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Evaluation of early microstructural changes in the R6/1 mouse model of Huntington's disease by ultra-high field diffusion MR imaging

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ABSTRACT

Diffusion MRI (dMRI) has been able to detect early structural changes related to neurological symptoms present in Huntington's disease (HD). However, there is still a knowledge gap to interpret the biological significance at early neuropathological stages. The purpose of this study is two-fold: (i) establish if the combination of Ultra-High Field Diffusion MRI (UHFD-MRI) techniques can add a more comprehensive analysis of the early microstructural changes observed in HD, and (ii) evaluate if early changes in dMRI microstructural parameters can be linked to cellular biomarkers of neuroinflammation. Ultra-high field magnet (16.7T), diffusion tensor imaging (DTI), and neurite orientation dispersion and density imaging (NODDI) techniques were applied to fixed ex-vivo brains of a preclinical model of HD (R6/1 mice). Fractional anisotropy (FA) was decreased in deep and superficial grey matter (GM) as well as white matter (WM) brain regions with well-known early HD microstructure and connectivity pathology. NODDI parameters associated with the intracellular and extracellular compartment, such as intracellular ventricular fraction (ICVF), orientation dispersion index (ODI), and isotropic volume fractions (IsoVF) were altered in R6/1 mice GM. Further, histological studies in these areas showed that glia cell markers associated with neuroinflammation (GFAP & Iba1) were consistent with the dMRI findings. dMRI can be used to extract non-invasive information of neuropathological events present in the early stages of HD. The combination of multiple imaging techniques represents a better approach to understand the neuropathological process allowing the early diagnosis and neuromonitoring of patients affected by HD.

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1. Introduction

Huntington's disease (HD) is a hereditary neurodegenerative disease most recognized by slow and progressive abnormal movements described in detail by George Huntington (Walker, 2007) in the 19th century . HD is associated with a distinctive phenotype, including chorea and dystonia, discoordination, cognitive decline,

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(*HTT*) gene is responsible for an atypically long polyglutamine (poly Q) expansion in the huntingtin (HTT) protein (Finkbeiner, 2011). Such polyQ expansion (more than 35 CAG repeats) makes HTT prone to misfolding, aggregation, and accumulation (Zheng and Diamond, 2012). Eventually, this leads to neurodegeneration, particularly in medium spiny neurons (MSNs) of the grey matter (GM) nuclei of the striatum (STR) which has been well described (Vonsattel et al., 1985). HD is particularly devastating due to its multiple cognitive features, with a relentless course of over 10–20 years. Early HD symptoms have been attributed to impairment of the frontal cerebral cortex (particularly layers III, V, and VI) (Zheng and Diamond, 2012). However, new evidence suggests

and behavioral difficulties. The genetic mutation of the huntingtin





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that other brain areas might be also involved, particularly the hippocampus, which has been under analysis in different mouse models and HD patients (Begeti et al., 2016; Harris et al., 2019; Quirion and Parsons, 2019). Unfortunately, there are no current therapeutic strategies to stop or slow down the disease progression, and fatality frequently occurs about 20 years after diagnosis (Roze et al., 2010).

Advancements in MRI techniques have opened a new window of preclinical microstructural exploration in neuroanatomical regions particularly susceptible to HD (Shim et al., 2019). Current improvements in gradient fields and imaging resolution have been able to non-invasively capture features, with improved accuracy in vivo and ex vivo at the early stages of neurodegenerative diseases (Gatto et al., 2018a; Gatto et al., 2018b; Muller et al., 2020). The development of new diffusion MRI techniques has opened a new realm with a great number of parameters associated with microstructural anomalies in HD (Rosas et al., 2006). Diffusion tensor imaging (DTI), a mono-exponential Gaussian imaging technique, has been applied to obtain parameters mostly related to WM changes during the disease in mouse models (Douaud et al., 2009; Gatto, Rodolfo G. et al., 2020; Magnotta et al., 2009), whereas other more complex models like neurite orientation dispersion and density imaging (NODDI) are contributing with additional information in different diseases at the clinical level (Kamiya et al., 2020), with few but growing number of examples in the context of HD preclinical models. Thus, a more comprehensive approach combining the different diffusion MRI techniques could be applied to increase our understanding of the intricate cellular changes in the early stages of HD.

