados cada 3 meses con el mismo protocolo de RM después de realizada la cirugía, y de forma previa a sus seguimientos en la consulta externa de radioneurocirugía de acuerdo al siguiente esquema.

Periodo	Basal	Cita radio reso	Citas en la consulta externa de radioneurocirugía después de la resección quirúrgica del tumor		
		3	6	9	12 mosos
		meses	meses	meses	12 1116365
	Para el tejido tumoral: Vol. Tumoral, Vol. Edema y el Indice tumoral	La evalu en 1 ROI	ación se de tejido	realizara i normal.	dealmente solo
Variables evaluadas	Para 3 ROIs diferentes en tumor, edema y tejido normal: rCBV, ADC, FA, Cho/Naa, Naa/Cr, Lipidos, lactato.	Si el tu completa en zona tumoral r	imor NO , se utiliz s que m esidual.	pudo te aran 2 R uestren e	ener resección Ols adicionales edema y tejido

Grado de Resección Postoperatoria

Después de la cirugía, misma que será realizada por el personal adscrito a Neurocirugía del INNN se procederá a determinar el grado de resección tumoral en el postoperatorio.

Grado de resección tumoral documentada con imagen

Se realiza mediante una TAC o RM postoperatoria con contraste dentro de las primeras 24 horas posteriores a la cirugía para clasificar la resección en 1 de 4 grados posibles:

- 1. GTR (Gross Total Resection) Resección macroscópica total.
- 2. NGTR (Near Gross Total Resection) Resección casi total, mínimo refuerzo con medio de contraste.
- 3. SGR (Subtotal Gross Resection) Resección subtotal, moderado remanente o descompresión.
- 4. BX (Biopsy) Biopsia o mínima resección tumoral sólo para diagnóstico.

Exámenes de laboratorio en sangre o suero.

Se obtendrán los exámenes de laboratorio preoperatorios según estipulan las guías de diagnóstico y tratamiento del INNN.

Manejo de la biopsia

La biopsia se procesara por el departamento de patología del INNN con técnicas convencionales.(104) Una vez teñida y posterior a su diagnostico histológico, la biopsia será transportada al Departamento de Neuroquímica del INNN, donde se realizará evaluación morfométrica,(105, 106) para cuantificación del porcentaje de biopsia en la laminilla.

Egreso y vigilancia del paciente.

El manejo del pacientes seguirá los estándares del Departamento de Radioneurocirugía del INNN.

Tratamiento

El tratamiento de quimio/radioterapia utilizado en el INNN para pacientes con GBM estará protocolizado de la manera siguiente:

Radioterapia estereotáctica fraccionada 60 Gy a la periferia, 30 sesiones de 2 Gy, delineando la enfermedad macroscópica en secuencia T1 con contraste, el edema perilesional en T2 y un márgen de 20 mm para conformar el PTV (Planning Treatment Volume). Se les indica prednisona 50 mg iniciales y en reducción progresiva a lo largo de un mes, iniciándolo el día de fin de tratamiento. Quimioterapia: Temozolamida calculada a 75 mg/m2/d los días de radioterapia, y al terminar calculada a 150-200 mg/m2/d por 5 días por ciclos de 28 días. Cinco ciclos en total. En caso de que el paciente no pueda pagar el medicamento se les administrara Carboplatino-Vincristina después de terminar la radioterapia.(90)





Respuesta a Radioterapia

Correlación Regresión lineal múltiple Regresión de Cox

Flujograma del analisis de datos



- 1. Coeficiente de difusión aparente,
- 2. Volumen tumoral relativo,
- 3. Anisotropia fraccional,
- 4. Volumetria en T1 y T2,
- 5. Indice de Colina/Naa,
- 6. Indice de Colina/Cr
- 7. Indice de Lactato/Cr
- 8. % de grasa con espectroscopia
- 9. % de Grasa con Morfometria,
- 10. Respuesta a tratamiento (sobrevida)
- 11. Edad
- 12. Sexo
- 13. Karnofsky



Morfometria en biopsia para cuantificacion de grasa

Cuantificación de grasa tumoral en la espectroscopia cerebral

10) RESULTADOS QUE SE ESPERAN OBTENER Y SU POSIBLE IMPACTO EN LA INVESTIGACIÓN BIOMÉDICA (CONTRIBUCIÓN DE LA INVESTIGACIÓN EN EL AVANCE DEL CONOCIMIENTO CIENTÍFICO).

Conocer el valor pronostico de nuevos biomarcadores de resonancia magnética, obtenidos de las evaluciones que se toman comúnmente a los pacientes pero que no se postprocesan:

- ✓ Coeficiente de difusión apararente,
- ✓ Volumen tumoral relativo
- ✓ Índice de colina/N-acetilaspartato
- ✓ Anisotropia fraccional

Con ello, es muy posible re-diseñar el estudio de Resonancia Magnetica de Cerebro que se lleva a cabo para estos pacientes en el INNN. Así como apoyar la planeación de los tratamientos de Radioneurocirugia.

11) SATISFACCIÓN DE UNA NECESIDAD DE SALUD CON EL DESARROLLO DE ESTE TRABAJO.

La atención de tumores cerebrales malignos, principalmente el glioblastoma multiforme, es la principal causa de consulta, se invierten recursos humanos y materiales en la atención de pacientes con una sobrevida corta.

Este estudio pretende obtener evidencia cuantitativa para apoyar la planeación de tratamientos y conocer el pronostico de sobrevida, lo que impacta en los tiempos de tratamiento y seguimiento.

12) MENCIONE QUE APORTARÁ EL DESARROLLO DEL TRABAJO A LA COMPRENSIÓN, PREVENCIÓN, DIAGNÓSTICO Y TRATAMIENTO A LAS AFECCIONES DEL SISTEMA NERVIOSO.

Comprensión

Ayudar a entender que las características por imagen del tumor pueden tener mayor valor para el pronostico y tratamiento que el resultado histológico inicial.

Prevención

Pacientes en quienes se midieran inicialmente, factores pronósticos pobres, ser mas agresivos en la planeación quirúrgica, de quimio y radioterapia.

Diagnostico

Utilizar los biomarcadores de imagen que recomienda la literatura internacional para una evaluación completa de tumores cerebrales.

Tratamiento

Tener una herramienta cuantitativa, que puede medir objetivamente la evolución de los pacientes, incluso antes que los datos clínicos (síntomas) sean reportados por el paciente (en el caso de recidiva tumoral).

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INSTITUTO NACIONAL DE NEUROLOGÍA Y NEUROCIRUGÍA MANUEL VELASCO SUÁREZ DIRECCIÓN DE INVESTIGACIÓN

> Insurgentes Sur 3877 Col. La Fama, C.P. 14269 Mexico, D.F., Tel. 55-28-80-36

Consentimiento informado

Fecha: _____ PROTOCOLO DE INVESTIGACIÓN No.: <u>84/11</u>

El departamento de Radioneurocirugía dirigido por el Dr. Sergio Moreno Jimenéz lo invita a participar en el protocolo de investigación:

Biomarcadores avanzados de resonancia magnética como factores pronósticos para la sobrevida de pacientes con glioblastoma multiforme.

Se está invitando a participar en este proyecto a los pacientes como usted con diagnóstico de glioblastoma multiforme (cáncer cerebral).

CONFIDENCIALIDAD

Si acepta contribuir con este proyecto debe saber que la información relacionada con su enfermedad y datos personales son confidenciales y solo será usada con fines de investigación. Los pacientes que sean incluidos en el estudio serán identificados por números sin relación a su expediente clínico por lo que su nombre y demás datos de índole personal no serán revelados. Esta información no será compartida con otras personas sin su consentimiento.

Todas las personas involucradas están comprometidas a guardar la información de manera confidencial y velar por la seguridad física, moral y psicológica de los participantes en este estudio.

OBJETIVO DEL ESTUDIO

Este proyecto estudiará la relación entre los parametros avanzados de la resonancia magnetica (difusión, perfusión, espectroscoía y tractografía) que se realizara a su cerebro como parte del diagnostico y seguimiento de su evolución clínica en este instituto. El diagnostico y tratamiento de cada paciente no será verá modificado o afectado.

PROCEDIMIENTOS

Si usted acepta participar en este proyecto se le pedirá que tenga una entrevista con los médicos encargados del estudio.

Se le harán preguntas en relación a su enfermedad y se recabarán esta información en un formato con fines de recolección de datos. Usted será diagnosticado y tratado como cualquier otro paciente del Instituto con diagnóstico de cáncer cerebral si acepta participar en este estudio.

BENEFICIOS PREVISTOS PARA LOS PACIENTES

Mediante este estudio no obtendra beneficio economico y su condicion de salud no mejorara como resultado de su participacion en mismo.

RIESGOS Y MOLESTIAS POSIBLES

No hay riesgos ni molestias adicionales a las del estudio de resonancia magnetica que su médico solicitó.

ALTERNATIVAS DE TRATAMIENTO

El participar en este protocolo no influirá en las indicaciones medicas de su tratamiento.

RECOMENDACIONES

Solamente las indicadas por su medico tratante.

OBLIGACIONES FINANCIERAS Y COMPENSACIÓN

Su participación en este estudio no tiene costo alguno para usted, no recibirá remuneración economica alguna por su participación.

ABANDONO DEL ESTUDIO

Usted puede abandonar o revocar su autorización a participar en el estudio en el momento que así lo deseé sin que esto interfiera con la atención médica recibida o que recibirá en este Instituto.

El médico responsable tiene la facultad de suspender su participación en el estudio con el fin de proteger su salud, si así lo considera necesario.

IDENTIFICACIÓN DE LOS INVESTIGADORES

Si tiene una pregunta o duda sobre cualquier asunto relacionado con este estudio por favor contacte a los médicos encargados del proyecto

Dr. Ernesto Roldan Valadez al telefono 5424-7230 o 5424-7231.

Dr. Sergio Moreno Jiménez al teléfono 56 06 38 22 extensión 4472 y 4473.

CONSIDERACIONES ÉTICAS

Este estudio se realizará bajo las recomendaciones de la Asociación Médica Mundial (conforme a la Declaración de Helsinki de 1964 y en apego a la Ley General de Salud, al Reglamento de la Ley General de Salud en Materia de Investigación para la Salud). Además los procedimientos de este estudio han sido revisados y aprobados por el Comité de Etica del Instituto Nacional de Neurologia y Neurocirugia.

FIRMA DE CONSENTIMIENTO INFORMADO

Manifiesto haber leído y entendido completamente el objetivo del estudio, haber tenido la oportunidad de hacer preguntas adicionales y recibido respuestas satisfactorias. Entiendo la razón por la que se requiere recopilar la información de este estudio así como el manejo que se le dará a la misma, con lo cual estoy totalmente de acuerdo.

Por este conducto, acepto participar en el presente protocolo de investigación titulado:

Biomarcadores avanzados de resonancia magnética como factores pronósticos para la sobrevida de pacientes con glioblastoma multiforme.

Nombre del paciente (Apellido paterno, materno y nombre(s))	Firma	
Teléfonos de contacto	Fecha (DD/MMM/AAAA)	

 Nombre del Testigo I
 Domicilio

 (Apellido paterno, materno y nombre(s))
 (Calle, N°, Colonia, Còdigo Postal)

 Fecha (DD/MMM/AAAA)
 Firma

Nombre del Testigo II (Apellido paterno, materno y nombre(s)) Domicilio (Calle, Nº, Colonia, Còdigo Postal)

Fecha (DD/MMM/AAAA)

Firma

Parentesco

Firma del Representante Legal (si aplica)

Nombre del Representante Legal (Apellido paterno, materno y nombre (s))

Telefonos de contacto

Firma

Fecha (DD/MMM/AAAA)

INSTITUTO NACIONAL DE NEUROLOGÍA Y NEUROCIRUGÍA

MANUEL VELASCO SUÁREZ DIRECCIÓN DE INVESTIGACIÓN



Insurgentes Sur 3877 Col. La Fama, C.P. 14269 Mexico, D.F., Tel. 55-28-80-36

Secuencias de RM para el Procotolo:

Biomarcadores avanzados de resonancia magnética como factores pronósticos para la sobrevida de pacientes con glioblastoma multiforme.

Fecha: Agosto 21 de 2012

PROTOCOLO DE INVESTIGACIÓN No.: 84/11

Objetivo: Este protocolo pretende estudiar los cambios por GBM en tejido tumoral, en edema y en tejido sano.

Protocolo de Secuencias para el Resonador de 3.0T *

Secuencias a realizar	Comentarios	Tiempo
Difusión	✓ Utilizar valor-b de 1000 s/mm², FOV=25.6 cm, matriz 128x128, grosor=5 mm, espaciamiento=1 mm, TE mínimo, TR suficiente para adquirir todos los cortes, NEX=3, ASSET.	~1:10
Secuencia T2 Cubo-sagital, isométrica	✓ FOV=25.6 cm, matriz=256x256, grosor=1 mm, TR=2500 ms, cortes suficientes para cubrir toda la cabeza, ajustar ETL para que el TE sea cercano a 120 ms, ARC.	~2:00
Secuencia FLAIR T2 Cubo- sagital	✓ FOV=25.6 cm, matriz=256x256, grosor=1 mm, TR=7000 ms, cortes suficientes para cubrir toda la cabeza, ajustar ETL para que el TE sea cercano a 120 ms, ARC.	~3:00
Tensor de Difusión (Tractografía)	✓ Utilizar valor-b de 700 s/mm², 35 direccionse de difusión, FOV=25.6 cm, matriz 128x128, grosor=3 mm, espaciamiento=0 mm, TE mínimo, cubrir todo el encéfalo y el cerebelo, TR suficiente para adquirir todos los cortes, NEX=1, ASSET.	~8:00
Perfusión	✓ EPI-GRE, FOV=25.6 cm, matriz 128x128, grosor=5 mm, espaciamiento=1 mm, TE=50 ms, TR suficiente para adquirir todos los cortes, 40 fases por corte, Tiempo de duracion aprox. 1:08 minutos para 20 cortes.	~1:10
Secuencia T1 3D isométrica axiala pre y post gadolinio-axial	✓ Fast SPGR, FOV=25.6 cm, matriz=256x256, phase FOV=0.75, TE Min Full, TI=450 ms, FA=15°, grosor=1 mm, cortes suficientes para cubrir todo el encéfalo y el cerebelo.	~8:00
Espectroscopia 3D Multivoxel	 Incluye zona del tumor y edema perilesional, si es posible alguna zona de tejido sano. FOV=16 cm, matriz=8x8, TE=144 ms, TR=1000 ms, NEX=1.6 ó 2. 	~8:30
1 Espectro univoxel para medición de grasa del tumor (zona solida) y 1 espectro univoxel para medicion de grasa en tejido sano (hemisferio cerebral contralateral).	✓ TE=30 ms, TR=1500 ms, vóxel=2x2x2 cm, 128 scans.	~4:00
	Total Minutos	~36:00

* Las secuencias de RM fueron planeadas con apoyo del Especialista en Aplicaciones Avanzadas M. en C. Rafael Favila (GE Healthcare, México).

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Secuencias de RM para el Procotolo:

Biomarcadores avanzados de resonancia magnética como factores pronósticos para la sobrevida de pacientes con glioblastoma multiforme.

Fecha: Agosto 21 de 2012

PROTOCOLO DE INVESTIGACIÓN No.: 84/11

IMPORTANTE: En caso que el magneto de 3.0 T no este disponible, las secuencias pueden adquirirse en el magneto de 1.5T con los siguientes parametros:

Protocolo de Secuencias para el Resonador de 1.5T *

Secuencias a realizar	Comentarios	Tiempo
Difusión	✓ Utilizar valor-b de 1000 s/mm², FOV=25.6 cm, matriz 128x128, grosor=5 mm,	
	espaciamiento=1 mm, TE mínimo, TR suficiente para adquirir todos los	~1:05
	cortes, NEX=3, ASSET.	
Secuencia T2 Cubo-sagital,	✓ FOV=26.8 cm, matriz=224x224, grosor=1.2 mm, TR=2500 ms, cortes	
isométrica	suficientes para cubrir toda la cabeza, ajustar ETL para que el TE sea	~1:55
	cercano a 120 ms, ARC.	
Secuencia FLAIR T2 Cubo-	✓ FOV=26.8 cm, matriz=224x224, grosor=1.2 mm, TR=7000 ms, cortes	
sagital	suficientes para cubrir toda la cabeza, ajustar ETL para que el TE sea	~2:55
	cercano a 120 ms, ARC.	
Tensor de Difusión	✓ Utilizar valor-b de 700 s/mm², 35 direccionse de difusión, FOV=25.6 cm,	
(Tractografia)	matriz 128x128, grosor=3 mm, espaciamiento=0 mm, TE minimo, cubrir todo	~7:55
	el encetalo y el cerebelo, IR suficiente para adquirir todos los cortes,	
Pertusion	EPI-GRE, FOV=25.6 cm, matriz 128X128, grosor=5 mm, espaciamiento=1	1.05
	mm, TE=50 ms, TR sufficiente para adquirir todos los cortes, 40 rases por	~1:05
Convencia T1 2D	Conte, Tiempo de duración aprox. 1.06 minutos para 20 contes.	
Secuencia II SD	 Fast SPGR, FOV-23.0 cm, main2-230x230, phase FOV-0.75, TE Mill Full, TI-450 mg, EA-159, grosser-1 mm, sortes suficientes nors subrir toda al. 	
	encéfalo y el cerebelo	~7.55
Espectrosconia 3D	\checkmark Incluive zone del tumor y edema perilesional, si es posible alguna zone de	
Multivoxel	tejido sano FOV=16 cm matriz=8x8 TF=144 ms TR=1000 ms NFX=1.6 ó	~8.25
Malityokor	2	0.20
2 Espectro univoxel para	✓ TE=30 ms. TR=1500 ms. vóxel=2x2x2 cm. 128 scans.	
medición de grasa: 1 en	· _ · · · · · · · · · · · · · · · · · ·	
zona solida del y 2 en tejido		~3:55
sano (hemisferio		
contralateral).		
	Total Minutos	~36:10

* Las secuencias de RM fueron planeadas con apoyo del Especialista en Aplicaciones Avanzadas M. en C. Rafael Favila (GE Healthcare, México).

Preguntas y comentarios:

Dr. Ernesto Roldan-Valadez, Radiólogo, Posgrado en RM. Alumno del Doctorado en Ciencias Medicas UNAM. Departamento de Neuroquímica, INNN.

Tel. 5606-3822, Ext. 2006. Celular. 04455-4390-4729.

Dr. Sergio Moreno Jimenez, Depto. de Radioneurocirugia.

Dr. Camilo Rios Catañeda, Depto. de Neuroquímica.

15) APÉNDICE 3: OTROS CRITERIOS DE ACUERDO AL TIPO DE ESTUDIO

Formato de captura de datos

Biomarcadores avanzados de resonancia magnética como factores pronosticos para la sobrevida de pacientes con glioblastoma multiforme.

Variable	Resultado
Edad	
Sexo	
Volumen tumoral en T1 post gadolinio	
Volumen en T2 (del edema perilesional)	
Coeficiente de Difusión Aparente (ADC)	
Índice de Colina/Naa	
Índice de Naa/Cr	
Índice de Lactato/Cr	
Índice de mioinositol/Naa	
% de grasa en espectroscopia	
Volumen cerebral relativo	
Anisotropia Fraccional	
Grado de resección Qx.	
Respuesta al tratamiento (mediana de sobrevida)	

16) ETAPAS EN QUE PROGRAMA SU INVESTIGACION (NUMERO DE ETAPAS QUE USTED CONSIDERE INTERVIENEN EN SU PROYECTO):

1a.	Diseño de protocolo y presentación a Comité Éti	<u>ca_</u>	_20%
2a.	Inclusion de Pacientes		_40%
За.	Analisis de los Datos		20%
4a.	Publicación de los Resultados		20_%
5a.			%
6а.			%

100 %

FECHA DE INICIO PROGRAMADA:	Agosto MES	<u>2011</u> AÑO
FECHA DE INICIO REAL:	Julio MES	2014 AÑO
FECHA DE TERMINACIÓN PROGRAMADA:	Julio MES	2014 AÑO
FECHA DE TERMINACIÓN REAL:	Julio MES	2016 AÑO



INSTITUTO NACIONAL DE NEUROLOGÍA Y NEUROCIRUGÍA MANUEL VELASCO SUÁREZ

Insurgentes Sur 3877 Col. La Fama, C. P. 14269 México, D.F., Tel. 56-06-14-07 http://www.innn.salud.gob.mx

México, D. F., a 21 de agosto de 2012

DR. DANIEL SAN JUAN ORTA, JEFE DEL DEPTO. DE INVESTIGACIÓN CLÍNICA, P R E S E N T E.