Murine models are the most widely used to study HD, based on their availability and practical use (Figiel et al., 2012; Switonski et al., 2012). Initial manipulation of the transgenic founder line (R/6 mouse) led to the R6/1 mice that express exon 1 of the human gene (HTT) carrying a 115 CAG repeat, previously characterized and widely used to test therapeutic compounds (Naver et al., 2003). Aggregates/inclusions have been detected in the striatum at about 8 weeks of age (Li et al., 2005). No qualitative evidence of massive cell death in the R6/1 mice was detected in the model (Naver et al., 2003). A longitudinal study showed that overall, striatal, cortical, and hippocampal volumes were different between WT and R6/1 mice. Additionally, agerelated deficits in both motor and cognitive behaviors were detectable by 11 weeks of age. However, no association between mHTT load and MRI-based regional brain changes were identified (Rattray et al., 2013). The slower progression of the disease in the R6/1 mouse makes it more suitable for investigating the long-term effects of treatments, as well as for analyzing changes in both, the presymptomatic as well as in the symptomatic state (Gatto and Weissmann, 2019). Thus, the evaluation of these models using different imaging techniques in the field of HD could help interpret MRI findings and early neuropathological mechanisms.

One of the central mechanisms involved in HD neuronal death is neuroinflammation; reactive inflammatory processes are one of the most studied at the early stages of HD (Politis et al., 2015). The activation of the components of the glia, particularly astrocytes with HTT inclusions have been detected and related to alterations in HD (Cisbani and Cicchetti, 2012). Thus, the evaluation of early associated neuroinflammatory biomarkers in physiological intact organisms is paramount to move research in this field forward. Overall, the purpose of this study is two-fold: (i) combine different diffusion techniques to agive a more comprehensive insight into the early microstructural changes observed in HD; (ii) evaluate if early microstructural changes are linked to cellular markers of neuroinflammation. To the best of our knowledge, this is the first study using microstructural multi-modality imaging methods in the R6/1 transgenic animal model of HD.

2. Material and methods

2.1. Animals

All animal procedures were performed according to the Guide for the Care and Use of Laboratory Animals, as adopted by the US National Institutes of Health, the Mexican Regulation of Animal Care and Maintenance (NOM-062-ZOO-1999, 2001), and in agreement with internal Cinvestav regulations. Mice had easy access to food and water and were checked daily to assess their level of well-being and health. Mice had ad libitum access to food and water in aseptic conditions with climate and light-controlled system and were checked daily to assess their level of well-being and health. R6/1 mice were obtained from Jackson Laboratories (stock #002809) with a C57BJ6 background, overexpressing 115 \pm 6 CAG repeats. Mice were genotyped by end-point PCR based on previous protocols (Segovia, 2002). To amplify the mutant fragment of the human huntingtin gene we used the following primers: forward 5GCA GCA GCA GCA GCA ACA GCC GCC ACC GCC, and reverse 5CGG CTG AGG CAG CAG CGG CTGT. To reproduce our results with more biological accuracy (as well as for future translational and reproducibility purposes), a mixed pool of male and female mice was used in this study, as has been previously done (Beery, 2018; Xiang et al., 2011). The R6/1 mice have been previously characterized and they develop motor symptoms between 14 to 20 weeks of age and die around 48 weeks (Ferrante, 2009). After a proper review of the literature, among the two-common method of sample size calculation in animal studies (Charan J et al 2013), A power analysis approach was used to calculate the sample size for a significance level (type 1 error) fixed at the level of 5%(p = 0.05) and a power (β) < 0.80 (Charan and Kantharia, 2013; Dattalo, 2018).

To disentangle alterations due to HD progression from normal aging, we initially considered two control groups of C57/BJ6 mice to match the experimental groups at the early presymptomatic stage of 11 weeks (P77), and a late R6/1 symptomatic group at postnatal week 30 (P320) with a total of 16 animals used: wild type (WT) controls and R6/1 mice (n = 4 per group). No significant differences were detected between both control groups.