A fin de cumplir con lo que exige la Ley General de Salud de México, y el Reglamento para Investigación Clínica del Instituto Nacional de Neurología y Neurocirugía Manuel Velasco Suárez, envío a usted:

- El protocolo titulado "Biomarcadores avanzados de resonancia magnética como factores pronósticos para la sobrevida de pacientes con glioblastoma multiforme", elaborado de acuerdo al formato oficial, firmado por todos lo autores. Original y dos copias.
- 2. El formato de identificación del ámbito de estudio del protocolo.

Para su evaluación de acuerdo con los Principios Básicos Científicos aceptados en las Declaraciones sobre Investigación Biomédica en seres humanos, de Helsinki II en 1975, Venecia en 1983, Hong Kong en 1989, Sudáfrica 1996, Edimburgo 2000, que señalan textualmente "el protocolo se remitirá a un comité independiente del investigador y de la entidad patrocinadora, para consideración, comentarios y asesoramiento", en lo cual se basará la aprobación, en cuyo caso debe informar trimestralmente del avance del estudio.

ATENTAMENTE

<u>Dr. Ernesto Roldán Valadez</u> Nombre y firma del Investigador Principal

FORMATO DE IDENTIFICACIÓN DEL ÁMBITO DE ESTUDIO DE LOS PROTOCOLOS DE INVESTIGACIÓN CLÍNICA

Título del proyecto:

1) Mencione cual es la vinculación de su proyecto de investigación con las líneas de avance médico actual y futuro: El protocolo propone la utilización de las secuencias de resonancia magnética que marca la literatura medica actual como las ideales para el estudio de los tumores cerebrales, estas secuencias en la actualidad no se utilizan en el INNN.

2) Mencione el vínculo de su proyecto con las líneas de interés nacional:

Cada año se detectan 30 mil nuevos casos en promedio de cáncer cerebral, en México, de estos, los tumores gliales de alto grado (glioblastoma multiforme) representan el tipo más común de tumor cerebral maligno y representan el 40-50% de las neoplasias cerebrales primarias en adultos. A nivel nacional, es uno de los 3 tipos de tumores mas comunes junto con linfomas y leucemias. En las últimas tres décadas, se ha observado un incremento en su mortalidad, más evidente en los pacientes mayores de 60 años, en quienes la incidencia se incrementa 2.5% por año desde 1980.

3) Diga como se vincula su investigación con las líneas de interés para la Secretaría de Salud:

La atención a los tumores cerebrales es una de las líneas prioritarias de la Secretaria de Salud de México, este protocolo se alinea directamente con un interés nacional para atender esta patología.

4) Diga cómo se vincula su proyecto con las líneas de interés para el Instituto:

El Instituto Nacional de Neurología y Neurocirugía (INNN) de la Secretaría de Salud cuenta con el equipo tecnológico más moderno para la atención de tumores cerebrales benignos y malignos, lo que coloca a México como líder en Latinoamérica en la atención de esos problemas. En el año 2010 se atendieron aproximadamente 100 casos de glioblastoma multiforme, un promedio de 2 pacientes por semana.

5) Mencione cuales son las líneas de investigación de interés para usted mismo:

- Analisis avanzado con resonancia magnetica de tumores cerebrales.
- Evaluacion diagnostica del análisis difusión-perfusión en el infarto cerebral.
- ✓ Diagnostico temprano de Alzheimer con análisis cuantitativo de resonancia magnética.
- ✓ <u>Caracterizacion de patología psiqiatrica con espectroscopia por resonancia magnética.</u>

Fecha:

Firma del investigador:

Agosto 21 de 2012

Dr. Ernesto Roldán Valadez

Avances del proyecto de tesis

- Se ha creado una línea de investigación original que previamente no existía en el Instituto Nacional de Neurología y en la Fundación Clínica Médica Sur, esta última lugar de trabajo del alumno.
- En las primeras etapas del proyecto se evaluó la capacidad pronostica para sobrevida de las secuencias de resonancia magnética conocidas como análisis avanzado en 28 pacientes con diagnóstico de glioblastoma multiforme y en 34 controles sanos. Se obtuvieron resultados preliminares de los mejores biomarcadores que dan respuesta a los objetivos planteados en el protocolo de investigación.
- Al termino del 3er año del doctorado, se completaron <u>5 artículos originales</u> relacionados directamente con el protocolo de investigación, 3 de ellos han sido publicados, y los otros 2 se encuentran en proceso de revisión, todos enviados a revistas indexadas en el *Journal Citation Reports*.
- El alumno fue responsable de escribir cada uno de los 5 artículos terminados, incluyendo su análisis estadístico correspondiente; en todo los manuscritos el alumno figura como 1er autor y/o autor correspondiente. Los manuscritos fueron revisados y comentados con el tutor.
- Durante el desarrollo del doctorado, el alumno presento avances semestrales del protocolo de investigación a la Coordinación de Estudios de Posgrado de la Facultad de Medicina, durante la actividad denominada *Coloquios de investigación*. Al final de cada presentación un grupo de expertos designado por el Subcomité de Ciencias Médicas califico los avances semestrales del proyecto y emitió recomendaciones que fueron seguidas por el alumno.
- Hasta el momento de la presentación de esta tesis, se ha obtenido: la aprobación del Examen de Candidatura, el reconocimiento de 3 publicaciones por el Subcomité de Ciencias Médicas y la autorización para presentar el Examen de Grado en la *modalidad de producción científica*. Con la presentación de esta tesis, se finaliza la ultima etapa del proceso de titulación del programa Doctorado en Ciencias Médicas avalado por la División de Estudios de Posgrado de la Facultad de Medicina de la UNAM.



Submission #256 version #1 submitted on January 6, 2014

Title

Choline-to-N-acetylaspartate and lipids-lactate-to-creatine ratios together with age assemble a significant Cox's proportional-hazards regression mode

Authors

- Dr. Ernesto Roldan-Valadez Roldan-Valadez* (Magnetic Resonance Unit, Medica Sur Clinic & Foundation. Mexico City, Mexico.)
- Dr. Camilo Rios (Department of Neurochemistry. National Institute of Neurology and Neurosurgery. Mexico City, Mexico.)
- Dr. Daniel Motola-Kuba (Oncology Unit, Medica Sur Clinic & Foundation. Mexico City, Mexico.)
- Dr. Juan Matus-Santos (Oncology Unit, Medica Sur Clinic & Foundation. Mexico City, Mexico.)

Dr. Antonio Villa-Romero (Coordination of Research, Department of Public Health, National Autonomous University of Mexico. City, Mexico.)

Dr. Sergio Moreno-Jimenez (Radioneurosurgery Unit, National Institute of Neurology and Neurosurgery. Mexico City, Mexico.)

Keywords

Biological markers; Glioblastoma; Magnetic Resonance Imaging; Prognosis; Survival Analysis.

Abstract

Introduction. There has been a longstanding interest in the identification of prognostic biomarkers for high-grade gliomas (HGGs) using magnetic resonance (MR) imaging, but a consensus of which parameters assemble a significant survival model is still missing in the literature; we investigated the significant positive or negative contribution of several MR biomarkers in the prognosis of these tumors. Materials and Methods. A cohort of 28 patients with supratentorial HGGs (11 GBM and 17 anaplastic astrocytomas) was included (19 men and 9 women, mean age of 50.4 years, SD: 16.28; range: 13 - 85 y). Edema and viable tumor measurements were acquired using ROIs in T1-w, T2-w, Flair, apparent diffusion coefficient (ADC) and MR Spectroscopy. Kaplan-Meier and Cox's proportional hazards model were performed. Results. 17 patients died within the follow-up period (3 to 98 months). The median survival time was 1.73 years (range, 0.287-8.947 years). Only 3 out of 20 covariates (Choline-to-N-acetylaspartate and lipids-lactate-to-creatine ratios, and age) contribute significantly to explaining variability in the survival hazards model, score test: $\chi 2$ (3) = 9.098, p = .028. Conclusions. For this data-set MR-spectroscopy metabolites overcome volumetric biomarkers (peritumoral edema and viable tumor) as well as tumor-regions ADC measurements. A consensus about which MR biomarkers should be included in the imaging follow-up is still in need for these tumors.

Files list

File name	File size	File category	File upload date	File comment	View file
00000256_Manuscript-GBM Survival Jan- 05-2014.docx	0.1 MB	manuscript	2014-01-06 02:10:04	Manuscript-blinded	<u>,</u>
00000256_Cover letter-ANE Jan-05- 2014.pdf	0.5 MB	cover letter	2014-01-06 02:15:24	Cover letter	0
00000256_Figure 1 flatten.tif	1.8 MB	figures	2014-01-06 02:16:20	Figure 1. Conventional MRI sequences in a patient with GBM, axial-plane images in: A, T2-w; B, post gadolinium T1-w; C, Flair; D, gradient echo.	0
00000256_Figure 2 BW.tif	0.1 MB	figures	2014-01-06 02:16:35	Figure 2. Survival curve of patients with high-grade gliomas.	0
00000256_Figure 3A-C flatten.tif	1.2 MB	figures	2014-01-06 02:17:06	Figure 3. Schoenfeld residual plots for the selected covariates in the Cox regression model for survival in HGGs. A, Cho/Naa ratio and B, Lipids-lactate/Creatine ratio; in both plots the Lowess curve varies around the zero reference line in a nonsystematic way; after the 3rd-4th year the smooth line appear to demonstrate a trend over time with a negative slope and thus that it had a negative proportional hazard. C, plot of the Age (years); the fitted line appeared to have a positive line, suggesting that the hazard ratio for this covariate increases over time and thus it has a proportional hazard.	

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Running Title: MRI model for survival in GBM patients.

Choline-to-N-acetylaspartate and lipids-lactate-to-creatine ratios together with age, assemble a significant Cox's proportional-hazards regression model for survival prediction in patients with high-grade gliomas.

Abstract

Introduction. There has been a longstanding interest in the identification of prognostic biomarkers for high-grade gliomas (HGGs) using magnetic resonance (MR) imaging, but a consensus of which parameters assemble a significant survival model is still missing in the literature; we investigated the significant positive or negative contribution of several MR biomarkers in the prognosis of these tumors.

Materials and Methods. A cohort of 28 patients with supratentorial HGGs (11 GBM and 17 anaplastic astrocytomas) was included (19 men and 9 women, mean age of 50.4 years, SD: 16.28; range: 13 – 85 y). Edema and viable tumor measurements were acquired using ROIs in T1-w, T2-w, Flair, apparent diffusion coefficient (ADC) and MR Spectroscopy. Kaplan-Meier and Cox's proportional hazards model were performed.

Results. 17 patients died within the follow-up period (3 to 98 months). The median survival time was 1.73 years (range, 0.287-8.947 years). Only 3 out of 20 covariates (Choline-to-N-acetylaspartate and lipids-lactate-to-creatine ratios, and age) contribute significantly to explaining variability in the survival hazards model, score test: χ^2 (3) = 9.098, p = .028.

Conclusions. For this data-set MR-spectroscopy metabolites overcome volumetric biomarkers (peritumoral edema and viable tumor) as well as tumor-regions ADC measurements. A consensus about which MR biomarkers should be included in the imaging follow-up is still in need for these tumors.

Key words: Biological markers; Glioblastoma; Magnetic Resonance Imaging; Prognosis; Survival Analysis.

Introduction

There has been a longstanding interest in the identification of prognostic markers for gliomas;(1) supported by the fact they constitute the most common primary brain tumors with a frequency of approximately 38%.(2) Most of them being glioblastoma multiforme (GBM) grade IV, as classified by the World Health Organization (WHO). It has been considered that conventional imaging reveals only the "tip of the iceberg" of each glioma, with a significant portion of the glioma cells already invading peripheral tissue.(3) The biological behavior of gliomas produces microscopic invasion to surrounding tissues, especially white matter (WM) tracts,(4) from the obvious area of disease;(5) as results there is normal appearance white matter (NAWM), where infiltration is not able to be detected using conventional imaging protocols of MR.(6) With this regard, areas of extensive edema surrounding the tumor on T₂-weighted imaging (T₂-w) are often assumed to contain tumor cells that have infiltrated into the brain tissue.(7) These factors may affect the survival rate of patients, because even when all radiologically visible portions of a tumor have been excised, the surgical margins may not be "clean," and further neoplastic growth can occur in the adjacent brain, leading from microscopic residual to gross recurrence.(8)

GBM shows a mean patient survival of still only 14 months, despite recent advances in surgery and radiochemotherapy.(9) The mean life expectancy for patients with anaplastic astrocytoma (WHO grade III) is slightly longer, at 41 months.(10) A complete surgical excision of high-grade gliomas (HGGs) (WHO grades III and IV) without tumor recurrence is impossible, due to their biological behavior. Thus, the interdisciplinary therapeutic concept today combines microsurgery followed by fractionated external beam radiation and chemotherapy.(11)

Several studies for predicting the outcome in HGGs patients have been described considering clinical factors as Karnofsky performance scale, tumor localization, tumor size, histopathologic subgroups, extent of resection, radiotherapy and reoperation, molecular and genetic factors; however, many of then have shown conflicting results.(1) Even given the same pathological diagnosis and equivalent treatments, some patients have a relatively favorable outcome, while others have a very poor prognosis.(12) Conventional MR examination for diagnosis of gliomas around the world is usually limited to identify edema boundaries and regional tumor infiltration to those patterns observed in the T_2 -w pre-contrast and T_1 -w post-contrast images;(13, 14) recent studies have proved predictive value of the apparent diffusion coefficient (ADC),(15) the measurement of choline-to-N-acetylaspartate ratio (Cho/NAA) using MR spectroscopy (MRS),(16) and the presence of immediate (<1 cm) and distant edema (>1 cm),(14) in the survival analysis of patients with GBM.

Despite the growing evidence of studies exploring the prognostic ability of magnetic resonance (MR) biomarkers inpatients with HGGs, a consensus of the best survival biomarkers is still missing in the literature; we considered necessary to investigate additional evidence of the quantitative MR biomarkers contribution's towards a positive or negative effect on survival in these

group of patients. In this study, we specifically investigated the role of basic demographics of a small group of patients together with conventional MR volumetric measurements, diffusion weighted imaging and MRS as biomarkers able to predict overall survival in patients with HGGs.

Material and Methods

Patients

A cohort of 28 patients with supratentorial HGGs underwent follow-up from December 1999 to January 2012. The group consisted of 19 men and 9 women, with a mean age of 50.4 years (SD: 16.28; range: 13 – 85 y). The patients' postoperative survival times (in years) were retrospectively obtained according to the Oncology Unit tumor register database. Pathologic findings included 11 GBM and 17 anaplastic gliomas. Exclusion criteria were any surgery, radiation therapy, or chemotherapy of a brain tumor before inclusion in the study as well as lack of histopathologic diagnoses, missing imaging data or presence of artifacts. Treatment was not used to exclude patients. Most of the patients were treated under the same protocol; after resection, they received chemotherapy including Temozolomide (Temodal, Schering-Plough, NJ, USA), and radiation therapy (60 Gy), the Stupp protocol. All patients underwent biopsy or surgical resection of the tumor with histopathologic diagnosis based on WHO criteria. A radiologist interpreted the MR images blinded to the patient's history. MRI examinations with other structural abnormalities were excluded. The local Institutional Review Board approved the study (reference number 2011.044).

Brain image and data acquisition

MR was performed by using a 3T unit (Signa HDxt, GE Healthcare, Waukesha, WI, USA) with a high-resolution eight-channel head coil (Invivo, Gainesville, FL, USA). MR sequences included conventional axial T_2 -w, axial Fluid-Attenuated Inversion Recovery (Flair), axial Spoiled Gradient Echo (SPGR), diffusion-weighted imaging (DWI) and axial T_1 -w, using 0.1 mmol/kg of body weight of gadopentetate dimeglumine (Magnevist; Schering, Berlin, Germany). DWI was performed using a singleshot SE EPI sequence with b-values of 1000 s/mm² and an image without diffusion weighting with b-value of 0 s/mm². Figure 1 shows the appearance of conventional MRI sequences in a patient with GBM.

Multi-voxel spectroscopic imaging (MV-MRS) was performed using a point-resolved spectroscopic sequence technique (PRESS), the volume of interest (VOI) size was specifically adjusted positioning the voxel over the lesion and trying to minimize partial-volume effects resulting from other neighboring tissues including bones and cerebral spinal fluid (CSF) of the ventricles. Proton spectra were recorded in the axial plane with T1-w post-constrat images via TR; 1500 ms, TE; 26 and 144 ms, FOV; 24×24 cm, 1-1.5 cm section thickness, 256×256 matrix and 24×24 phase encoding.

Image postprocessing and data analysis

One radiologist, blind to the clinical history of each patient, manually traced the boundaries of the tumor regions; volumetric calculations were included in the analysis: the T_1 -w post-contrast volume, the T_2 -w pre-contrast volume, the T_1 -w/ T_2 -w ratio, the Flair pre-contrast volume, and the Flair/ T_1 -w ratio.

For the DWI sequence, the ADC was measured at five tumor regions: viable tumor region (area of enhanced rim at T1-w postcontrast); necrosis (considered the cystic cavity); an immediate zone of edema (arbitrarily chosen as a 10-mm-wide band); a distant zone of edema (area adjacent to the immediate zone of edema, chosen also as a 10-mm-wide band) and an area of normal-appearing white matter (NAWM), drawn in the patient's contralateral hemisphere.

For the MRS data, metabolite signal peaks were analyzed and centered within a range of 0-4.35 ppm as follows: methyl protons of N-acetylaspartate (NAA) at 2.0 ppm, N-trimethyl protons of choline-containing metabolites at 3.2 ppm (Cho), creatine (Cr) at 3-3.1 ppm, a compound peak containing lipids and lactate (LL) at 0.8-1.4 ppm, and a compound peak of the protons of myo-inositol (mI) at 3.56 and 4.06 ppm.(17) Automatic shimming of the linear x, y, z channels was used to optimize field homogeneity, water resonance and water suppression pulses were optimized. All volumetric and MRS imaging data were analyzed in a GE Advantage Workstation with the software FuncTool 9.4.04b (GE Medical Systems, Milwaukee, WI, USA).

Definition of Survival Time

For the purpose of this study, the preoperative images acquired at the time of the first operation were considered as the baseline and were used to monitor the evolution of the disease. Overall survival was evaluated from the baseline to death or, for cases that were not followed until death (eg, living patients), from the baseline to the time of last available follow-up.

Statistical analysis

According to the aims of the study, only 20 variables related with basic demographics and MRI biomarkers were included in the analyses: three related to clinical findings (gender, age, tumor grade); two obtained from the visual inspection of conventional MR imaging (affected side and tumor location), 3 variables from the post-processing of volumetric measurements (T_1 -w post-contrast volume, T_2 -w pre-contrast volume, T_1 -w/ T_2 -w ratio, Flair pre-contrast volume, Flair/ T_1 -w ratio); 5 variables from ADC measurements at five tumor regions (viable tumor region, necrosis, immediate zone of edema, a distant zone of edema, normal-appearing white matter), and 5 variables from MRS metabolites quantification (NAA/Cr, Cho/Cr, LL/Cr, Cho/Naa, and mI/Cr). Univariate analysis using the Kaplan-Meier method was used for the calculation of median survival time.(18) Multivariate analysis was performed using a Cox's proportional hazards model.(19) Hazard ratios (HRs) and their adjusted 95% confidence intervals (CIs) were calculated. For all analysis, statistical significance was indicated by a *p*-value < 0.05.

Software. All analyses were carried out using the IBM® SPSS® Statistics software (version 22.0.0.0 IBM Corporation; Armonk, NY, USA). *Presentation of data*. General presentation of the manuscript followed the guidelines set by the International Committee of Medical Journal Editors.(20)

Results

General data

The postoperative follow-up period range was from 3 to 98 months, 17 patients died within the follow-up period. The median survival time was 1.73 years (range, 0.287-8.947 years). The descriptive statistics of the selected variables are summarized in Table I, the survival curve for this cohort is depicted in figure 2.