2.2. Animal preparation for MRI imaging

Animals were deeply anesthetized with an overdose of pentobarbital sodium, then transcardially perfused with PBS and 4% paraformaldehyde (PFA) solution. After the skull was opened, mouse brains were extracted intact and immersed in PFA (> 48 hours). Before scanning, brains were soaked overnight in PBS to remove free fixative. Brain stacks (3) were placed in 10 mm diameter NMR tubes (New Era #NEML10-7, 300-400 MHz) and the tubes were filled with fluorocarbon oil (Fluorinert, 3M, Maplewood, MN) and gently stabilized using coronoid tips. Three (3) 10mm tubes were gently placed in a 25mm coil. Images from all brains were acquired with a 17.6 T vertical-bore Avance II scanner using a 25-mm RF coil, Micro-2.5 gradients, and Paravision 6.0 software (Bruker, Karlsruhe, Germany). The brain settings allowed the placement of a stack of 3 brains per 10mm MRI compatible tubes.

2.3. MR image acquisition and processing

Two (2) MRI scans sessions (involving 6 mice brains each) were performed (total mouse brains N = 12). Three (3) experimental groups (WT pre and symptomatic control, R6/1(11weeks

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presymptomatic), and R6/1 symptomatic (30 weeks), coronally centered and oriented along the rostral-caudal axis of each brain (n = 4 per animal group). A total of 170 MRI slices were acquired using a spin-echo sequence with TR= 10000 ms and TE =20 ms, interleaved 0.2 mm thick slices, with a field of view = $25 \times 25 \times 34 \text{ mm}^3$ in each block of slices, in-plane acquisition matrix = $125 \times 125 \times 170$, for an isotropic image resolution of 200 microns, $b = 0 \text{ s/mm}^2$ for each shell: $b = 1000 \text{ s/mm}^2$ in 20 directions, and $b = 2500 \text{ s/mm}^2$ in 64 directions, with 3.5 ms duration (δ) and 11 ms of separation (Δ) of the gradient pulses. This acquisition was averaged twice for a total time of approximately 19 hrs. per set in the cool bore (20 degrees Celsius) of the 17.6 T magnet (750 MHz/89 mm Bruker Avance III HD). Due to the potential hardware vibrations, and field inhomogeneity, susceptibility corrections were performed. All the DWIs to the baseline image (b0), correction of the eddy currents, as well as motion and artifacts corrections were accounted for using standard preprocessing procedures (Andersson et al., 2003; Andersson and Sotiropoulos, 2016). DTI parameters data processing was performed using FSL (Jenkinson et al., 2012) to calculate fractional anisotropy (FA).

For NODDI calculations, we followed the MATLAB scripts provided in (Zhang et al., 2012). For NODDI fitting, the initial condition for the intrinsic diffusivity (D_{in}) of the neural tissue was set to 0.6×10^{-3} mm²/s, particularly optimized for an overall murine brain tissue as described in earlier work (Wang et al., 2019). A default value of isotropic diffusivity (2.0 \times 10–3 mm2/s) was used for D_{iso} value as indicated in previous studies (Holz et al., 2000). Applying equations already implemented in the original MATLAB algorithms in ex vivo fixed tissue, an additional dot-compartment (diffusion is restricted in all directions) was included for the fitting (Dhital et al., 2018). Manual segmentation, volumetric calculations, and diffusivity parameters were acquired for three different brain structures: a) Superficial GM cortical areas (superficial GM), b) subcortical areas (deep GM); and c) WM tracts, using ITK-SNAP (www.itksnap.org) (Yushkevich et al., 2006; Yushkevich et al., 2016). Regions of interest (ROIs) were manually defined following stereotaxic coordinates and landmarks presented by the Paxinos mouse brain atlas (Paxinos and Franklin, 2004). Four (4) superficial GM ROIs areas were selected; frontal accessory cortical area (FrAc), the Cingular cortex (Cing), the primary motor cortex (M1), and the supplementary motor cortex (M2). Two (2) deep GM areas were selected; the striatum (STR) and the hippocampus (HC). Additionally, we analyzed the corpus callosum (CC), a WM area generally affected by HD disease (Gatto and Weissmann, 2019; Phillips et al., 2013; Phillips et al., 2014).