Survival analysis

We performed a Cox proportional hazards regression analysis to fit a null model containing only an intercept parameter (residual χ^2 (13) = 22.647, p = .046); which allowed us to identify the significant independent prognostic factors of survival time; only three out of 20 available covariates were selected: Cho/Naa ratio (p = .017), age (p = .072) and lipids-lactate/creatine (p = .109).

A second Cox regression analysis using only the three selected covariates and the complete method (Method = Enter) of regression for age (years) and Cho/Naa and lipids-lactate/creatine ratios; provided an "Omnibus Test of Model Coefficients" that simultaneously assessed the effects of the parameters in the model. We found that our three covariates contribute significantly to explaining variability in the survival hazards, score test: χ^2 (3) = 9.098, p = .028. For this new model, the -2 x Log Likelihood relative to the null model also indicated a significant improvement in the model fit after adding the covariates, LR test: χ^2 (3) = 14.591, p = .002. The Wald's tests calculated the effects of the variables in the model after adjusting for the effects of the remaining ones. At the 5% test level, the three covariates significantly affected the hazard function, table 2.

Cho/Naa ratio was the strongest survival predictor with a log-hazard function of 2.672. This value indicated that, for every additional unit increase in the Cho/Naa ratio, patients increased 2.672 times the hazard function to report a death stage than those who did not have an increase, controlling for all other factors in the model.

The LL/Cr ratio depicted a log-hazard function of 0.584; considering the negative sign of its beta value, this hazard value can be interpreted as a 41.6% reduction in the risk to report a death stage. The age has a positive value with a log-hazard function of 1.062, indicanting a 6.2% increase in the risk to report death for every additional year in the patient's age.

We completed our Cox's analysis by plotting the *Schoenfeld residuals* to check for proportional hazards;(21) they contain values for subjects with uncensored survival times. For the first two covariates (Cho/Naa and LL/Cr), the Lowess curve varies

around the zero reference line in a nonsystematic way, figures 3 A and B; although after the 3rd-4th year the smooth line appear to demonstrate a trend over time with a negative slope and thus it had a negative proportional hazard; for both variables, the plots are only based in few events ant it is seen; it is largely influenced by the residuals at the first time points. For the covariate age, the fitted line appeared to have a positive line, suggesting that the hazard ratio for this covariate increases over time and thus it has a proportional hazard; figure 3 C.

Discussion

Survival time for GBM is still approximately 1 year despite all the treatment modalities including surgical resection, radiotherapy and chemotherapy.(1) The median survival time of 1.73 years in this study is explained by the combined analysis of grade III and IV tumors. However, until recently, a 3 year life period for glioblastoma was indicated in 3% of patients and a 5 years life period was indicated in 0%; nowadays, there are documentations of patients surviving 10 years with GBM.(22, 23) To date, the usually selected variables for predicting prognostic subgroups of patients with HGGs patients include age, Karnofsky performance score, tumor localization, tumor size, histopathologic subgroups, extend of resection, adjuvant therapy and reoperation; recent studies tend to evaluate molecular genetics or molecular biology excluding in most cases an integrative assessment of quantitative MR biomarkers in the prognosis such as those from spectroscopy.(24)

Among all the factors studied in our Cox analysis, the Cho/Naa was confirmed as the most significant factor influencing the survival time (P < 0.028) through the hazard function; a recent work demonstrated that the signal intensity ratio of Cho to the metabolic marker for neuronal integrity N-acetyl-aspartate (NAA) changed upon antiangiogenic treatment and that these changes may predict the outcome of therapy;(25) our data support a similar study using the same index almost a decade ago but that manuscript did no report on the influence of lipids-lactate-to-creatine ratio;(16) a previous study has reported that MRS findings may precede the perfusion abnormalities and contrast enhancement features in HHGs.(26) It is interesting for our group that, despite the evidence supporting the use of these MR biomarkers there is still not international consensus whether the quantification of MRS metabolites should be mandatory in the imaging evaluation of brain tumors performed at most tertiary-care hospitals; the reality is that spectroscopy, perfusion and diffusion techniques are not used to anticipate tumor progression in a day-to-day basis.(11) then classic T1 and T2 imaging features remain as the criterion standard.(27)

Recent studies has proved a significant association of ADC in subgroups stratification,(15) and in the methylation status of the MGMT promoter in patients with GBM,(24) however these studies missed to complement their analyses including a Cox's regression model; our data showed that the simultaneous assessment of ADC values from several tumor regions were not considered as significant biomarkers using a Cox's regression method.
The association between low levels of lactate and hypoxia in tumor tissue was reported four decades ago,(28) this evidence is relevant because under hypoxic conditions an hypoxia inducible factor, up-regulates genes encoding vascular endothelial growth factor that participates in infiltration of tumor cells and angiogenesis.(29) In patients with HGGs, it has been observed a higher lipid and lactate levels measured by MRS associated with significantly worse overall survival.(30) Then, our finding that low levels of lipids-lactate/Cr ratio represents up to a 41.6% reduction in the risk of death is a reverse statement that increased levels of tumor hypoxia have been associated with worse clinical outcomes.(31)

The age at the time of diagnosis has been considered one of the most important prognostic factors.(32) Survival time is significantly longer in younger patients,(33) with a clear correlation between increasing age and decreasing overall survival; our findings supported this knowledge with a 6% increase in the risk of death per year of age.

Although the gender was not significant, we found a male dominance (67.8%), which is in agreement with most of the studies about incidence for glial malign tumors in the literature.(34)

The assessment of quantitative MR imaging biomarkers is becoming standard in tumor patients, but a real understanding of the significant effects of the several MR modalities is still in its early days. Several reports have reported that localization of tumor may effect patient's survival time, (34, 35) other multivariate analyses found that the location of the tumor may have better prognosis but its effect size is not enough to define this feature as an independent factor; (1, 36) in our analysis there was no significant effect between the location of the tumor and the survival.

We evaluated the effect of peritumoral edema through measurements of 3D volumes in T_2 -w and Flair sequences; although an association has been observed with a significantly shorter survival time in many studies,(36) we did not find significant influence for our data set.

Some limitations in this study need to be addressed: the data were obtained and analyzed retrospectively, so a control group was not available, although the study answers the questions posed at the introduction (which MRI biomarkers has a significant association with the survival in patients with HGGs), several clinical and surgical variables were intentionally out of the scope of this study (extent of resection, radiation dose and adjuvant chemotherapy), as we want to report the significance of MR biomarkers without the influence of other clinical/surgical/radiotherapy variables. The metabolism of phospholipid cell membrane turnover is associated to several metabolites such as glycerophosphocholine, phosphocholine, and phosphoethanolamine each one playing major roles as indicators for tumor growth; its discrimination however, can not be performed using the conventional (1) H MRS, it requires specialized hardware with (31) P MRS.(37) Then, the research on MR biomarkers still needs further investigation with larger patient population to determine the robustness of the selected variables in predicting survival and to increase understanding of the morphologic and functional characteristics of brain tumors, also to

promote generalizability of the results.

In conclusion for these data, a higher preoperative Cho/Naa ratio, a lower Lipids-lactate/Creatine ratio and the years of age, are significant predictors of survival that overcome the classic T_2 -w and T_1 -w post gadolinium volumetric standards (representatives of peritumoral edema and viable tumor respectively) and the ADC measurements (diffusibility of water). Considering MRS is able to detect the conversion of normal tissue to tumor infiltration before conventional MR sequences show conspicuous findings, and its acquisition do not require contrast; MRI units around the world might adapt the post-processing methods of MRS to their own contexts in order to validate data about implementation and outcomes that can be generalisable beyond one setting. Due to the large number of variables (qualitative and quantitative) to which a radiologist is exposed during an advanced imaging evaluation of brain tumors, the use of multivariate assessment combined not only with Kaplan-Meir but also with Cox's regression methods is a strong consideration in the diagnosis and follow-up of these patients.

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Tables

Table 1. Descriptive statistics of the selected variables.

Categorical variables	Number	Percent
Gender		
Male	19	67.9%
Female	9	32.1%
Tumor Grade		
III	11	39.3%
IV	17	60.7%
Cerebral Hemisphere		
Left	13	52%
Right	8	32%
Both	4	16%
Tumor location		
Zone I	10	35.7%
Zone II	3	10.7%
Zone III	12	42.9%
Zone IV	3	10.7%
Continuous variables	Mean	Std. Deviation
Age (years)	50.54	16.283
ADC Tumor (mm ² /s)	.001310	.000391
ADC Necrosis (mm ² /s)	.002290	.000927
ADC Proximal Edema (mm ² /s)	.001294	.000460
ADC Distal Edema (mm ² /s)	.001323	.000283
ADC Normal Tissue (mm ² /s)	.000885	.000200
Naa/Cr	.874	.637
Cho/Cr	2.353	1.190
Lipids-Lactate/Cr	3.806	3.204

1.572	1.272
3.613	2.033
24.047	29.551
74.948	46.699
.337	.316
79.031	44.647
.298	.270
	1.572 3.613 24.047 74.948 .337 79.031 .298

Table 2. Wald's tests assessing the covariates effects on the hazard function.

	В	SE	χ^2	df	p-value	Log-hazard	95.0% CI		
			(Wald test) functions					Upper	
Cho/Naa ratio	.983	.448	4.821	1	.028	2.672	1.111	6.423	
LL/Cr ratio	538	.268	4.039	1	.044	.584	.345	.987	
Age (years)	.060	.034	3.059	1	.080	1.062	.993	1.136	

Figure legends



Figure 1. Conventional MRI sequences in a patient with GBM, axial-plane images in: A, T_2 -w; B, post gadolinium T_1 -w; C, Flair; D, gradient echo.



Figure 2. Survival curve of patients with high-grade gliomas.



Figure 3. Schoenfeld residual plots for the selected covariates in the Cox regression model for survival in HGGs. A, Cho/Naa ratio and B, Lipids-lactate/Creatine ratio; in both plots the Lowess curve varies around the zero reference line in a nonsystematic way; after the 3rd-4th year the smooth line appear to demonstrate a trend over time with a negative slope and thus that it had a negative proportional hazard. C, plot of the Age (years); the fitted line appeared to have a positive line, suggesting that the hazard ratio for this covariate increases over time and thus it has a proportional hazard.

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Title: Correlates and interactions among global DTI-derived metrics, gender and age: a pilot study in healthy and glioblastoma multiforme brains.

Article Type: Original Contribution

Keywords: Brain neoplasms; Diffusion Tensor Imaging; Magnetic Resonance Imaging; Statistics as Topic; Software Tools.

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Abstract

Purpose. Although some relationships among DTI-derived tensor metrics in healthy brains and brains with GBM (glioblastoma multiforme) have been reported in the last decade, they are still not completely understood.

Subjects and methods. Case-control study performed on brain MRI data by using DTI sequence with 25 directions. Image postprocessing using the FSL software allowed the calculation of eleven tensor metrics: mean diffusivity, fractional anisotropy, pure isotropic diffusion, pure anisotropic diffusion, the total magnitude of the diffusion tensor, linear tensor, planar tensor, spherical tensor, relative anisotropy axial and radial diffusivity. Partial correlation analyses and multivariate Mancova controlling the effect of age and gender were performed.

Results. 27 patients with GBM (mean, SD; 48.41 ± 15.18 years; range, 18-78 years) and 34 controls (Mean, 41.47; SD, \pm 21.94 years; range, 21-80 years) were included. From 56 comparisons among bivariate correlations, only seven were significantly different between groups. There was a main effect of diagnosis: F-value (11, 23) = 11.842, p = < .001, with a large effect size (partial eta squared = .850); a similar result was obtained for age.

Conclusion. DTI-derived metrics depict significant differences between healthy brains and brains with GBM, with specific magnitudes and correlations. This study makes a contribution to decrease the underlying empiricism about these novel biomarkers of brain impairment.

Key words: Brain neoplasms; Diffusion Tensor Imaging; Magnetic Resonance Imaging; Statistics as Topic; Software Tools.

Introduction

Quantitative magnetic resonance (MR) techniques assessing the impairment of intraaxial brain tumors have been accepted in the diagnosis and follow-up, as it may complement the management of the surgical neuro-oncology team aimed to conserve vital cerebral tissue while maximizing tumor resection.(1, 2) A caveat should be raised in the pathologic and regional MR imaging evaluations of astrocytomas grades II to IV, as evidence suggest these tumors would benefit from the use of a *global* measurement of brain impairment: they frequently contains multiple areas of variable histologic features, conditioning that sampling error in a biopsy may mean that the degree of malignancy seen by the neuropathologist may not reflect the degree of malignancy present elsewhere in the tumor, which may result in significant undergrading of some lesions.(3)

The rising interest in the imaging evaluation of brain tumors using of diffusion tensor imaging (DTI)derived metrics measurements has showed their clinical applicability in a tumor-region assessment (cystic cavity, enhancing rim, edema, normal-appearing white matter).(4-6) Almost a dozen of combinations of the terms of the diagonalized diffusion tensor, that is, the eigenvalues λ_1 , λ_2 , and λ_3 , have been reported with several scalar measures of diffusion, such as mean diffusivity (MD), fractional anisotropy (FA), pure isotropic diffusion (p), pure anisotropic diffusion (q), the total magnitude of the diffusion tensor (L), linear tensor (Cl), planar tensor (Cp), spherical tensor (Cs), relative anisotropy (RA); and axial (AD) and radial diffusivity (RD).(4, 5, 7-9) However, to best of our knowledge, there is currently neither a clear understanding of the expected measurements among these variables in each tissue, nor consensus about which tensor metrics might be useful in the evaluation of brain tumors.

In this study we aimed to (1) determine the normal limits (magnitudes) of previously reported DTIderived tensor metrics in healthy brains and brains of patients with glioblastoma multiforme (GBM) by using a *global* approach, that is, a single measure of the whole brain for each metric; (2) evince the statistical significance of the mean differences between these groups of subjects; and (3) to analize the DTI-metrics correlates considering the influence of clinical diagnosis.

Materials and methods

Subjects

Case-control study design; inclusion criteria considered preoperative brain MR examinations between January 2010 and September 2012 of patients with at first (suspected) diagnosis and later pathology confirmation of astrocytoma grade IV, GBM according to the WHO. Exclusion criteria applied to corticosteroid or antibiotic treatment, lesions with areas related to calcification and/or haemorrhage and previous brain surgery. A control group included young and elderly healthy volunteers recruited among the enrolled interns and medical residents of the hospital as well as elderly subjects from our Geriatric unit. All volunteers received detailed health examinations; exclusion criteria considered major neurological, psychiatric, or cardiovascular diseases. A radiologist interpreted the MR images blinded to the patient's history. MR examinations with other structural abnormalities were excluded. The local Institutional Review Boards approved the study (Project #2011.044).

Brain image acquisition

MR sequences included conventional axial T2-weighted imaging, axial Fluid-Attenuated Inversion Recovery (FLAIR), axial Spoiled Gradient Echo (SPGR), DWI and axial T1-weighted imaging, using 0.1 mmol/kg of body weight of gadopentetate dimeglumine (Magnevist; Schering, Berlin, Germany); healthy volunteers did not received endogenous contrast. DTI was performed using a single-shot SE EPI sequence. Diffusion gradients were applied in 25 directions with b-values of 1000 s/mm^{2 and} an image without diffusion weighting with b-value of 0 s/mm². DTI sequences were acquired in the axial plane with 44 contiguous sections, 2.4 mm section thickness, no intersection gap, TR/TE of 17,000/80 ms, with parallel imaging to reduce off-resonance artifacts (PI factor was 2); 25 x 25 cm FOV, and 128 x 128

matrix size. MR was performed on a single occasion by using a 3T unit (Signa HDxt, GE Healthcare, Waukesha, WI, USA); and a high-resolution eight-channel head coil (Invivo, Gainesville, FL, USA).

Image postprocessing and data analysis

We use the software dcm2nii (10) and the FMRIB Software Library (FSL) v. 4.1.9;(11) extraction of DTI images used the *Brain Extraction Tool (BET)* v. 2.1.(12) Eddy currents were corrected using the *FMRIB's Diffusion Toolbox v. 2.0*; the *Reconstruct Diffusion Tensor (DTIFIT)* and the *fslmaths tool* generated the eigenvector and eigenvalue maps for each tensor metric. The *fslstats tool* calculated the scalar measures (mean values) of each whole-brain biomarker. By using the major (λ_1), intermediate (λ_2), and minor (λ_3) eigenvalues it is possible to calculate a single global measure for each selected tensor metrics;(13), table 1 shows the tensor-metric formulas.

Statistical analysis

Sample size. The adequacy of the sample size to expect validity from our results was based on two considerations: First, we defined the phase of our research, as a Phase I, "exploratory phase", according to Zhou et al;(14) in which new method is first evaluated in a clinical setting to determine whether the test discriminates between diseased from nondiseased patients. Second, by matching this phase with the summarized list of computed sample sizes needed for an exploratory retrospective study reported by Obuchowski et al.,(15) at least 10 diseased patients and 10 control patients are required to keep statistical validation considering the type I error rate is set at 0.05 and type II error rate is ≤ 0.10 , and power ≥ 0.90 . Additionally, our sample size allowed us to run a Manova analysis to investigate whether any mean differences between the healthy brains and brains of patients with GBM on the combination of dependent variables were likely to have occurred by chance. According to Pallant,(16) the absolute minimum of cases to have in each cell must equal at least the number of dependent variables, in our study, we had twenty-two cells (two levels of our independent variable: healthy brains/brains with GBM, and eleven dependent variables for each). The study was run in 34 controls and 27 patients; this size also follow the

recommendation of Tabachnik and Fidell,(17) for whom a sample size of at least 20 in each cell should ensure "robustness" in the results. Assumptions testing assessed the normality of the distribution of the DTI-derived scores by using the Kolmogorov–Smirnov's and Shapiro-Wilk normality tests;(18) outliers; multivariate normality; linearity; multicollinearity and singularity and the homogeneity of variancecovariance matrices.(16)

Partial correlation analysis, the Pearson's correlation coefficient described the degree of linear relationship between each pair of variables. Considering the age range and the gender of the subjects in our study, we wanted to determine the correlations among tensor metrics *without the effect of age and gender*, so, a *partial correlation analysis* controlling these variables was carried out for each group (healthy brains and brains with GBM). The strength of the linear relationship corresponding to each correlation coefficient value was interpreted as *Very strong* (at least of 0.8), *Moderately strong* (0.6 up to 0.8), *Fair* (0.3 up to 0.6) and *Poor* (less than 0.3). Squaring r-values represented the *coefficient of determination*, the proportion of variance that each two compared variables had in common.(19) The statistical significance of the difference between r coefficients from both groups was tested converting each pair of r values into a standard z scores, then using the formula proposed by Pallant and colleagues (20):

$$Z_{obs} = \frac{Z_1 - Z_2}{\sqrt{\frac{1}{N_1 - 3} + \frac{1}{N_2 - 3}}}$$

Observed Z value \leq -1.96 or \geq 1.96 were considered statistically signicantly different.

A two-way between-groups *multivariate analysis of covariance* (Mancova) was performed to investigate diagnosis and gender differences in tensor metrics measurements.(21) The eleven tensor-metrics represented the dependent variables used; independent variables were the diagnosis and gender; age was used as a covariate to control for individual differences. The *effect size* was obtained using the Eta squared

value.(22) It considered .01 as a small effect, .06 as a medium effect and .14 as a large effect.(23) Statistical significance was indicated by a *p*-value < 0.05.

Software. All analyses were carried out using the IBM® SPSS® Statistics software (version 21.0.0.0 IBM Corporation; Armonk, NY). General presentation of the manuscript followed the guidelines set by the International Committee of Medical Journal Editors.(24)

Results

Demographic data and quantitative DTI tensor maps

The study was conducted in 61 subjects; 27 patients: 13 females (mean age 50.0 ± 15.4 years, range 31-73 years) and 14 males (mean age 46.93 ± 15.4 years, range 18-78 years); and 34 controls: 26 females (mean age 41.04 ± 22.3 years, range 21-80 years) and 8 males (mean age 42.88 ± 21.89 years, range 24-72 years). All tensor maps could be generated using the FSL software, which added up 671 tensor-metrics measurements included in the analysis. Figure 1 shows an example of some of the MR sequences and tensor-metric maps used in the data analyses.