For tractography reconstructions and exploration purposes, a multi-shell diffusion scheme was used, with the b-values of 1000 and 2500 s/mm² and sampling directions used were 20, and 20, respectively. The in-plane resolution was 0.15 mm. The slice thickness was 0.15 mm. The b-table was checked by an automatic quality control routine to ensure its accuracy (Schilling et al., 2019). The diffusion tensor was calculated. A deterministic fiber tracking algorithm (Yeh et al., 2013) was used. A seeding region was placed at the whole brain with a volume size of 4.2e+02 mm^{3.} The ROI was placed in the striatum region with a volume size of 5 mm³. The anisotropy threshold was 0.0960938. The angular threshold was 90 degrees. The step size was randomly selected from 0.5 to 1.5 voxels. The fiber trajectories were smoothed by averaging the propagation direction with a percentage of the previous direction. The percentage was randomly selected from 0% to 95%. Tracks with a length shorter than 0 or longer than 20 mm were discarded. A total of 50000 seeds were placed. ROIs were used for brain parcellation, and the connectivity matrix was calculated by using the count of the connecting tracks. The connectivity matrix and graph theoretical analysis was conducted using DSI Studio (http://dsi-studio.labsolver. org).

2.4. Immunohistochemistry

After MRI scanning, oil media was removed, and brains were placed in increasing concentration of sucrose solutions [5-30 %] for an additional 24 hrs for cryo-protection. After embedding in optical cutting temperature (OCT) polymer compound (Tissue Tek, Sakura, Finetek, cat #4583), 50 µm thick brain sections were obtained using a microtome (Leica cryostat CM 1850 Cryostat, Buffalo Grove, IL). Brain sections were mounted on slides (Fisher-brand Superfrost, cat# 12-550-15) and dried for 15 minutes. Then, the OCT was removed by washing three times with Tris base buffer (TBS) and permeabilized with 0.25% Triton x-100 (20 min) and treated with blocking solution (1% BSA, 0.01% Triton x-100 in TBS) for an hour at room temperature (RT). Slides were incubated then with the primary antibody overnight in blocking solution, washed in TBS, and incubated with the secondary antibody for 60 min in blocking buffer. After washing, slides were incubated in propidium iodide (PI) for 15 min to detect nuclei. After a final wash in TBS, slides were dried and mounted in Vecta-Shield mounting media (Vector Laboratories, Burlingame, CA). The antibodies used were rabbit polyclonal antibody against glial fibrillary acid proteins (GFAP) (DAKO, Glostrup, Denmark; 1:700), mouse monoclonal against (Ionized calcium-binding adaptor molecule 1 (anti-Iba1) (Wako, 019-19741, 1:200). PI (Molecular Probes) 20 mM was used in a solution 10 uM in TBS containing Tx-100 (0.01%) and RNAse.

Cortical and hippocampal images in mice were acquired with an Olympus FV300/BX61 microscope. Alexa-647 and Alexa-488conjugated secondary antibodies, and PI were excited using Argon (lambda: 488 nm), Helium-Neon (lambda: 543 nm), Helium-Neon (lambda: 633 nm) lasers, and a transmission light detector as previously described (Castellanos et al., 2020). For quantitative experiments, identical laser power and acquisition settings were used. Representative histological sections were selected matching stereotaxic coordinates using MRI brain imaging (The Mouse Brain: 1st Ed, Watson & Paxinos, 2008). Measurements were done using open-source ImageJ software (Collins, 2007; Schneider et al., 2012). Mean intensity values were used for quantitative purposes. Four representatives (4) ROIs from four (4) consecutive immunohistochemistry (IHC) slides (n = 16 ROIs) were selected for analysis (Gatto, R. G. et al., 2020). In the case of CA1, CA3, and DG areas, ROIs were selected from the PI channel to determine the different layers (pyramidalis, radiata) and mean grey values obtained (background corrected) from the antibody channel (n = 15-20 ROIs) in an analysis similar to that performed by Ugolino et al. (Ugolini et al., 2018). For fractal and branch analysis, we used the plugin FracLac and Skeleton respectively in ImageJ. The analysis performed was similar to that described by Young et al (Young and Morrison, 2018). Representative fluorescent micrographs were increased in brightness and contrast equally for all conditions (WT, presymptomatic, and symptomatic) to visually show the differences mentioned in the legends.