Normality tests, magnitudes of means and SD

There was a normal distribution for all metrics (requisite for the Pearson's correlation analyses). We counted the number of decimal places to the right of the decimal point (to understand the corresponding magnitudes of each DTI biomarker); five tensor metrics reported mean values within the tenths place: Cs, FA, RA, Cp and Cl; none tensor measurement fell in the hundredths place; five tensors values fell in the thousandths place: L, p, AD, MD and RD; and one tensor metrics had values in the ten thousandths place: q. Table 2 shows the means and SD ordered by descending means.

Partial correlation analyses

A scatterplot for each group showed no serious violation of the assumptions of linearity, homoscedasticity, and outliers (Fig 2). In healthy brains, significant very strong bivariate correlations

were observed for: Cs \Leftrightarrow RA (-), Cs \Leftrightarrow Cp (-), Cs \Leftrightarrow L (-), RA \Leftrightarrow Cp (+), RA \Leftrightarrow Cl (+), Cp \Leftrightarrow Cl (+), L \Leftrightarrow p (+), L \Leftrightarrow AD (+), L \Leftrightarrow MD (+), L \Leftrightarrow RD (+), p \Leftrightarrow MD (+), p \Leftrightarrow RD (+) and MD \Leftrightarrow RD (+); and moderately strong significant correlations were calculated for: FA \Leftrightarrow q (+), p \Leftrightarrow AD (+), AD \Leftrightarrow MD (+) and AD \Leftrightarrow RD (+). In brains with GBM, the corresponding significant very strong bivariate correlations included: Cs \Leftrightarrow RA (-), Cs \Leftrightarrow Cp (-), Cs \Leftrightarrow L (-), FA \Leftrightarrow Cl (+), FA \Leftrightarrow q (+), RA \Leftrightarrow Cp (+), RA \Leftrightarrow Cl (+), Cp \Leftrightarrow Cl (+), L \Leftrightarrow p (+), L \Leftrightarrow AD (+), L \Leftrightarrow MD (+), p \Leftrightarrow AD (+), p \Leftrightarrow MD (+), p \Leftrightarrow RD (+), AD \Leftrightarrow MD (+), and MD \Leftrightarrow RD (+); the moderately strong significant correlations were observed in: Cl \Leftrightarrow q (+), AD \Leftrightarrow RD (+) and AD \Leftrightarrow q (+). Tables 2 and 3 present the means, SD, and correlations among the global tensor-metrics (controlled for the effect of age and gender).

From the 55 pairs of bivariate correlations in each a group, statistical significances of the difference between r coefficients were observed in only seven pairs of variables: Cs \Leftrightarrow Cp, FA \Leftrightarrow Cl, FA \Leftrightarrow q, RA \Leftrightarrow q, Cl \Leftrightarrow q, L \Leftrightarrow p and L \Leftrightarrow MD.

Mancova analysis

After adjusting for age, there was not interaction effect between the gender and clinical diagnosis F (11, 23) = 1.115, p = .394. There was not main affect of gender F (11, 23) = 2.060, p = .069; however, the main effect of diagnosis was statistically significant F (11, 23) = 11.842, p < .001, with a large effect size (partial eta squared = .850). The age also had a significant influence as a covariate F (11, 23) = 10.523, p < .001, with a large effect size (partial eta squared = .834). Table 4 and figure 3 depict the estimated marginal means; the age was controlled at the value of 43.92 years.

Discussion

DTI uses three eigenvectors and three eigenvalues to represent the main diffusion directions and magnitudes, respectively, of the molecular motion of water in a three dimensional space;(25) with this information, DTI produces neural tract images instead of using this data solely for the purpose of

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assigning contrast or colors to pixels in a cross sectional image. The principal eigenvector is assumed to correspond to the main axis of diffusion and, therefore, the axis of the white matter tract.(26) Although early publications on this topic, used to measure only two DTI metrics, FA and MD as accepted biomarkers of brain impairment, recent protocols have included almost a dozen of DTI- metrics in their analyses;(6) the caveat of using several biomarkers to characterize brain pathology in clinical applications is that, it is still not completely understood a priori which are the magnitudes and associations among these measurements to quantify pathological changes in brain tissue.(7) MD is understood as a synonym of the coefficient of diffusion in different space guidelines.(27) FA measures the directionality of molecular motion in the brain,(28) it is a quantitative index for diffusion anisotropy that correlates with microstructural integrity of myelinated fiber tracts.(29)

We believe the clinical relevance of this study has several components: first, the importance to understand these metrics is supported by several studies in the MR-DTI literature: Cs, Cp, Cl, FA and MD, have been related with brain abscesses, GBM and brain metastasis;(4) p, q and L measurements have been previously applied to the evaluation of GBM and brain metastasis;(5, 30) AD has been reported in enchepalomylielitis of the spinal cord,(31) AD and RD have been correlated with brain development,(32) infantile spasm,(33) amyotrophic lateral sclerosis,(34) schizophrenia,(35) and brain tumors;(36) with only one recent study integrating a tumor-region diagnostic evaluation of 11 DTI-metrics.(6) Interestingly, the way researchers report their findings in most of these studies, is by presenting the correlates among these metrics, that is, the increase/decrease in values of one variable, compared with those of others; more complex multivariate evaluation like Mancova analysis is missing in medical literature.

Second, our study intended to make a small contribution toward answering a clinically-relevant question in MR DTI: which are the relationships among the most recently published tensor metrics in the medical literature that help to characterize damage to brain tissue? We share our findings about the observed global magnitudes of these metrics in each group of subjects (useful data for detecting errors in software calculations during the image postprocessing), the significant associations and direction (direct or inverse) for each bivariate correlation, and present a reference of estimated means after a multivariate analysis. We believe our data represent useful information for radiologist and/or bio imaging experts to explain the relationships between tensor metrics to clinicians (neurologists, neurosurgeons, psychiatrist, pediatricians, etc.) as well as in the preparation of prospective studies with clinical application.

Third, our study tested that a global approach is able to detect significant differences between healthy and diseased brains; we also got evidence that the gender neither had interaction nor main effect in the observed measurements of DTI metrics, however the age did, so multivariate analysis controlling this statistic effects should be included in future studies. We evaluated 56 different bivariate correlations for each studied group, with only 7 of them showing a significant difference between healthy brains and brains with GBM, none of these results have been previously reported. It will be interesting to know the data analyses from other research groups that help validate and support the clinical significance of the results presented in this study.

Fourth, despite that most of our results could not be compared with previous studies, some interesting associations are worthy to be noted: for example, the precise value of MD and FA in a clinical setting is still controversial;(5) Traditionally, MD and FA have showed negative correlation with increased MD and decreased FA in the peritumoral signal-intensity abnormality when compared with normal-appearing white matter;(37) we observed a poor-negative non-significant correlation when controlling the effect of age and gender. One explanation might be that, in contrast to MD that measures the magnitude of molecular motion of water and does not depend directly on the integrity of myelinated fiber tracts, FA depends on the restriction of water proton movement along myelinated fiber tracts, it is the weighted average of the anisotropic diffusion.(8) The absence of partial correlation analyses in previous reports could explain some conflicts in the values of MD and FA in characterizing tumor type.(5, 38)

For MD, the highest-significant correlation (r = 1) was observed with p and RD, a positive correlation between RD and MD have been previously observed in human brains.(39) We found an inverse relationship between MD with Cp, which seems opposite to previously suggested direction measuring MD, FA and the shape tensor metrics (Cp, Cs, Cl) between tumoral brain tissue, metastasis and abscesses.(4) A high correlation (more than 0.8) observed between some pairs of variables, is a warning for those investigators willing to conduct multivariate assessments (Anova, Mancova, Regression) as the presence of multicollinearity suggest the removal of some variables before the analysis.(40)

For most of the tensor metrics correlates, we observed a decrease in the strength of the linear relationship after controlling the effect of age and gender (unpublished data). A poor correlation between a pair of metrics should be interpreted cautiously; in clinical settings these measurements might have the physiological implication to represent independent biological biomarkers.

FA's mechanisms of decreasing its value in brain tumors are still unclear: they could be related to an increase in extracellular space secondary to neuronal and fiber tract destruction;(38) or a decrease in extracellular space secondary to tumor infiltration (observing a negative relationship between FA and tumor cellularity).(41) The FA-value can change by virtue of its definition as a ratio between q and L;(7) the inclusion of the parameter q may be able to provide a more complete picture of the diffusion profile of a brain tumor, in our study this parameter showed the highest-significant strength of correlation with FA (r = .928) in brains with GBM, it would seem an expected finding from its formula (ratio of q/L); however, our findings differ from other authors who observed, when comparing measurements in metastasis, that the q value did not change significantly, suggesting that the variation in FA may be due to changes in the magnitude of the total diffusion (L) rather than the anisotropy (q). The inverse correlation between FA and RD observed in our study had been reported,(39) although there are not statements of the expected magnitudes among these correlations.

Several limitations in our study and factors that influence the clinical application of DTI-metrics need to be addressed: our decision to evaluate a single, *global* approach among tensor-metric relationships came up from reports depicting a non-significant difference in FA between the enhancing and non-enhancing peritumoral *regions*;(42) no difference in MD and FA values in peritumoral areas with T2 signal-intensity change,(43) and findings of high variance of FA values between several brain regions,(5) making all them a patent lack of agreement between investigators. Also, some studies have evaluated only small regions of interest within the tumors, which may miss much of the tumor or the peritumoral abnormalities, and those that have used regions of interest that cover the entire tumor, have not always studied the region closest to the tumor, which should contain the most infiltrated white matter tracts.(5)

Nowadays, there is still scarce evidence regarding the potential of these DTI biomarkers, for example, besides the clinical acceptance of FA; some studies have showed a marked increase in p (mean 68%) with marked reduction in q (mean 42%) in disrupted white matter.(44) Also, some proved biomarkers, for example, the increase of RD as a marker of demyelination and axonal loss has not had generalization in its use.(39) These facts reflect the limited acceptance of DTI by the medical community; it does not reach yet the anatomic validity of the myeloarchitectonic studies and currently cannot differentiate individual axons or synaptic connections.(45) The intricacies of DTI-derived tensor metrics to characterize the brain tissue in health and disease might be affected among other variables by the ratio of extracellular to intracellular space, vascularity, edema, microcysts, and extracellular matrix.(46)

We most acknowledge that a deep physics explanation of the several relationships presented in this research is beyond not only from the scope of the article but also from our knowledge as clinical investigators. Some readers might prefer more complex multivariate analyses such as anova and linear regression, we considered them, but they would require additional analyses and explanations. Our study is part of a research line trying to understand the role of tensor metrics in characterizing brain impairment

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due to brain tumors; we are aware that, the clinical value of our findings has yet to be determined, and the biological impact of the different metrics should be explained in more detail in clinical journals.

In conclusion, our study make a contribution toward the understanding of the fundamentals of the tensormetric's relationships; the presented data that will help to know *a priori* which relationships can be expected in a prospective analysis decreasing the underlying empiricism in this area. A comprehensive understanding of the currently available DTI-derived tensor metrics will allow them to be considered as imaging biomarkers in the diagnosis and treatment planning of brain tumors. Readers should be aware that, it is still not completely understood the different relationships of the brain structure with each tensor metric, it is possible they answer different questions; also, variations in DTI measures are not specific of one histologic type of tumor, which broaden the application of these biomarkers to a wider variety of intracranial pathologies. Given the increased availability of open source software in MRI units around the world, it is anticipated that DTI-derived tensor metrics may become a low-cost and common used approach. A standardized package of selected DTI processing tools might be included as an imaging protocol to be routinely performed in clinical settings, however this approach still needs further work and histologic examination to assess the depth and extent of its application and finally obtain objective and reproducible results. **Conflict of interest**

Acknowledgments

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Tables

Table 1. DTI-derived tensor metric formulas.

$$Mean Diffusivity (MD)$$

$$MD = D = \frac{\lambda_1 + \lambda_2 + \lambda_3}{3}$$
Fractional Anisotropy (FA)
$$FA = \sqrt{\frac{3}{2}} \frac{1}{L} = \sqrt{\frac{3}{2}} \sqrt{\frac{(\lambda_1 - D)^2 + (\lambda_2 - D)^2 + (\lambda_3 - D)^2}{\lambda_1^2 + \lambda_2^2 + \lambda_3^2}}$$
Relative Anisotropy (RA)
$$RA = \frac{q}{p} = \frac{\sqrt{(\lambda_1 - D)^2 + (\lambda_2 - D)^2 + (\lambda_3 - D)^2}}{\sqrt{3}D}$$
Radial Diffusivity (RD)
$$RD = \frac{\lambda_2 + \lambda_3}{2}$$
Axial Diffusivity (AD)
$$AD = \lambda_1$$
Spherical Tensor (Cs)
$$Cs = \frac{3\lambda_3}{\lambda_1 + \lambda_2 + \lambda_3}$$
Pure Isotropic Diffusion (p)
$$p = \sqrt{3}D = \frac{\lambda_1 + \lambda_2 + \lambda_3}{\sqrt{3}}$$
Pure Anisotropic Diffusion (q)
$$q = \sqrt{(\lambda_1 - D)^2 + (\lambda_2 - D)^2 + (\lambda_3 - D)^2}$$
Total Magnitude of the Diffusion Tensor (L)
$$L = \sqrt{p^2 + q^2} = \sqrt{\lambda_1^2 + \lambda_2^2 + \lambda_3^2}$$

$$Linear Tensor (Cl)$$

$$Cl = \frac{\lambda_1 - \lambda_2}{\lambda_1 + \lambda_2 + \lambda_3}$$

$$Planar Tensor (Cp)$$

$$Cp = \frac{2(\lambda_2 - \lambda_3)}{\lambda_1 + \lambda_2 + \lambda_3}$$

	Tensor Metr	ic											Mean	SD
		Cs	-										.756690	.032259
FA	Pearson's R <i>p</i> -value	456 .029	FA										.284317	.018917
RA	Pearson's R <i>p</i> -value	992 .000	.434 .039	RA									.224873	.029523
Ср	Pearson's R <i>p</i> -value	979 .000	.460 .031	.938 .000	Ср								.133346	.017086
Cl	Pearson's R <i>p</i> -value	946 .000	.442 .024	.978 .000	.855 .000	Cl							.111574	.016583
L	Pearson's R <i>p</i> -value	.372 .117	115 .578	416 .076	317 .200	186 .408	L						.002275	.000100
р	Pearson's R <i>p</i> -value	.179 .477	095 .652	214 .395	020 .938	014 .953	.945 .000	р					.002096	.000087
AD	Pearson's R <i>p</i> -value	.322 .179	012 .952	348 .144	287 .249	.019 .933	.804 .000	.761 .000	AD				.001553	.000056
MD	Pearson's R <i>p</i> -value	.181 .473	096 .649	215 .391	022 .934	014 .953	.945 .000	1.000 .000	.762 .000	MD			.001210	.000050
RD	Pearson's R <i>p</i> -value	.053 .821	297 .132	119 .606	.426 .054	131 .542	.877 .000	.954 .000	.628 .001	.954 .000	RD		.001046	.000072
q	Pearson's R <i>p</i> -value	077 .721	.698 .000	.019 .929	.257 .225	.138 .492	.138 .502	.153 .466	.393 .047	.152 .467	.406 .029	q	.000445	.000055

Table 2. Means, SD and correlations (controlled for the effect of age and gender) in healthy brains.

	Tensor Metr	ic											Mean	SD
		Cs	-										.771562	.066812
FA	Pearson's R <i>p</i> -value	570 .003	FA										.253531	.028425
RA	Pearson's R <i>p</i> -value	-1.000 .000	.592 .002	RA									.201778	.053287
Ср	Pearson's R <i>p</i> -value	936 .000	.576 .003	.901 .000	Ср								.133265	.043503
CI	Pearson's R <i>p</i> -value	904 .000	.819 .000	.969 .000	.803 .000	Cl							.098462	.011929
L	Pearson's R <i>p</i> -value	.096 .648	.151 .472	.009 .967	159 .446	.182 .469	L						.002111	.000140
р	Pearson's R <i>p</i> -value	.156 .456	129 .540	094 .664	216 .299	177 .482	.834 .000	р					.001961	.000123
AD	Pearson's R <i>p</i> -value	111 .596	.343 .094	.191 .372	.053 .803	.173 .492	.871 .000	.881 .000	AD				.001397	.000080
MD	Pearson's R <i>p</i> -value	.154 .462	125 .550	090 .675	215 .303	173 .491	.837 .000	1.000 .000	.883 .000	MD			.001132	.000071
RD	Pearson's R <i>p</i> -value	.305 .139	411 .041	260 .219	360 .077	395 .105	.706 .000	.954 .000	.699 .000	.953 .000	RD		.000999	.000073
q	Pearson's R <i>p</i> -value	484 .014	.928 .000	.539 .007	.466 .019	.736 .001	.430 .032	.187 .371	.629 .001	.191 .361	114 .588	q	.000363	.000048

Table 3. Means, SD and correlations (*controlled* for the effect of age and gender) in brains with GBM.
Tables 4. Estimated marginal means SE and CI of DTI-derived tensor metrics (the effect of age was controlled at the value of 43.92 yr).

		Health	y brains			Brains wi	th GBM	
Tensor metric	Maaa	Std.	95%	CI	Maaa	Std.	95%	CI
	Mean	Error	Lower	Upper	Mean	Error	Lower	Upper
Cs	.739511	.009798	.719577	.759446	.766043	.007824	.750126	.781960
FA	.290160	.005721	.278520	.301799	.255660	.004568	.246366	.264954
RA	.241025	.008008	.224734	.257317	.211498	.006394	.198490	.224506
Ср	.141366	.007941	.125210	.157522	.137961	.006341	.125061	.150860
Cl	.119041	.003243	.112443	.125638	.099495	.002589	.094227	.104763
L	.002287	.000029	.002227	.002347	.002103	.000024	.002055	.002151
р	.002122	.000027	.002068	.002177	.001946	.000021	.001903	.001990
AD	.001558	.000020	.001518	.001598	.001395	.000016	.001363	.001426
MD	.001225	.000015	.001194	.001257	.001124	.000012	.001099	.001149
RD	.001059	.000015	.001029	.001089	.000988	.000012	.000964	.001012
q	.000454	.000011	.000432	.000476	.000371	.000009	.000354	.000389

Figure Legends

Fig. 1. Example of some sequences and DTI maps from healthy brains (upper row) and brains with GBM (lower row): a and d, T1-postgadolinium images; b and e, Flair sequence; c, pure isotropic diffusion; f, color map of the V1-vector.

Fig. 2. Scatter matrix of the variable's data grouped by diagnosis.

Fig. 3. Graphs of the estimated marginal means for each tensor metric, the influence of age was controlled at the value of 43.92 years.







Original article

Diffusion tensor imaging-derived measures of fractional anisotropy across the pyramidal tract are influenced by the cerebral hemisphere but not by gender in young healthy volunteers: a split-plot factorial analysis of variance

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Keywords: anisotropy; diffusion tensor imaging; magnetic resonance imaging; multivariate analysis; analysis of variance

Background Diffusion tensor imaging (DTI) permits quantitative examination within the pyramidal tract (PT) by measuring fractional anisotropy (FA). To the best of our knowledge, the inter-variability measures of FA along the PT remain unexplained. A clear understanding of these reference values would help radiologists and neuroscientists to understand normality as well as to detect early pathophysiologic changes of brain diseases. The aim of our study was to calculate the variability of the FA at eleven anatomical landmarks along the PT and the influences of gender and cerebral hemisphere in these measurements in a sample of young, healthy volunteers.

Methods A retrospective, cross-sectional study was performed in twenty-three right-handed healthy volunteers who underwent magnetic resonance evaluation of the brain. Mean FA values from eleven anatomical landmarks across the PT (at centrum semiovale, corona radiata, posterior limb of internal capsule (PLIC), mesencephalon, pons, and medulla oblongata) were evaluated using split-plot factorial analysis of variance (ANOVA).

Results We found a significant interaction effect between anatomical landmark and cerebral hemisphere (F (10, 32)=4.516, P=0.001; Wilks' Lambda 0.415, with a large effect size (partial η^2 =0.585)). The influence of gender and age was non-significant. On average, the midbrain and PLIC FA values were higher than pons and medulla oblongata values; centrum semiovale measurements were higher than those of the corona radiata but lower than PLIC.

Conclusions There is a normal variability of FA measurements along PT in healthy individuals, which is influenced by regions of interest location (anatomical landmarks) and cerebral hemisphere. FA measurements should be reported for comparing same-side and same-landmark PT to help avoid comparisons with the contralateral PT; ideally, normative values should exist for a clinically significant age group. A standardized package of selected DTI processing tools would allow DTI processing to be routinely performed in clinical settings.

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The evaluation of the integrity of motor fibers within the pyramidal tract (PT) as the main descending fiber ² plays an important role in measuring the degree bundle,^{1,} of motor impairment; this impairment can be measured with standardized, reliable, and valid clinical assessments.^{3,4} An advanced magnetic resonance (MR) assessment of the PT can confirm lesion location, but this finding may not sufficiently demonstrate the extent to which motor fibers remain intact.⁵ For accurate MR assessment. additional variables should also considered in the investigation of motor impairment, specifically: lesion size,^{6,7} the amount of previous lesion burden, and other factors such as patient age and comorbidities.8,9

Damage to major axonal pathways can now be studied noninvasively using diffusion tensor imaging (DTI), which permits the imaging and quantitative examination of major axonal pathways of the living brain and provides information about tissue microstructure by measuring fractional anisotropy (FA);^{10,11} this permits the topographic relation of lesion location and corticospinal fibers to be evaluated.^{12,13} FA is an index of the diffusion characteristics of water molecules preferentially directed along the axis of major axonal pathways.¹⁰

Reduced FA has been interpreted as evidence of nerve degeneration and axonal loss.^{14,15} Despite numerous studies that have reported FA altered values at a variety of

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clinical entities (e.g., within stroke lesions), significant decreases in FA have been demonstrated as early as 7 to 14 hours poststroke.¹⁶ In addition, secondary degeneration of the PT-altered FA has been recorded over a 12-week period,¹⁷ yet at present little is known about specific changes in FA along the entire tract based on anatomical landmarks, gender, or cerebral hemispheres.^{18,19}

To the best of our knowledge, there do not exist any published findings describing the normative values of FA as it relates to the regions along the PT; that is, the inter-variability measures of FA along the PT remain unexplained. Establishment of these reference values would help radiologists and neuroscientists to understand normality as well as early pathophysiologic changes associated with a variety of brain diseases.

The aim of our study was to know the variability of the FA at eleven anatomical landmarks along the PT, and to demonstrate the possible influences of gender and/or the cerebral hemisphere in these measures in a sample of young, healthy volunteers. We sought to evaluate these influences in order to obtain DTI-derived reference-biomarker values of motor tract integrity, which we expect could be compared to patterns of motor tract damage in future studies.

METHODS

Subjects

A retrospective, cross-sectional study was performed in twenty-three right-handed healthy volunteers (medical students at our institution) for MR evaluation of the brain: 9 men ((26.8±2.1) years; range, 24-30 years) and 14 women ((24.94±2.75) years; range, 20-30 years). A preliminary neuropsychological evaluation included the Wechsler Adult Intelligence Scale (WAIS) III intelligence quotient (IQ) test validated for Mexican patients (considering an average IQ of 100 and a standard deviation of 15).^{20,21} In addition, a validated, modified version of the Mini-Mental State Examination (MMSE) in Spanish was used (score greater than or equal to 25 points was considered normal),²² and the presence of depression was assessed with the Beck's Depression Inventory-II (BDI) test (subjects with cutoff test values below 13 points were included).²³ The local Institutional Review Board approved the study, and all participants provided written informed consent.

Brain image acquisition

Conventional MR evaluation of the brain was performed using a 3-Tesla Signa HDxt scanner (GE Healthcare, Waukesha, WI, USA) and a high-resolution eight-channel head coil (Invivo, Gainesville, FL, USA). Contraindications to magnetic resonance imaging (MRI) were claustrophobia and the presence of a pacemaker or metallic implant. Exclusion criteria considered any potential medical disorders that could affect brain structure or function as well as subjects with brain-structural abnormalities on their MRI scans. All participants were included.

The examination included standard clinical sequences: sagittal T1-weighted Fluid-Attenuated Inversion Recovery (FLAIR) (TE/TR = 9.9 ms/2500 ms) with a 5/3-mm slice thickness/gap and a field of view (FOV) of 24×24 ; an axial Spoiled Gradient Echo (FSPGR) sequence (TE/TR = 3.9 ms/9.4 ms) with a 1.3/0-mm slice thickness/gap and an FOV of 24 × 18; a coronal T2-weighted fast spin-echo (FSE) sequence (TE/TR = T_{T} 164.1 ms/2617 ms) with a 3/0-mm slice thickness/gap and an FOV of 22×16 ; and a FLAIR sequence (TE/TR = 115.8 ms/11002 ms) with a 5/1-mm slice thickness/gap and an FOV of 22×22 . The DTI sequence resulted in 30 axial slices covering the entire brain and brainstem with 1.87 mm \times 1.87 mm \times 5.0 mm voxel size, acquiring 25 noncollinear diffusion directions with a b-value of 1000 s/mm^2 , as well as one with a b-value of 0 s/mm^2 .

DTI analysis and FA measurements

We used two main tools for our imaging analysis: dcm2nii²⁴ to convert our data from DICOM to NIfTI formats and FMRIB Software Library (FSL).^{25,26} to calculate the FA, normalize the data to Montreal Neurological Institute (MNI) space,²⁷ and select regions of interest (ROIs). Once we had obtained the subjects' diffusion images in DICOM format,²⁸ we proceeded to convert these files into the NIfTI format.²⁹

DTI images were brain-extracted using the standard brain extraction tool (BET).^{30,31} Eddy current effects were corrected by taking the first volume as a reference using FSL's Diffusion Toolbox. The diffusion tensor model was adjusted to generate FA images for each subject. All subjects' FA maps were then aligned into the MNI's common space using the nonlinear registration tool FNIRT, which uses a b-spline representation of the registration warp field.³²

Eleven ROIs were selected on axial color-coded FA images, generating eleven masks. These eleven anatomic-landmarks corresponded approximately to the next coordinates in the space of the MNI brain atlas:²⁷ centrum semiovale (Z=+45 and Z=+55); corona radiata (Z=+35 and Z=+25); posterior limb of internal capsule (PLIC) (Z=+15 and Z=+05); mesencephalon (Z=-05 and Z=-15); pons (Z=-25 and Z=-35) and medulla oblongata (Z=-45); all these data points are shown in Figure 1. All ROIs were distributed along the PT according to the John Hopkins University Tractography Atlas.³³ Mean FA values for each subject were extracted from corresponding maps and masks using *fslstats*.²⁶ Figure 2 shows the algorithm used in DTI measurements.

Clinical significance of DTI sequence and FA measurements

The use of the DTI sequence permits the noninvasive delineation of detailed anatomical and functional



Figure 2. Algorithm showing the software used in DTI-derived FA measurements.

information of both normal and pathological brain tissues. DTI data can be used to visualize the major white matter (WM) tracts of the brain,^{34,35} and indirectly evaluate the integrity of WM by measuring water diffusion and its directionality in three dimensions.³⁶

FA provides information about the shape of the diffusion tensor in each voxel. The anisotropic fraction is based on the normalized variance of the core values of the main vectors of diffusion (corresponding to the eigenvectors of the tensor),³⁷ its values vary between 0 (isotropic diffusion) and 1 (infinite anisotropy).³⁸ Diffusion is anisotropic in WM fiber tracts, as axonal membranes and myelin sheaths present barriers to the motion of water molecules in directions not parallel to their own orientation. Based on previous reports,³⁹ we considered values >0.2 as indication of oriented structure within the normal WM (myelination and axonal packing); however, we did not define superior or inferior limits to the expected FA values of PT, as these measurements were the purpose of this study. Reductions in FA values may result from fiber depletion (tumor destroy fibers, reducing their absolute number); fiber dilution (tumor or vasogenic edema spreads intact fibers apart, reducing their density); or fiber degradation (fibers degrade, fibers themselves become intrinsically less anisotropic and thus retain normal numbers and density).⁴⁰

Statistical analysis

A split-plot factorial analysis of variance (ANOVA)^{41,42} was conducted to assess the influences of gender and the cerebral hemisphere on FA measures across the chosen anatomical landmarks along the PT. Subject age was used

Figure 1. Selected anatomical landmarks for the DTI-derived FA measurements.

as a covariate to control for individual differences. Preliminary checks were conducted to ensure that there was no violation of the general assumptions underlying univariate and multivariate ANOVA results, sphericity, and reliable measurement of the covariate. The effect size assessment (proportion of the variance in the dependent variable that can be explained by the independent variable) of each of the results was obtained by using the Partial Eta Squared (η^2) proposed by Cohen,⁴³ where 0.01-0.06 = small effect; 0.06-0.14 = moderate effect; and a value greater than 0.14 = large effect.

Variation of the initial FA measurements was shown using descriptive statistics (mean \pm standard deviation (SD)); however, the final estimated marginal means of the factorial ANOVA analysis were reported using the mean \pm standard error (SE). Here we use SE instead of SD, as we consider that it is a probabilistic statement about how a specific number of samples would provide a better bound on estimates of the population mean, in light of the central limit theorem. SE was calculated by dividing the SD by the square root of N (the sample size), SE = SD/ \sqrt{n} .⁴⁴

All analyses were carried out using the IBM[®] SPSS[®] Statistics software (version 20.0.0 IBM Corporation; Armonk, NY, USA); for the graphics design, we used the STATISTICA data analysis software system (version 8.0; StatSoft, Inc.; Tulsa, OK, USA). Data were presented in accordance with the guidelines of the American Psychological Association.^{45,46} Statistical significance was indicated by P < 0.05 (two-tailed).

RESULTS

Subjects and FA measurements

Neuropsychological tests revealed that the corresponding IQs or other variables of mental status were comparable across the age range of the subjects. The PT was successfully traced from the medulla oblongata to the centrum semiovale with 11 FA measurements taken across the chosen anatomic landmarks. Forty-six PTs (from 23 subjects) representing a total of 506 FA measurements were included in the split-plot ANOVA analysis. The mean values and SD for the FA measurements totals across each anatomical landmark are

presented in Table 1.

Interactions, main effects, and effect sizes on FA measurements

The assumption of homogeneity of variance-covariance matrices was interpreted as non-significant (Box's M value=109.137, P=0.330), assuming the covariance matrices between the groups were equal.⁴⁷ The homogeneity of variance at each anatomical landmark was also non-significant (P > 0.05). The assumption of sphericity however, was found to be violated (P < 0.05). In light of this assumption violation, the findings of the multivariate testing were reported.

There were no significant interaction effects amongst anatomical landmark, gender, and cerebral hemisphere and the anatomical landmark and gender, P > 0.05. However, there was a significant interaction effect between Anatomical landmark and cerebral hemisphere (F (10, 32)=4.516, P=0.001; Wilks' Lambda 0.415, with a large effect size (partial $\eta^2 = 0.585$)). The covariate age was non-significant in its interaction with the anatomical landmarks, P=0.055. Finally, there was a significant main effect of the anatomical landmarks (F (10, 32)=8.313, P<0.001; Wilks' Lambda 0.278, with the largest effect size (partial $\eta^2 = 0.722$)).

The between-subject effects depicted a non-significant main effect for groups; that is, there were not significant differences in the FA measurements for each group (cerebral hemisphere with P=0.297 and gender with P=0.336). Table 2 presents a summary of the within- and between-subjects effects assessment.

Table 1. Means and SD for FA measurements across the	PT :
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Anatomical landmarks	Ν	Mean±SD
Medulla oblongata ($Z = -45$)	46	0.476±0.089
Pons ($Z = -35$)	46	0.493 ± 0.070
Pons ($Z = -25$)	46	$0.594{\pm}0.039$
Midbrain ($Z = -15$)	46	0.639±0.068
Midbrain ($Z = -05$)	46	0.646±0.032
PLIC ($Z = 05$)	46	0.643±0.023
PLIC ($Z = 15$)	46	0.599±0.028
Corona radiata (Z = 25)	46	0.499±0.025
Corona radiata (Z = 35)	46	0.456±0.028
Centrum semiovale ($Z = 45$)	46	0.563±0.035
Centrum semiovale ($Z = 45$)	46	$0.520{\pm}0.048$

Estimated marginal means

Considering the significant interaction between anatomical landmarks and cerebral hemisphere, marginal means were clustered by groups. An inspection of the mean scores indicated that except for the pons (Z=-35) and corona radiata (Z=+35), FA measurements were significantly higher in the right cerebral hemisphere. The post hoc analyses (Bonferroni's) revealed significant differences between mean comparisons across all eleven anatomical landmarks (P < 0.05). The trend analysis showed the FA measurements with a non-linear (cubic) trend, P=0.030, partial η^2 =0.110. That is, on average, the midbrain and PLIC FA values were higher than those of the pons and medulla oblongata; centrum semiovale measurements were higher than those of the corona radiata but less than PLIC. Table 3 depicts the estimated marginal means and SE adjusted for cerebral hemispheres and the covariate age, and Figure 3 depicts the comparative graphs of marginal means.

DISCUSSION

Achieving a better understanding of normative regional FA values is of particular importance to those specialists participating in imaging the PT, especially considering that recent studies foster the use of DTI-derived measures alone as valid structural surrogates of motor impairment.⁵

The clinical relevance of our study, specifically, the evidence of normal variability of FA measurements across PT in a sample of healthy subjects, has three main components: (1) we accomplished the mapping and measurement of FA along fiber tracts in a reliable, standardized way by selecting a group of open-source software tools for DTI analysis (Figure 2); (2) we confirmed the existence of significant regional FA variability along the PT, which had a main effect behavior in our analysis; (3) we discovered evidence of the significant influences of the cerebral hemisphere in the FA measurements (interaction of anatomical-landmark and cerebral hemisphere). These findings are in agreement with previous reports of asymmetries in fiber density and FA values for the arcuate fasciculus,⁴⁸ a finding consistent with lateralization of brain functions in most humans. Also,

Table 2. Within- and between-subjects effects
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Main effects and interaction	Wilks' Lambda	F values	P values	Effect size (partial η^2)
Within-subjects effects				
Anatomical landmark * Gender * Cerebral Hemisphere	0.672	1.563	0.163	0.328
Anatomical landmark * Gender	0.675	1.544	0.169	0.325
Anatomical landmark * Cerebral Hemisphere	0.415	4.516	0.001	0.585
Anatomical landmark * Age	0.604	2.099	0.055	0.396
Anatomical landmark	0.278	8.313	< 0.001	0.722
Between-subjects effects				
Gender * Cerebral Hemisphere	_	0.029	0.865	0.001
Gender	_	0.946	0.336	0.023
Cerebral Hemisphere	_	1.116	0.297	0.027
Age	_	3.318	0.076	0.075

Design = Intercept + age + gender + cerebral hemisphere + gender * cerebral hemisphere. * = interaction.

Table 3. Means and SE adjusted for the significant interactions found in the study: anatomical landmarks and cerebral hemispheres

An atomical lon descelo	Right	nemisphere	Left he	emisphere
Anatomicai fandmarks	Mean±SE	95% CI	Mean±SE	95% CI
Medulla oblongata ($Z = -45$)	0.448 ± 0.022	0.403-0.493	0.467 ± 0.022	0.422-0.511
Pons ($Z = -35$)	0.496±0.018	0.460-0.532	0.511±0.018	0.474-0.547
Pons ($Z = -25$)	0.605 ± 0.008	0.589-0.622	0.568 ± 0.008	0.551-0.585
Midbrain ($Z = -15$)	0.647±0.018	0.611-0.683	0.637±0.018	0.601-0.673
Midbrain ($Z = -05$)	0.647 ± 0.008	0.630-0.664	0.641 ± 0.008	0.624-0.658
PLIC ($Z = 05$)	0.641 ± 0.006	0.629-0.654	0.640 ± 0.006	0.628-0.652
PLIC ($Z = 15$)	0.618 ± 0.005	0.607-0.628	0.578 ± 0.005	0.568-0.589
Corona radiata (Z = 25)	0.504±0.006	0.491-0.516	0.488 ± 0.006	0.476-0.501
Corona radiata (Z = 35)	0.451±0.007	0.437-0.465	0.453 ± 0.007	0.439-0.468
Centrum semiovale $(Z = 45)$	0.570±0.009	0.552-0.589	0.553±0.009	0.535-0.572
Centrum semiovale ($Z = 45$)	0.536±0.011	0.515-0.558	0.506±0.011	0.485-0.528

The covariate (age) was evaluated at the value of 25.22 years.



Figure 3. Estimated marginal means of FA at selected anatomical landmarks of the PT clustered by cerebral hemisphere (error bars = 95% *CI*). The crossing lines (*) evidence the interactions between anatomical landmarks and cerebral hemispheres. These interactions support the influence of ROI location in the FA measurements.

our observations of similar measurements at certain anatomical landmarks, for example the FA values of PLIC at Z=+05, support previous reports of intact ascending and descending fiber tracts that traverse through the PT in close proximity.^{49,50}

Although the image-processing pathway presented here are not at present well known by experts in the field, we considered it noteworthy to demonstrate that we were able to achieve standardized measurements. The use of FSL's tools and scripts,³⁰ predesigned ROIs from Johns Hopkins University's White Matter Atlas,³³ and drawings of our own ROIs (we preferred to draw specific masks around the desired ROIs instead of using the skeletonized image generated by the TBSS procedure⁵¹ for some anatomic landmarks not included in the JHU atlas); permitted us to avoid common errors such as misidentification of cerebral structures and data alteration by other undesirable ROIs (e.g., gray-matter or adjacent white-matter tracts). The sequence of computational algorithms that we used here allowed us to conduct FA measurements within the same area and cerebral landmarks in all patients; we sought the maximum accuracy in our final results, so variations in the results were only a reflection of the distribution of the FA values in the selected PT.

We considered the quality of our data reliable as normative values; we presented the corresponding FA values at each of the eleven-tier anatomical landmarks using mean, SE, and *CI*. To avoid confusion with SD, the reader must remember that SE represents how well the sample mean approximates the population mean (the larger the sample, the smaller the SE; the closer the sample mean approximates the population mean). In practical situations, because the true value of the SD is usually unknown, the SE is often used to refer to an estimate of this unknown quantity.⁴⁴ Also considering that the probability distribution of the values in this study was known, SE was used to calculate a good approximation to 95% exact *CI*.

Our study differs methodologically over previous ones, in that we used a split-plot ANOVA test controlling the age as a covariate; this analysis allowed us to identify the interaction between anatomical landmarks and the cerebral hemisphere, that is, the difference between FA measurements will depend on group membership; also, the significant main effect of the anatomical landmarks means that the imaging evaluation of the PT integrity should be conducted by reading the FA values at each landmark.

This kind of imaging analysis might be of special importance. considering that previous several publications have reported only main effects; however, a formal statistical analysis considering simultaneously independent variables that can influence results (age, cerebral hemisphere, and gender) has not been reported, to the best of our knowledge. These moderator variables are noteworthy because they might help explaining why some researchers obtain statistically significant results while others do not. Although the anatomical landmarks of our study depicted a large effect size, the observed 72.2% of variance should not stop neuroscientists and clinicians from considering additional interacting factors to discover additional variance.

Although it is currently accepted that axonal membranes and myelin sheets are the primary determinant of anisotropic diffusion along nerve fibers,¹⁴ some reports consider that the variability of FA observed along the PT may be in fact associated with microstructural disruption of axonal transport; such disruption causes a significant disturbance of directed axonal protoplasmatic flow.⁵² The low FA values observed across some landmarks of the PT would mean that other tracts contribute to the diffusion signal decay in these healthy subjects. In this study, the higher FA values were observed in the regions between pons and internal capsule; the corona radiata, a region including crossing fibers from the superior longitudinal fasciculus (SLF), reflects a much lower FA; our explanation for this finding is that voxels do not represent just one tract but a sample of both the PT and the SLF (Figure 3).

In addition, when there is a nonuniform distribution of fiber directions, the eigenvector associated with the largest eigenvalue only corresponds to the averaged fiber direction within the voxel; this may not necessarily reflect the nerve fibers crossing the voxel.⁴⁸ Such mixture of tracts reflects a lower FA value because diffusion is not as anisotropic as it is in other regions where one single tract is represented, such as the PLIC. When PT is evaluated in patients, a low regional FA finding might not necessarily represent direct damage to the tract; other possibilities such as vasogenic edema and tract compression by a tumoral mass could be responsible for this FA finding.⁴⁸

Some limitations to our study and factors that influence the accuracy of DTI in clinical applications need to be addressed, specifically: (1) Although the age range of our study subjects could seem limited, we intentionally did not compute a wider age group (young and elderly subjects). As age could be a variable that influences the calculation of FA, we wanted to study a group of young, healthy volunteers to help us understand the regional FA variability in a homogeneous sample prior to attaining measures of FA in pathological scenarios. The fact that even in this limited age-range group we found significant differences in FA measurements supports our statement that careful analysis must be taken into account by radiologists and/or bio-imaging experts when reporting DTI-derived measurements to clinicians (neurologists, neurosurgeons, psychiatrist, pediatricians, etc.), as these results might influence diagnosis and treatment. (2) The factorial ANOVA analysis of FA measurements was performed considering the number of pyramidal tracts and not subjects - analyzed (after proof of normal distribution of the data), which should ease some concerns about the apparently limited sample size used in this study. (3) There are still some intrinsic limitations in DTI acceptance by the medical community, as this technique has not yet achieved the anatomic validity of myelo-architectonic studies and currently cannot differentiate individual axons or synaptic connections;⁵³ furthermore, there is still no gold standard for in vivo

fiber tracking.⁵⁴⁻⁵⁶ Finally, although the path of PT is well known and can be reliably reconstructed,³³ there is a coexistence of alternate motor fibers,¹ and the course of these fibers is not yet based on anatomic studies in humans and were not discriminated in our study (the cortico-reticulo-spinal and cortico-rubro-spinal tracts).^{57,58}

There is a normal variability of FA measurements along PT in healthy individuals, which is influenced by the location of ROIs (anatomical landmarks) and the cerebral hemisphere. The use of FA measurements as a biomarker in clinical settings will allow the follow-up of patients and comparison with controls by comparing only same-side and same-landmark PT. The comparisons of FA measurements with the contralateral PT ROI should be avoided when investigators are interested in accurate reports.

Ideally, normative values should exist for specific age and ethnic groups; further studies examining normal variations might provide additional understanding of brain physiology. Given the increased availability of open-source software in MRI units around the world, it is anticipated that DTI-derived FA measurement may become a low-cost and commonplace biomarker. A standardized package of selected DTI processing tools would allow tractography to be routinely performed in clinical settings. The combined use of DTI-derived FA measurements with other conventional MRI modalities such as 3D-volumetry, spectroscopy, and functional MRI Bold needs to have an integrative approach in order to obtain objective and reproducible results.

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NEURO

Diagnostic performance of regional DTI-derived tensor metrics in glioblastoma multiforme: simultaneous evaluation of p, q, L, Cl, Cp, Cs, RA, RD, AD, mean diffusivity and fractional anisotropy

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Abstract

Objectives Almost a dozen diffusion tensor-imaging (DTI) variables have been used to evaluate brain tumours with scarce information about their diagnostic ability. We aimed to perform a comprehensive evaluation of tensor metrics reported in the last decade.

Methods Retrospective case control study performed in 14 patients with glioblastoma multiforme (GBM) and 28 controls. Conventional brain MR sequences and image postprocessing of DTI allowed the calculation of: MD, FA, p, q, L, Cl, Cp, Cs, RA, RD and AD, classified into five regions:

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C. Rios Department of Neurochemistry, National Institute of Neurology and Neurosurgery, Mexico City, Mexico normal appearance white matter (NAWM), immediate and distant oedema, enhancing rim and cystic cavity. ANOVA and AUROC analyses were performed.

Results ANOVA depicted a significant difference among all metrics (p<0.05). RA had the highest performance in the NAWM and cystic cavity; immediate and distant zones of oedema were best diagnosed by RD and Cp respectively; q was the best biomarker of the enhancing rim zone; p<0.001 for all metrics.

Conclusions FA and MD, accepted biomarkers of brain injury, were surpassed by other metrics. RA, together with Cs, Cl and CP, might be the new leaders in the evaluation of brain tumours. DTI tensor metrics depict different clinical applicability at each tumour region.

Key Points

- DTI-derived tensor metrics can characterise the regional impairment of intraaxial brain tumours.
- A performance evaluation of new diagnostic tests should follow the STARD initiative.
- Each tumour region in GBM is detected by a different tensor metric.
- FA and MD are not the best biomarkers of tumour infiltration.
- The diagnostic performance of some tensor metrics allows them to be used interchangeably.

Keywords Brain neoplasms · Diffusion tensor imaging · Magnetic resonance imaging · Sensitivity and specificity · Statistics as topic

Abbreviations

AD	axial diffusivity
ADC	apparent diffusion coefficient
Anova	analysis of variance
AUROC	area under the ROC curve
CI	confidence interval
Cl	linear tensor
Ср	planar tensor
Cs	spherical tensor
DTI	diffusion tensor imaging
FA	fractional anisotropy
GBM	glioblastoma multiforme
L	total magnitude of the diffusion tensor
MD	mean diffusivity
NAWM	normal-appearing white matter
р	pure isotropic diffusion
q	pure anisotropic diffusion
RA	relative anisotropy
RD	radial diffusivity
ROC	receiver operating characteristic
SE	standard error
STARD	Standards for Reporting of Diagnostic Accuracy
	initiative
VEGF	vascular endothelial growth factor
WM	white matter

Introduction

Advanced MR techniques that produce image contrast reflecting attributes of tissue physiology and microstructure have begun to be widely applied in clinical brain tumour imaging at major academic centres [1]. Diffusion tensor imaging (DTI) provides quantitative measures of the molecular motion of water in a three-dimensional space [2]. The measure of this directionality [3] has been accepted as a quantitative index for diffusion anisotropy that correlates with microstructural integrity of myelinated fibre tracts [4]. Three eigenvectors and three eigenvalues are used to represent the main diffusion directions and magnitudes, respectively [5], by combining the eigenvalues λ_1 , λ_2 and λ_3 ; several scalar measures of diffusion have been proposed: mean diffusivity (MD), fractional anisotropy (FA), pure isotropic diffusion (p), pure anisotropic diffusion (q), the total magnitude of the diffusion tensor (L), linear tensor (Cl), planar tensor (Cp), spherical tensor (Cs), relative anisotropy (RA), axial diffusivity (AD) and radial diffusivity (RD) [6-10]. Some of these DTI-derived variables have been studied in brain tumours [7, 8]; however, there is still scarce information about the performance of most of these tensor metrics in the diagnosis of the infiltration zones caused by glioblastoma multiforme (GBM).

We conducted this study to compare the diagnostic accuracy among DTI-derived tensor metrics in the detection of regional tumour infiltration by using the infiltration pattern observed in the T2- and T1-weighted post-contrast images as the reference standard [11]. Our data intend to be a small contribution in understanding the clinical applicability of these available quantitative biomarkers of brain-tumour impairment, which could accelerate the adoption of these image-data analyses in clinical settings.

Material and methods

Subjects

Retrospective study using a case and control design. Inclusion criteria considered preoperative brain MR examinations between January 2011 and May 2012 of patients with at first (suspected) diagnosis and later pathological confirmation of primary GBM. Exclusion criteria applied to corticosteroid or antibiotic treatment, lesions with areas related to calcification and/or haemorrhage and previous brain surgery. A control group included young and elderly healthy volunteers recruited among the enrolled interns and medical residents of the hospital as well as elderly subjects from our Geriatrics unit. All volunteers received detailed health examinations; exclusion criteria considered major neurological, psychiatric or cardiovascular diseases. Two radiologists (E.R.V., M.M.L.) interpreted the MR images blinded to the patient's history. MRI examinations with other structural abnormalities were excluded. The local Institutional Review Board approved the study (project no. 2011.044).

Brain image acquisition

MR sequences included conventional axial T2-weighted imaging, axial fluid-attenuated inversion recovery (FLAIR), axial spoiled gradient echo (FSPGR), DWI and axial T1-weighted imaging, using 0.1 mmol/kg of body weight of gadopentetate dimeglumine (Magnevist; Schering, Berlin, Germany). DTI was performed using a single-shot SE EPI sequence. Diffusion gradients were applied in 25 directions with b-values of 700 s/mm² and an image without diffusion weighting with b-value of 0 s/mm². DTI sequences were acquired in the axial plane with 44 contiguous sections, 2.4-mm section thickness, no intersection gap, TR/TE of 17,000/80 ms, with imaging in parallel for decreasing the time of acquisition, 25×25 cm FOV and 128×128 matrix size.

MR was performed on a single occasion by using a 3-T unit (Signa HDxt, GE Healthcare, Waukesha, WI, USA) and a high-resolution eight-channel head coil (Invivo, Gainesville, FL, USA).

Image postprocessing and data analysis (DTI tensor-metric formulas) [6-10]

$$\begin{split} MD &= D = \frac{\lambda_1 + \lambda_2 + \lambda_3}{3} & p = \sqrt{3D} = \frac{\lambda_1 + \lambda_2 + \lambda_3}{\sqrt{3}} \quad AD = \lambda_1 \\ L &= \sqrt{p^2 + q^2} = \sqrt{\lambda_1^2 + \lambda_2^2 + \lambda_3^2} & q = \sqrt{(\lambda_1 - D)^2 + (\lambda_2 - D)^2 + (\lambda_3 - D)^2} \quad C_l = \frac{\lambda_1 - \lambda_2}{\lambda_1 + \lambda_2 + \lambda_3} \\ RA &= \frac{q}{p} = \frac{\sqrt{(\lambda_1 - D)^2 + (\lambda_2 - D)^2 + (\lambda_3 - D)^2}}{\sqrt{3D}} & FA = \sqrt{\frac{3}{2}} \frac{q}{L} = \sqrt{\frac{3}{2}} \sqrt{\frac{(\lambda_1 - D)^2 + (\lambda_2 - D)^2 + (\lambda_3 - D)^2}{\lambda_1^2 + \lambda_2^2 + \lambda_3^2}} \quad C_p = \frac{2(\lambda_2 - \lambda_3)}{\lambda_1 + \lambda_2 + \lambda_3} \\ C_s &= \frac{3\lambda_3}{\lambda_1 + \lambda_2 + \lambda_3} & RD = \frac{\lambda_2 + \lambda_3}{2} \end{split}$$

We used the software dcm2nii [12] and the FMRIB Software Library (FSL) v. 4.1.9 [13]; for extraction of DTI images, we used the *Brain Extraction Tool (BET)* v. 2.1 [14]. Eddy currents were corrected using the *FMRIB's Diffusion Toolbox* v. 2.0; the *Reconstruct Diffusion Tensor (DTIFIT)* and the *fslmaths tool* generated the eigenvector and eigenvalue maps for each tensor metric. The *fslstats tool* calculated the scalar measures (mean values) of the selected regions.

For each tumour region, three regions of interest (ROI) with a maximum average size of 10 mm² were drawn and manually segmented using DTIFIT at three different locations, separated at least 10 mm from the adjacent one: cystic cavity, enhancing rim, an immediate zone of oedema (oedema most adjacent to the enhancing rim, it was arbitrarily chosen as a 10-mm-wide band) and a distant zone of oedema (oedema adjacent to the immediate zone of oedema, chosen also as a 10-mm-wide band). Each enhancing-rim zone was matched with three measurements of normal-appearing white matter (NAWM): one drawn in the patient's contralateral hemisphere and two additional ones drawn in the ipsilateral hemisphere of two healthy volunteers. A total of 15 ROIs were obtained from each patient with four additional ROIs of NAWM matched from normal subjects. An example of the sequences and regions of interest is shown in Fig. 1a-d.

Statistical analysis

Sample size had two considerations: First, our research was defined as phase I, "exploratory phase" [15], in which a new diagnostic test is first evaluated in a clinical setting to determine whether the test has any ability to discriminate diseased from nondiseased patients. Second, we matched this phase I with the summarised list of computed sample sizes needed for an exploratory retrospective study as

reported by Obuchowski et al. [16]. At least ten diseased patients and ten control patients are required to keep statistical validation in a receiver-operating characteristic (ROC) curve analysis assuming the type I error rate is set at 0.05 and type II error rate is ≤ 0.10 , with power ≥ 0.90 . Our study included 14 patients and 28 controls.

Evaluations of diagnostic tests used one-*way analysis of variance* (ANOVA) [17] and the Student-Newman-Keuls test [18–20] for pairwise comparison of subgroups. The *effect size* was obtained for each ANOVA result using the Eta squared value [21]. It considered 0.01 as a small effect, 0.06 as a medium effect and 0.14 as a large effect [22].

The area under the ROC curve (AUROC) was considered the measure of the overall performance of each tensor metric [23–25]; standard error (SE), p value and its 95 % confidence interval (CI) were also reported. The method of Delong [26] was used for the calculation of SE and pairwise comparisons. Values of sensitivity, specificity, (+) and (–) predictive value, and (+) and (–) likelihood ratios were also displayed. Cutoff ranges represented the optimal cutoff to maximise sensitivity plus specificity for assessment of the presence of the five selected brain regions. The accuracy was determined using a traditional academic point-system classification: 0.90-1=excellent (A), 0.80-0.90=good (B), 0.70-0.80=fair (C), 0.60-0.70=poor (D) and 0.50-0.60=fail (F) [27].

Software All analyses were carried out using MedCalc[®] (version 12.2.1 MedCalc Software bvba, Mariakerke, Belgium). Statistical significance was indicated by a *P*-value<0.05.

Presentation of data followed the Standards for Reporting of Diagnostic Accuracy (STARD) initiative [28]. General presentation of the manuscript was prepared in accordance with the guidelines set by the International Committee of Medical Journal Editors [29]. **Fig. 1 a, b** Conventional MR sequences used in the MRI evaluation of brain tumours showing axial plane T1weighted post-contrast and T2weighted images. **c** T2weighted image in the axial plane generated by the FSL showing the drawn ROIs corresponding to each tumour region. **d** Example of ROI location on FA maps



Results

Demographic and quantitative DTI tensor maps

The study was conducted in 14 patients: 6 men (mean age, SD; 51.50 ± 17.74 years; range, 33-78 years) and 8 women (mean age, SD; 49.75 ± 16.54 years; range, 31-73 years). There were 28 controls: 9 men (mean age, SD; 45.78 ± 21.57 years; range, 24-72 years) and 19 women (mean age, SD; 27.21 ± 10.25 years; range, 20-68 years). Ninety-eight ROIs of the NAWM region (from all tensors) and 40 ROIs from each tumour region added 258 measurements included in the analysis.

Anova analyses and pairwise comparison of subgroups

There was a statistically significant difference with a large effect size for all tensor metrics: MD=0.418; FA=0.532; p= 0.515; q=0.238; L=0.461; Cl=0.554; Cp=0.366; Cs= 0.627; RA=0.616; RD=0.576 and AD=0.345. Post-hoc tests also showed significant pairwise comparisons within

subgroups of all tensor metrics. Table 1 presents the means and SD within the subgroups of each tensor metric.

Diagnostic performance of tensor metrics

Among the 55 ROC curves calculated for each patient, the highest performer for each region corresponded to: RA in the NAWM and cystic cavity with an excellent fit (0.944 and 0.829 respectively); RD in the immediate zone of oedema with fair performance (0.708); Cp was the highest for the distant zone of oedema with a fair performance (0.785); q depicted a poor fit (0.642) in the enhancing rim zone. Tables 2 and 3 and Figs. 2 and 3 show the values of the AUROC analyses and the interactive dot diagrams corresponding to the three best performers for each tumour region.

Pairwise comparison of ROC curves

Significant differences of the tensor metrics in each region were observed only between: RA with Cp, p, MD, L and q in the *NAWM*; RD with AD in the *immediate zone of*

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Table 1 Means and SD within subgroups for each tensor metric

Tensor metric	NAWM		Immediate zo:	ne of oedema	Distant zone	of oedema	Enhancing	rim	Cystic cavi	ty
	Mean	SD	Mean	SD	Mean	SD	Mean	SD	Mean	SD
MD	0.000832	0.000147	0.001519	0.000420	0.001608	0.000253	0.001138	0.000332	0.001951	0.001093
FA	0.513454	0.143641	0.195618	0.122330	0.177375	0.081779	0.241313	0.100363	0.202516	0.215315
р	0.001440	0.000255	0.002630	0.000726	0.002783	0.000438	0.001970	0.000574	0.003698	0.001693
q	0.000666	0.000202	0.000376	0.000150	0.000394	0.000156	0.000384	0.000192	0.000422	0.000483
L	0.001609	0.000240	0.002669	0.000703	0.002819	0.000428	0.002019	0.000581	0.003677	0.001755
C1	0.267629	0.114121	0.082389	0.065540	0.081726	0.047998	0.096443	0.051993	0.058288	0.051932
Ср	0.228388	0.110703	0.099713	0.064011	0.070233	0.032118	0.138498	0.062885	0.080032	0.103336
Cs	0.520621	0.130280	0.819050	0.116676	0.851493	0.061990	0.770997	0.093871	0.861679	0.141119
RA	0.480533	0.155383	0.166486	0.114198	0.147988	0.070564	0.205776	0.091396	0.124237	0.117666
RD	0.000575	0.000190	0.001380	0.000441	0.001456	0.000271	0.000999	0.000313	0.001977	0.000893
AD	0.001352	0.000192	0.001799	0.000402	0.001911	0.000268	0.001417	0.000410	0.002452	0.001226

oedema; Cp with Cs, FA, RA, Cl and q at the *distant zone of oedema*; q with FA, RA, Cs and RD in the *enhancing rim zone*; RA with q in the *cystic cavity zone*.

Discussion

Diagnostic accuracy studies are a vital step in the evaluation of new diagnostic technologies [30]. They compared results from one or more tests with the results obtained with the reference standard evaluated with the purpose of detecting the target condition [30]. Our findings allowed us to identify which DTI-derived tensor metrics better classified each selected tumour region in a group of patients with GBM and healthy volunteers.

The clinical relevance of our study has several components: first, we report a comprehensive diagnostic evaluation for each selected region; the top three tensor metrics presented in each region (largest AUROC) did not show statistically significant differences in the pairwise comparisons, meaning they could be used interchangeably. Our study has differences from the one published by Toh et al. [7]: we only compared GBM with a control group of healthy volunteers. The NAWM included measurements of matched ROIs (to the enhancing rim area) from the contralateral hemisphere of patients and the ipsilateral hemisphere of

Table 2 AUROC analyses of the three best biomarkers at each tumour region

Tumour region	Best DTI-tensor performers	Cutoff value	AUROC	SE	P-value	95 % CI	
						Lower bound	Upper bound
NAWM	RA	>0.3079	0.944	0.0146	< 0.0001	0.908	0.969
	Cs	≤0.6286	0.944	0.0153	< 0.0001	0.908	0.969
	Cl	>0.1518	0.933	0.0153	< 0.0001	0.895	0.96
Immediate zone of oedema	RD	>0.0009	0.708	0.0387	< 0.0001	0.648	0.764
	FA	≤0.2595	0.700	AUROC SE P -value 95 % CI Lower bound Uppe 0.944 0.0146 <0.0001	0.756		
	MD	>0.0011	0.695	0.0439	< 0.0001	0.635	0.751
Distant zone of oedema	Ср	≤0.0929	0.785	0.0304	< 0.0001	0.73	0.835
	MD	>0.0012	0.769	0.028	< 0.0001	95 % CI Lower bound 0.908 0.895 0.648 0.640 0.635 0.73 0.713 0.713 0.708 0.580 0.551 0.550 0.777 0.776 0.768	0.819
	р	>0.0021	0.765	0.0284	< 0.0001	0.708	0.816
Enhancing rim	q	≤ 0.0005	0.642	0.0434	0.0011	0.580	0.701
	AD	≤ 0.002	0.614	0.0482	0.0177	0.551	0.675
	Cl	≤0.1865	0.613	0.0384	0.0033	0.550	0.673
Cystic cavity	RA	≤0.1623	0.829	0.0385	< 0.0001	0.777	0.874
	Cs	>0.8392	0.828	0.0404	< 0.0001	0.776	0.873
	Cl	≤0.0711	0.821	0.0363	< 0.0001	0.768	0.866

Tumour region	Best DTI-Tensor performers	Sensitivity		Specifici	ty	+LR		-LR		V^{+}		۸d-	
		Value	95 % CI	Value	95 % CI	Value	95 % CI	Value	95 % CI	Value	95 % CI	Value	95 % CI
NAWM	RA	86.60	78.2–92.7	90.97	85.3-95.0	9.59	8.7-10.5	0.15	0.07 - 0.3	85.7	77.2–92.0	91.6	86.0-95.4
	Cs	84.54	75.8-91.1	96.13	91.8-98.6	21.84	19.9–23.9	0.16	0.06 - 0.4	93.2	85.7–97.5	90.9	85.4-94.8
	CI	81.44	72.3-88.6	89.03	83.0-93.5	7.43	6.7 - 8.3	0.21	0.1 - 0.4	82.3	73.2-89.3	88.5	82.4-93.0
Immediate zone of oedema	RD	84.62	69.5-94.1	56.07	49.1-62.8	1.93	1.6 - 2.3	0.27	0.1 - 0.6	26.0	18.6 - 34.5	95.2	89.9–98.2
	FA	84.62	69.5-94.1	59.63	52.8-66.2	2.10	1.8 - 2.5	0.26	0.1 - 0.5	27.3	19.6 - 36.1	95.6	90.6 - 98.4
	MD	87.18	72.6-95.7	57.80	50.9-64.4	2.07	1.8 - 2.4	0.22	0.10 - 0.5	27.0	19.5-35.6	96.2	91.3-98.7
Distant zone of oedema	Cp	88.89	73.9–96.9	68.52	61.9–74.7	2.82	2.4-3.3	0.16	0.06 - 0.4	32.0	23.0-42.1	97.4	93.4-99.3
	MD	100.0	90.3 - 100	64.25	57.6-70.6	2.8	2.5 - 3.1	0.00	0.0 - 0.0	31.3	22.9-40.7	100.0	97.4-100.0
	Ь	100.0	90.3 - 100	63.59	56.8-70.0	2.73	2.5 - 3.0	0.0	0.0 - 0.0	31.3	23.0-40.6	100.0	93.1-100.0
Enhancing rim	Ь	73.81	58.0 - 86.1	49.77	42.9–56.6	1.47	1.2 - 1.8	0.53	0.3 - 0.9	22.3	17.7 - 30.1	90.7	83.9–95.3
	AD	97.62	87.4–99.9	28.91	22.9–35.5	1.37	1.1 - 1.7	0.082	0.01 - 0.6	21.5	15.9 - 28.0	98.4	91.2-100.0
	CI	97.62	87.4–99.9	37.14	30.6 - 44.1	1.55	1.3 - 1.9	0.064	0.009 - 0.4	23.7	17.6 - 30.7	98.7	93.1–100.
Cystic cavity	RA	84.21	68.7–94.0	71.50	64.9–77.4	2.95	2.5-3.5	0.22	0.1 - 0.5	34.4	24.9-45.0	96.2	92.0–98.6
	Cs	81.58	65.7–92.3	73.36	66.9–79.2	3.06	2.6 - 3.6	0.25	0.1 - 0.5	35.2	25.3-46.1	95.7	91.4–98.3
	CI	81.58	65.7–92.3	72.43	65.9–78.3	2.96	2.5-3.5	0.25	0.1 - 0.5	34.4	24.7-45.2	95.7	91.3–98.2

Fable 3 Evaluation of diagnostic tests for the three best biomarkers at each tumour region

healthy volunteers; we followed the basic requisites of a diagnostic test to obtain validity in its results [25].

Second, each diagnostic evaluation followed the STARD initiative [28], reporting AUROC, CI, SE and p-values as well as the sensitivity and specificity, (+) and (-) predictive values and (+) and (-) likelihood ratios with their corresponding CI. To the best of our knowledge, although just a few studies have followed this initiative [7, 8], comprehensive evaluations like the one presented in this study are scarce in imaging reports.

Third, we obtained evidence that FA and MD are not the best tensor metrics in any of the five selected regions for this study. These results differ from those presented by Toh et al. [7]. They found that FA was the best biomarker for the cystic cavity and enhancing-rim zones and ADC for the immediate oedema zone. Our results are not necessarily contradictory but might complement the current available information: our pairwise analysis of ROC curve comparisons showed that the top sorted biomarkers for each region could be clinically interchangeable. Our results provide additional hints to clarify tensor metrics' abilities and might unveil underlying behaviours of these biomarkers. An integrative approach could explain why some researchers found significant relationships while others did not.

Fourth, there is still a significant amount of empiricism surrounding the understanding of tensor metrics in each tumour region and their expected cutoff values. In the MR-DTI literature, usually FA and MD are displayed as the accepted biomarkers of brain microstructure; however, the caveat to exclusively using MD and FA to characterise pathology in clinical applications is that it is still not known a priori which tensor measure is the most appropriate to quantify pathological changes in brain tissue [6]. Eventually, there will be a deeper understanding as more studies are reported, especially those including multivariate analysis of biological features presented in patients with brain tumours (gender, age, regional location of tumour, race, etc.).

A grading of the peritumoral oedema in MR images is relevant, considering its role as an independent prognostic factor has been proved. We supported our classification of immediate and distant oedema with the study by Schoenegger et al. [31], who found that when oedema is classified as minor (<1 cm) and major (>1 cm), patients with the presence of major oedema had significantly shorter overall survival compared to patients with minor oedema; also, distant oedema has been correlated with a higher degree of necrosis and vascular endothelial growth factor (VEGF) expression [11]. The higher performance of FA observed in the immediate zone of oedema could explain its mechanism of decrement related to an increase in extracellular space Fig. 2 a, c and e Images correspond to the interactive dot diagrams. b, d, and f ROC curves from to the NAWM, immediate and distant zones of oedema respectively



secondary to neuronal and fibre tract destruction [32] or to a reduction in extracellular space secondary to tumour infiltration [33]. The best performance of MD in both oedema zones is explained by its measuring the magnitude of the molecular motion of water, although it does not depend directly on the integrity of myelinated fibre tracts [9].

The best biomarker for the enhancing rim zone was q, followed by AD and Cl. Previous studies on the features of FA have mentioned that the inclusion of the parameter q may be able to provide a more complete picture of the diffusion profile of a brain tumour

[6]. Price et al. [34] found that in brain tumours, compared with normal white matter, there is a less markedly decreased q (mean 14 %) in infiltrated white matter, with marked reduction in q (mean 42 %) in disrupted white matter. AD describes microscopic water movement parallel to axonal tracts and has been associated with acute axonal injury [35]. Cl had the worst performance among FA, MD, Cs and Cp in the study by Toh et al. [7]. We hypothesise that the geometric component of linearity in Cl accounts for its better discriminant ability for the enhancing rim zone compared with other tensor metrics. Fig. 3 a, c Images correspond to the interactive dot diagrams. b and d ROC curves from to the enhancing rim and cystic cavity respectively



For the cystic cavity and the NAWM, RA was the best biomarker followed by Cs and Cl; seven metrics outperformed FA and MD in the former zone and four tensor metrics did in the latter zone. RA is a ratio of the normalised standard deviations between the anisotropic part of the diffusion coefficient and its isotropic part [9]. It is a function of the variance of the eigenvalues of the diffusion tensor, which is not equal to the variance of the diffusivities along all directions [36]. There have not been publications about the clinical applicability of RA in brain tumours; RA has been found to have inferior noise immunity (lower signal-tonoise ratio) than FA for any tensor anisotropy value [37]. We believe the features of the RA definition and the spherical geometric properties of the Cs tensor allowed them to better characterise the cystic cavity and NAWM. The performance of Cs (second position) after RA agrees with the results for Cs published by Toh et al. [7].

Some limitations in this study need to be addressed: DTI's acceptance by the medical community is still limited, mainly because it does not yet reach the anatomic validity of the myeloarchitectonic studies and currently cannot differentiate individual axons or synaptic connections [38]. We consider this project a phase-1 study; it was not our purpose to either correlate clinical features of our subjects or to explore prognostic abilities; some readers would like to know which combinations of metrics increase the sensitivity in a specific region. A binary logistic regression might unveil this answer, but it would require additional analyses and explanations. Thus, the clinical value of our findings has yet to be determined, and the biological impact of the different metrics should be explained in more detail in clinical journals.

There is still scarce information regarding the potential of tensor metrics as biomarkers. This is prehaps because several publications have shown contradictory results for some biomarkers such as FA and MD [8, 39-41]. Our control and patient groups had differences between gender and age distributions; this could influence our measurements, as has been proved for FA and AD with aging. Also, variability due to the size of the selected ROIs should be considered; some papers reported ROIs encompassing the whole enhancing rim or oedema that varied at each subject; in a previous publication from our group [41], we proved the existence of significant FA variability across normal pyramidal tracts, which is influenced by anatomic landmarks and the cerebral hemisphere. Small ROIs (no larger than 10 mm in diameter) might provide less variability by encompassing a smaller number of axonal tracts (each one hypothetically with significantly different DTI values) and explain the lack of agreement among previous reports. A complex

mathematical explanation of the differences in AUROCs for each selected region is beyond the scope of this manuscript, which was written for clinical researchers. Further studies should address if the theoretical definitions of each tensor metric can explain the intricacies of brain tumour behaviour affected by the ratio of extracellular to intracellular space, vascularity, oedema, microcysts and extracellular matrix [42]. This study is part of an on-going project trying to understand the role of tensor metrics in characterising GBM. We are aware of the omission in this study of metabolic data provided by MRI using spectroscopy and perfusion; however, we limited our definitions of regional tumour infiltration and oedema boundaries to those patterns observed in the T1-weighted post-contrast and T2-weighted images, as they have been used as the reference for diagnosis in previous reports [11, 31]. Additional studies including metabolic sequences might help to understand which approach offers the best performance.

Our study makes a contribution toward understanding the fundamentals of the tensors' relationships; this information may help radiologists and/or bio-imaging experts to explain their clinical applicability to clinicians (neurologists, neurosurgeons, oncologists, psychiatrist, paediatricians, etc.). Almost 20 years after the initial definitions of FA and MD [43], there is incomplete understanding of the relationships between brain structure and tensor metrics; they might answer different questions. Also, variations in DTI measures are not specific to one histological type of tumour, which broadens the application of these biomarkers to a wide variety of intracranial lesions. Nowadays, DTI-derived biomarkers can characterise the regional impairment of intraaxial brain tumours; in the short term we expected that an advanced quantitative analysis of DTI-derived tensor metrics might be accepted in clinical practice, depending on the interest of the clinicians, for screening of patients, triaging patients presenting with symptoms suggestive of progression and/or follow-up to evaluate treatment response (radiotherapy).

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research article

Global diffusion tensor imaging derived metrics differentiate glioblastoma multiforme vs. normal brains by using discriminant analysis: introduction of a novel whole-brain approach

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Background. Histological behavior of glioblastoma multiforme suggests it would benefit more from a global rather than regional evaluation. A global (whole-brain) calculation of diffusion tensor imaging (DTI) derived tensor metrics offers a valid method to detect the integrity of white matter structures without missing infiltrated brain areas not seen in conventional sequences. In this study we calculated a predictive model of brain infiltration in patients with glioblastoma using global tensor metrics.

Methods. Retrospective, case and control study; 11 global DTI-derived tensor metrics were calculated in 27 patients with glioblastoma multiforme and 34 controls: mean diffusivity, fractional anisotropy, pure isotropic diffusion, pure anisotropic diffusion, the total magnitude of the diffusion tensor, linear tensor, planar tensor, spherical tensor, relative anisotropy, axial diffusivity and radial diffusivity. The multivariate discriminant analysis of these variables (including age) with a diagnostic test evaluation was performed.

Results. The simultaneous analysis of 732 measures from 12 continuous variables in 61 subjects revealed one discriminant model that significantly differentiated normal brains and brains with glioblastoma: Wilks' $\lambda = 0.324$, χ^2 (3) = 38.907, p < .001. The overall predictive accuracy was 92.7%.

Conclusions. We present a phase II study introducing a novel global approach using DTI-derived biomarkers of brain impairment. The final predictive model selected only three metrics: axial diffusivity, spherical tensor and linear tensor. These metrics might be clinically applied for diagnosis, follow-up, and the study of other neurological diseases.

Key words: brain neoplasms; diffusion tensor imaging; discriminant analysis; magnetic resonance imaging; predictive value of tests

Introduction

Some pathologic and magnetic resonance (MR) imaging characteristics of astrocytomas grades II to IV (highest degree known as glioblastoma multiforme, GBM) suggest these tumors would benefit from the use of a global measurement of brain impairment.¹ The first imaging approaches to characterize high-grade glial lesions, especially GBM, were fraught with pitfalls resulting from the marked heterogeneity of both glial-infiltrated and normal brains.^{2,3} These tumors frequently contain multiple areas of variable histologic features, so that a sampling error in a biopsy may mean that the degree of

malignancy seen by the neuropathologist may not reflect the degree of malignancy present elsewhere in the tumor, resulting in significant undergrading of some lesions.² Thus, even when all radiologically visible portions of a tumor have been excised, the surgical margins may not be "clean", and further neoplastic growth can (and usually does) occur in the adjacent brain tissue, leading from microscopic residual to gross recurrence.⁴ Therefore, none of the MR protocols for GBM in every day practice should be only morphologic.3,5,6 As a consequence, surgery usually only reduces the tumor; this information is relevant as recent evidence has proved gross total resection (surgical margin status) significantly correlates with progression-free, recurrence pattern and overall survival in patients with GBM.7,8

GBM is considered a whole brain disease. Radiotherapy and chemotherapy follow surgery.

Although MR perfusion and spectroscopy⁹, and sometimes diffusion tensor imaging (DTI)¹⁰ are routinely used methods to locate parts of the tumor-GBM with high malignancy that should be biopsied, the development of specific and sensitive biomarkers remains a critical unmet need.¹¹

Our purpose in this study was to explore the diagnostic ability of a global (whole brain) assessment of DTI-derived tensor metrics in normal and infiltrated brains with GBM. We used the multivariate technique of linear discriminant analysis (DA), previously reported in MRI diagnosis¹², to classify the study participants into groups, describe group differences and to assess the relative importance of DTI variables for discriminating between groups. This analysis might unveil findings and associations that cannot, in a partial-regional assessment, be recognized at surgery, neurologic, MRI and/or pathologic examination. Considering there is still scarce information in the medical literature about the global calculation of tensor metrics^{13,14}, a predictive discriminating model may offer an innovative diagnostic approach to the surgical-neurooncology team.

Subjects and methods

Subjects

This was a case-control study. We included patients with suspected diagnosis and later pathological confirmation of primary GBM who had undergone preoperative brain MR examinations between January 2010 and September 2012. Exclusion criteria were corticosteroid or antibiotic treatment, lesions with areas related to calcification and/or hemorrhage and previous brain surgery. A control group included young and elderly healthy volunteers recruited from the enrolled interns and medical residents of the hospital, and elderly subjects from our Geriatrics unit. All volunteers received detailed health examinations; exclusion criteria were major neurological, psychiatric, or cardiovascular diseases. A radiologist interpreted the MR images blinded to the patient's history and MRI examinations with other structural abnormalities were excluded. The local institutional review board approved the study (Project #2011-EXT-05).

Brain image acquisition

MR sequences included conventional axial T2-weighted imaging, axial Fluid-Attenuated Inversion Recovery (FLAIR), axial Spoiled Gradient Echo (SPGR), DWI and axial T1-weighted imaging, using 0.1 mmol/kg of body weight of gadopentetate dimeglumine (Magnevist; Schering, Berlin, Germany). Healthy volunteers did not receive endogenous contrast. DTI was performed using a single-shot SE EPI sequence. Diffusion gradients were applied in 25 directions with b-values of 1000 s/mm² and an image without diffusion weighting with b-value of 0 s/mm². DTI sequences were acquired in the axial plane with 44 contiguous sections, 2.4 mm section thickness, no intersection gap; TR/TE of 17,000/80 ms, with parallel imaging to reduce off-resonance artifacts (PI factor was 2); 25 x 25 cm FOV; and 128 x 128 matrix/pixel size. MR was performed on a single occasion using a 3T unit (Signa HDxt, GE Healthcare, Waukesha, WI, USA); and a high-resolution eight-channel head coil (Invivo, Gainesville, FL, USA).

Image postprocessing and data analysis

We used the software dcm2nii¹⁵ and the FMRIB Software Library (FSL) v. 4.1.9.¹⁶ DTI images were extracted using the *Brain Extraction Tool (BET)* v. 2.1.¹⁷ Eddy currents were corrected using the *FMRIB's Diffusion Toolbox v. 2.0;* the *Reconstruct Diffusion Tensor (DTIFIT)* and the *fslmaths tool* generated the eigenvector and eigenvalue maps for each tensor metric. The *fslstats tool* calculated the scalar measures (mean values) of each whole-brain calculation. The apparent diffusion coefficient (ADC) value, a simple index calculated from diffusionweighted images¹⁸, was considered equivalent to the MD (mean diffusivity) metric, as it was obtained from the DTI sequence.¹⁹ DTI-derived tensor

Tensor	metric											
		Cs										
FA	Pearson´s R	552	FA									
	p-value	< .001										
RA	Pearson´s R	-1.000	.557	RA								
	p-value	< .001	< .001									
Ср	Pearson's R	937	.584	.912	Ср							
•	p-value	< .001	< .001	< .001	·							
CI	Pearson´s R	898	.541	.943	.673	CI						
	p-value	< .001	< .001	< .001	< .001							
L	Pearson´s R	.211	.075	175	191	< .001	L					
	p-value	.165	.596	.256	.213	.999						
p	Pearson´s R	.195	079	183	194	.006	.890	р				
-	p-value	.205	.580	.240	.214	.972	< .001	-				
AD	Pearson´s R	.034	.209	008	002	.106	.882	.880	AD			
	p-value	.826	.137	.958	.989	.508	< .001	< .001				
MD	Pearson´s R	.195	078	183	193	.007	.892	1.000	.881	MD		
	p-value	.206	.589	.241	.214	.968	< .001	< .001	< .001			
RD	Pearson´s R	.213	226	209	149	105	.815	.973	.779	.973	RD	
	p-value	.151	.103	.163	.316	.502	< .001	< .001	< .001	< .001		
q	Pearson´s R	306	.804	.310	.339	.328	.403	.281	.627	.284	.214	q
-	p-value	.031	< .001	.030	.016	.026	.003	.046	< .001	.044	.116	-

TABLE 1. Correlations of tensor metrics, controlled for the effect of diagnosis, age and gender

AD = axial diffusivity; CI = linear tensor; Cp = planar tensor; Cs = spherical tensor; FA = fractional anisotropy; L = the total magnitude of the diffusion tensor; MD = mean diffusivity; p = pure isotropic diffusion; q = pure anisotropic diffusion; RA = relative anisotropy; RD = radial diffusivity

metrics formulas using the major (λ 1), intermediate (λ 2), and minor (λ 3) eigenvalues allowed the calculation of the eleven most common tensor metrics for brain imaging: mean diffusivity (MD), fractional anisotropy (FA), pure isotropic diffusion (p), pure anisotropic diffusion (q), the total magnitude of the diffusion tensor (L), linear tensor (Cl), planar tensor (Cp), spherical tensor (Cs), relative anisotropy (RA), axial diffusivity (AD) and radial diffusivity (RD)¹⁰; each one representing a single global measure of the whole-brain. Figure 1A shows the algorithm for measuring the DTI-derived tensor metrics.

Statistical analysis

Study design

The study was considered a Phase II aimed to determine the capacity of DTI-derived biomarkers to distinguish between people with cancer and those without.²⁰

Sample size

Considering our predictive model to discriminate between normal brains *vs*. brains infiltrated with GBM underwent a diagnostic performance assessment, the adequacy of the sample size to expect validity from our results was based on matching this phase with the summarized list of computed sample sizes needed for an exploratory retrospective study reported by Obuchowski *et al.*²¹, at least 10 diseased patients and 10 control patients were required to maintain statistical validation in a diagnostic test evaluation where the type I error rate was set at 0.05, type II error rate was \leq 0.10, and power \geq 0.90. Our study included 27 patients and 34 controls. 3



FIGURE 1. (A), FSL software algorithm used in the image postprocessing and data analyses. (B-E), Examples of acquired sequences in a patient with GBM and the tensor-metric maps generated for the data analyses: (B), axial T2-weighted; (C), post contrast axial T1-weighted; (D), axial diffusivity (AD) tensor map; and (E), fractional anisotropy (FA) tensor map. Notice how it might not be possible to perform an imaging diagnosis based only on a visual inspection of these maps.

Multivariate DA

We ran a DA, which was optimal under the same conditions where Manova was optimal; then attempted to detect any deviation from Manova assumptions that might distort the tests of statistical significance.²² We assessed the normality of the distribution of the DTI-derived scores using the Kolmogorov-Smirnov's and Shapiro-Wilk normality tests²³; eliminated significant outliers, evaluated multivariate normality and linearity, and tested the homogeneity of variance-covariance matrices using the Box's M test.²⁴ Considering the similarity of the tensor-metric formulae, we ran scatterplots and correlations to check the strength of correlations among the dependent variables in order to detect the presence of multicollinearity and singularity (Table 1). Partial correlation analyses were carried out to calculate the Pearson's correlation coefficient (r) controlling for the effect of age, gender and clinical diagnosis. The strength of the linear relationship corresponding to each correlation coefficient value was interpreted as very strong (at least of 0.8), moderately strong (0.6 up to 0.8), fair (0.3 up to 0.6) and poor (less than 0.3). A squared r value represented the coefficient of determination, the proportion of variance that each two compared variables had in common.25

We applied the stepwise method in DA, it considered the value of Wilk's lambda and changing criteria: minimum partial F to enter of 3.84 and minimum partial F to remove of 2.71.22 Continuous variables were included with the predictive aim to identify specific tensor-metric attributes in GBM and normal brains. The dependent variable (DV) used in the DA was the clinical diagnosis, which classified subjects as patients or controls. The independent variables (IVs) included 11 DTI-derived tensor metrics: MD, FA, p, q, L, Cl, Cp, Cs, RA, RD and AD, and the patients' age (in years). The effectsize measure for discriminant analysis was calculated using the squared canonical correlation as the equivalent of the R² in regression.²⁶ By convention, effect sizes of 0.02, 0.15, and 0.35 are termed small, medium, and large, respectively.²⁷ For all analyses, statistical significance was indicated by a *p*-value < 0.05.

Diagnostic model evaluation

The cross-validated contingency Table generated by the DA was used to evaluate the diagnostic performance of the DA model. We reported values of sensitivity, specificity, positive and negative likelihood ratios, and positive and negative predictive values, with their corresponding confidence in-

Veriable	Healthy	/ brains	Brains w	ith GBM	Wilks'	T to at	
Valiable	Mean	SD	Mean	SD	Lambda	r iesi	p-value
Cs (spherical tensor)	.747091	.026938	.768395	.042299	.915	3.341	.076
FA (fractional anisotropy)	.287029	.011517	.254082	.026761	.607	23.341	< .001
RA (relative anisotropy)	.233436	.025491	.209370	.033559	.855	6.088	.018
Cp (planar tensor)	.138366	.013823	.136635	.036218	.999	.036	.850
CI (linear tensor)	.114543	.013991	.098463	.011930	.711	14.621	.001
L (total magnitude of the diffusion tensor)	.002277	.000087	.002117	.000147	.691	16.077	< .001
p (pure isotropic diffusion)	.002107	.000077	.001959	.000134	.681	16.893	< .001
AD (axial diffusivity)	.001548	.000044	.001399	.000087	.461	42.052	< .001
MD (mean diffusivity)	.001217	.000044	.001132	.000078	.685	16.526	< .001
RD (radial diffusivity)	.001051	.000050	.000997	.000078	.852	6.237	.017
q (pure anisotropic diffusion)	.000452	.000036	.000367	.000047	.483	38.529	< .001
Age	40.333	21.502	47.150	15.187	.965	1.294	.263

TABLE 2. Multivariate analysis (between-groups) of diffusion tensor imaging (DTI)-derived tensor metrics and age showing the statistical differences between means of normal-brain and brain-with- as glioblastoma multiforme (GBM) groups for the independent variables included in the analysis

SD = standard deviation

tervals (CI). Evaluation of the diagnostic tests followed the Standards for Reporting of Diagnostic Accuracy (STARD) initiative.²⁸

Software

All analyses were carried out using the IBM[®] SPSS[®] Statistics software (version 22.0.0.0 IBM Corporation; Armonk, NY, USA). Diagnostic performance was assessed using MedCalc[®] (version 12.3.0.0 MedCalc Software bvba, Mariakerke, Belgium).

Results

Subjects and MRI acquisition

The study was conducted in 61 subjects; 27 patients: 13 females (mean age 50.0 ± 15.400 years, range 31–73 years) and 14 males (mean age $46.93 \pm$ 15.403 years, range 18–78 years); and 34 controls: 26 females (mean age 41.04 ± 22.37 years, range 21–80 years) and 8 males (mean age 42.88 ± 21.89 years, range 24–72 years). The eleven DTI tensor-maps plus the age (per subject) added up 732 measurements included in the analyses. Figure 1 B-E shows examples of some of the MR sequences and tensormetric maps used in the data analyses.

Partial correlation analyses

A scatterplot showed no serious violation of the assumptions of linearity, homoscedasticity, and

outliers. Among 55 pairs of bivariate correlations, we found only 15 with a significantly very strong (at least 0.8) r value: Cs \Leftrightarrow RA (-), Cs \Leftrightarrow Cp (-), Cs \Leftrightarrow L (-), FA \Leftrightarrow q (+), RA \Leftrightarrow Cp (+), RA \Leftrightarrow Cl (+), L \Leftrightarrow p (+), L \Leftrightarrow AD (+), L \Leftrightarrow MD (+), L \Leftrightarrow MD (+), p \Leftrightarrow AD (+), p \Leftrightarrow MD (+), p \Leftrightarrow MD (+), AD \Leftrightarrow MD (+), and MD \Leftrightarrow RD (+). Table 1 and Figure 2 depict correlation values and the scatterplot of the eleven tensor-metrics.

Discriminant analysis

Although some r values were calculated at > 0.8, we included all variables in the DA, as we found evidence the stepwise variant of this method protects against multicollinearity and singularity;²⁹ a brief explanation is presented in the discussion section. The assumption of homogeneity of variance-covariance matrices was interpreted as significant (Box's M value = 35.110, F = 5.317, df (6, 9087.738), p = < .001). In the stepwise statistics, at each step, the best variable that minimized the overall Wilks' Lambda was entered: AD was entered at Step 1, F = 42.052 (1, 36) p < .001; Cl entered at Step 2, F = 29.609 (2, 35) p < .001; and Cs entered at Step 3, F = 23.672 (3, 34) p < .001.

DA revealed one discriminant function that significantly differentiated the normal brains and GBM brains: Wilks' λ = 0.324, χ 2 (3) = 38.907, p < .001. By indicating the significance of the discriminant function, Wilks' lambda provided a moderate proportion of total variability not explained by the model of 10.49%. A canonical correlation of 0.822



6

FIGURE 2. Scatter matrix of the data variables grouped by diagnosis.

suggested the model explains 67.56% of the variation in the grouping variable.

Summary of discriminant functions

The tests of equality of group means provided statistical evidence of significant differences between means of normal brains and brains with GBM in 9 of the IVs, with AD producing the highest F's value; Table 2 depicts the means, standard deviations (SD) and F's tests values (between-groups multivariate analysis).

Standardized canonical discriminant function coefficients showed an index of the importance of each predictor for diagnosis with the sign indicating the direction of the relationship. A significant increase in values of Cs (spherical tensor), Cl (linear tensor) and AD (axial diffusivity) were the strongest diagnostic predictors. The variable coefficients stood out (for these data) as those that strongly predicted allocation to the normal-brain or tumorbrain group. The coefficient score decrement was proportional to less successful diagnostic predictors (Table 3A).

Structure Matrix Data provided another way of indicating the relative importance of the diagnostic predictors by showing the correlations (Pearson coefficients) of each variable with each discriminate function. Many researchers consider the structure matrix correlations more accurate than the standardized canonical discriminant function coefficients.²⁶ By identifying the largest loadings for each discriminate function, different patterns of loading variables can be seen. We found AD, MD, p, q, Cl, RD and FA, as the functions that best discriminate between normal brains and brains with tumor. A value of 0.30 was considered as the cutoff between important and less important variables (Table 3B).³⁰

The canonical discriminant function coefficients Table showed the unstandardized coefficients (b) that were used to create the discriminant function (equation), they operated just like a regression equation, allowing us to build a predictive model of brain status:

Brain status (normal brain vs. tumor infiltration) = -48.295 +11,443.557 (axial diffusivity, AD) +105.124 (longitudinal tensor, Cl) +26.804 (spherical tensor, Cs)

The discriminant function coefficients (b) indicated the partial contribution of each variable to the discriminate function controlling all other variables in the equation (Table 3C).

The group centroids values described each group in terms of its profile, using the group means of the predictor variables called centroids. The cutoff value was defined as the mean of the two centroids; if the discriminant score of the function of a new case was less than or equal to the cut-off, the case was classed as 1 (brain with tumor), whereas if it was above the cut-off, it was classed as 0 (normal brain). In our study, normal brains had a mean of 1.483 while brains with GBM produced a mean of -1.334; the cut-off for the function at group centroids showed a calculated value of 0.149.

For the final part of the DA we performed a classification phase using the cross-validated set of data to present the power of the discriminant function. These results revealed that 92.7% of patients were classified correctly into "normal brain" or "brain with GBM" groups, this value corresponded to the overall predictive accuracy of the discriminant function. Additional results of diagnostic tests performance including the 95% confidence intervals (C.I.) showed: sensitivity = 100.00 (80.49 – 100.00); specificity = 87.50 (67.64 – 97.34); (+) likelihood ratio = 8.00 (82.78 – 23.06); (-) likelihood ratio = 0.00 (-); (+) predictive value = 85.00 (62.11 – 96.79); and (-) predictive value = 100.00 (83.89 – 100.00).

The average discriminant (D) scores for each group and the group centroids were used as visual demonstrations of the effectiveness of the disTABLE 3. Independent variables included in the discriminant analysis. A, ordered by their Standardized Canonical Discriminant Function Coefficients (variables with larger coefficients stand out as those that strongly predict allocation to each diagnosis). B, Within-groups correlation matrix depicts the participant variables ordered by absolute size of correlation (Pearson coefficients) within function. The largest loadings for each discriminate function (AD was the largest) suggest the preference of diffusivity values that discriminates between normal- and brain-tumor groups. A value of 0.30 is considered as the cut-off between important and less important variables, notice that variables with (*) were not used in the analysis. C, unstandardized coefficients used to create a discriminant function operating just like a regression equation. Coefficients indicate the partial contribution of each variable to the discriminate function controlling for all other variables in the equation

А		В		С	
Standardized Canonical Discriminant Function Coefficients		Structure Matrix		Canonical Discriminant Function Coefficients	
Variable –	Function	- Variable -	Function	– Variable -	Function
	1		1		1
CI (linear tensor)	1.361	AD (axial diffusivity)	.748	AD (axial diffusivity)	11443.557
Cs (spherical tensor)	.962	MD (mean diffusivity)*	.568	CI (linear tensor)	105.124
AD (axial diffusivity)	.806	p (pure isotropic diffusion)*	.566	Cs (spherical tensor)	26.804
		L (total magnitude of the diffusion tensor)*	.553	(Constant)	- 48.295
		q (pure anisotropic diffusion)*	.533		
		CI (linear tensor)	.441		
		RD (radial diffusivity)*	.427		
		FA (fractional anisotropy)	.320		
		RA (relative anisotropy)*	.278		
		Cs (spherical tensor)	211		
		Age*	.124		
		Cp (planar tensor)*	.020		

criminant function. Histograms and box plots of the average D scores for each group were used as graphical demonstrations of the effectiveness of the discriminant function, the absence of overlap of the plots revealed an excellent discrimination (Figure 3 A-B).

Discussion

The lack of consensus regarding which DTIderived tensor metrics are the most meaningful³¹, and the scarce information about their diagnostic abilities, compelled us to evaluate whether a *global* approach might have clinical applicability. We consider our study an introduction to the method and proof-of-principle that a *global* approach using selected DTI-derived tensor metrics can differentiate normal brains from brains infiltrated with GBM, the selected metrics may function as biomarkers assembling a predictive model of tumor infiltration.

The relevant findings in our study showed that a multivariate DA of global measurements excluded the pair-wise comparisons from conventional tumor-region evaluations; the assembled statistically significant discriminant model of tumor brain impairment (for these data) needed only three *global* DTI-derived metrics: AD, Cl, and Cs.

Some advantages of a global approach using DTI metrics need to be mentioned: it decreases the bias associated with manual placement of a region of interest encompassing tumor regions; the tumor and edema regions are implicitly included in the evaluation; lesions not perceived by the radiologist's eye on conventional sequences would be included in a global assessment; it may avoid problems associated with partial volume effects, and inaccurate image coregistrations; DTI biomarkers can be applied to other tumors/neurological diseases; its acquisition does not need contrast, and its post processing method can be semiautomatic; these facts broad the clinical applicability with no significant increase in the cost of MRI examinations.

The selected biomarkers in our final model deserve a brief explanation: AD depicted the main influence (larger value of its b unstandardized coefficient); it represents the directional diffusivity describing the microscopic water movement parallel to axonal tracts. AD is one of the best biomarkers in the diagnosis of enhancing rim in GBM, but not for other tumor regions¹⁰; (this fact provides evidence



that a regional measurements may not be the most effective way to use DTI metrics in brain tumor imaging).14 AD has been studied in animal models of encephalomyelitis of the spinal cord^{32,33}, in unfixed ex vivo human brains with multiple sclerosis³⁴, in a model of axonal injury caused by stroke³⁵, and in optic neuritis. Cl and Cs on the other hand, along with FA, have been reported among the biomarkers with best overall performance in differentiating the cystic cavity in abscess from GBM.¹³ They show best diagnostic performance in the detection of normal-appearance white matter (NAWM) and the cystic cavity in brains with GBM.¹⁰ Cs represents spherical normalized coordinates of a nonorthogonal DTI-derived tensor for each voxel, and Cl corresponds to the linear case.³⁶

Several limitations in this study need to be addressed: because there have not been studies investigating a whole set of tensor metrics (not only FA and ADC) in a global approach^{1,37,38}, it is difficult to compare our results with others in the literature. Further studies might include comparisons with other brain tumors, the influence of variables like radiation necrosis, inflammatory and demyelinating diseases; and tumor infiltration categories such as post-surgery and post-radiotherapy; all of them were beyond the scope of this study. A concern of using DTI-metric values with high correlations (correlations up 0.8 or 0.9), as we observed in our data, might be raised because in those situations one variable is a near-linear combination of the other variable (the variable provides information that is redundant to the information available in one or more of the others, making matrix inversion unreliable).²⁹ The usual solution is a deletion of the redundant variable, however, because we have a compelling theoretical reason to retain all variables in this study (to evaluate the simultaneous discriminant ability of 11 global tensor metrics), the IBM® SPSS® Statistics software protects against multicollinearity and singularity through computation of pooled within-cell tolerance (1-squared multiple correlation, SMC) for each variable. SMC is the squared multiple correlation of a variable where it serves as the dependent variable (DV) with the rest

FIGURE 3. Visual demonstration of the effectiveness of the discriminant function. (A), histograms showing the distribution of discriminant scores for normal- and tumor-brains. (B), box plots of the average D scores. Both kinds of plots illustrate the distribution of the discriminant function scores for each group. The box-plots depict a visual demonstration of the excellent discrimination of the model by showing no overlap between groups.

Healthy brains

Brains with GBM

Predicted Group for Analysis 1

-2.0000

-4.0000

 (A)

(B)

 (\mathbf{C})

as independent variables (IV) in multiple correlation. Variables with insufficient tolerance are deleted from the analysis; this procedure is a part of the stepwise method in DA.22 Our discriminant model was able to explain a significant proportion of the variability in the data (67.56%), but may still have some errors in predicting individual diagnosis, so model validation should be done in subsequent studies. We acknowledge the linking of tensormetric values with the axonal-integrity status represents an oversimplification with respect to what is happening in brains with GBM, where complex tissue changes occur and affect water diffusivity: density of fiber, average diameters, degree of myelination, directional similarity, cellularity, viscosity, permeability, and histologic architecture; the DTI-tensor values are the effects of the summation of all these microstructural barriers.¹³

Several questions remain unanswered, for example, what is the relation of these tumor-DTI biomarkers with those of MR perfusion and spectroscopy? What is the association of DTI-biomarkers with the pattern of relapse and extension of resection in GBM? So far, only one study, to the best of our knowledge, has correlated a few regional DTI-tensor metrics with the survival of patients with GBM³⁹; thus the clinical value of global DTImetrics in predicting the overall survival has yet to be determined. As a phase II study, our research line will look for a sequel, applying the proven concepts in the follow-up of tumor-infiltration categories (post-surgery, post-radiotherapy, etc.) and in differential diagnoses (primary brain tumors vs. metastasis vs. demyelinating diseases).

Conclusions

Although we cannot affirm the superiority of global vs. regional DTI-derived tensor metrics in the evaluation of GBM yet, we can ascertain with certainty that there is an immediate clinical applicability of these biomarkers in assembling statistically significant predictive models able to announce the conversion of normal tissue to tumor infiltrated tissue before the conventional MR sequences show conspicuous findings. These principles could easily be extended to other neurological diseases. A first step in the advanced evaluation of brain tumors might include a global measurement of DTIbiomarkers able to pick up major infiltration zones. Due to the large number of variables (qualitative and quantitative) that must be analyzed in contemporary brain MRI by radiologists and neuroscientists conducting research on novel imaging biomarkers; multivariate techniques, like DA, may help in the generalization of knowledge beyond one setting.

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Conclusiones

- La medición de biomarcadores avanzados de resonancia magnética en pacientes con glioblastoma multiforme permiten la construcción de modelos predictivos de sobrevida. En nuestro proyecto, obtuvimos un modelo con significancia estadística que seleccionó tres biomarcadores avanzados:
 - o Cholina/N-acetil-aspartato
 - Lípidos-Lactato/Creatina
 - o Edad
- El resto de variables utilizadas en nuestro modelo (derivados de secuencias convencionales) no fueron seleccionadas.
- Es posible calcular 11 biomarcadores diferentes, derivadas de la secuencia tensor de difusión en el encéfalo de sujetos sanos y en pacientes con glioblastoma multiforme.
- Existe una variabilidad significativa en mediciones derivadas del tensor de difusión en diferentes regiones y a lo largo de trayectos axonales de estructuras nerviosas del encéfalo normal. Conocer esta variabilidad permitirá un correcta interpretación de hallazgos en el estado normal y patológico.
- Cada uno de estos nuevos biomarcadores muestra un desempeño diagnostico diferente dependiendo de la región tumoral evaluada: tumor viable, necrosis, edema y tejido normal. Este hallazgo limita la realización de un abordaje regional para medir la infiltración tumoral en pacientes con glioblastoma multiforme y justifica la necesidad de un método global para detectar la infiltración tumoral.
- El análisis multivariado discriminante aplicado a los biomarcadores avanzados globales también permiten calcular un modelo significativo que permite la diferenciación de tejido sano y enfermo y que tiene capacidad predictiva.
- El proyecto de investigación permitió crear una línea de investigación original y con gran potencial para producción científica de calidad, no existente actualmente en México y que podría traer beneficio inmediato al diagnostico y tratamiento de pacientes.
- Nuestra línea de investigación continuará su tarea de responder varias preguntas referentes al uso de las técnicas avanzadas de resonancia magnética en el estudio de los tumores cerebrales, entre las mas inmediatas se incluyen:
- ¿Cual es la correlación entre espectroscopia, difusión, perfusión y tensor de difusión como biomarcadores de una misma región tumoral?
- ¿Cual es la capacidad diagnostica multivariada de estos biomarcadores para la sobrevida en GBM en un abordaje global versus regional?