2.5. Statistical analysis

Quantitative data were tabulated and analyzed using GraphPad Prism version 6 software (La Jolla, CA). According to animal protocols, a minimum group size of animals was established using power analysis and sample size calculations based on the results from previous pilot work (Gatto et al., 2015). Multiple comparison corrections using the false discovery rate (FDR) method were performed (Benjamini, 2010; Benjamini and Hochberg, 1995). For



Fig. 1. Regions of interest (ROIs) studied in the R6/1 mouse brain. (a) Sagittal section of a mouse brain showing ROIs measured in a WT mouse at 16.7T and a 200 microns voxel resolution. (b, c) Coronal and longitudinal sections of the same brain. Consecutive sections (a' anterior) to (i' posterior) are used to show the distribution of different white matter (WM) and grey matter (GM) structures. Abbreviations: FrAc, frontal accessory area; Cing, Cingular cortex area; M1, primary motor area; M2, Supplementary motor area; HC, hippocampus; STR, Striatal Area; CC, corpus callosum. Note that additional section's planes (1,2,3) are used to describe the anterior, medial, and posterior part of the hippocampus. Scale bar = 1mm

quantitative analysis, one-way ANOVA and Tukey's post-hoc tests were used to determine statistical differences among different experimental animal groups. A value of p < 0.05 was used to demonstrate statistical significance. Additionally, non-parametric statistical tools (Mann–Whitney test) were used. Error bars in all tables represent the standard error of the mean (s.e.m.). In the graphs, data are visualized as box plots representing the median value and whiskers representing the upper (25%) and lower (75%) quartile.

3. Results

3.1. UHFD-MRI can capture volumetric changes in GM cortical areas of the R6/1 mice

Presymptomatic (11 weeks) and symptomatic (30 weeks) R6/1 and control animal brains were scanned to analyze the different regions shown in Fig. 1. As illustrated in Fig. 2, volumetric changes were captured in different segmented areas. Whole-brain volumes were significantly reduced in symptomatic R6/1 compared to WT controls (WT-11wks) = $422.8 \pm 17.2 \text{ mm}^3 \text{ vs } \text{R6/1} (30\text{wks}) = 366.8$ \pm 7.27 mm³, p < 0.02) (13.3 % decrease), but not in the presymptomatic mice group (R6/1 (11wks) = $380.3 \pm 5.26 \text{ mm}^3$, p <0.05) (9.9 % decrease). In specific WM regions, like the corpus callosum (CC), a significant reduction was attained only in the R6/1 symptomatic group (WT-11wks = $13.7 \pm 0.5 \text{ mm}^3 \text{ vs } \text{R6/1}$ $(30 \text{ wks}) = 11.1 \pm 0.4 \text{ mm}^3$, p < 0.02) (18.97 % decrease). In deep GM regions, like the striatum (STR) and hippocampus (HC), only significant changes were seen in the symptomatic stages (STR, WT-11 wks = 33.7 \pm 2.62 mm³ vs. R6/1 (30 wks) = 24.9 \pm 2.4 mm³, p <0.04), (HC, WT-11wks) = $22.0 \pm 0.3 \text{ mm}^3 \text{ vs. } \text{R6/1} (30\text{wks}) = 18.3$ \pm 1.3 mm³, p < 0.03) with a decrease of 26.1% and 16.8% respectively. Changes were also detected in M1 and M2 motor cortex areas at the symptomatic stage. In other areas of the cortex (CCX), such as the frontal accessory (FrAc) (WT-11wks = 13.9 \pm 2.3 mm³ vs R6/1 (11wks) = 7.35 \pm 0.45 mm³, p < 0.02) (47.1% decrease), and cingulate cortex (Cing) (WT-11wks = 7.57 \pm 0.52 mm³ vs R6/1 (11wks) = 5.0 \pm 0.57 mm³, p < 0.01) (33.97% decrease), significant presymptomatic volumetric decreases were accounted for. Additional normalization of each brain structure by the total brain volume on each experimental group and effect size analysis (Daly and Cohen, 1987; Rosenthal and Rosnow, 1984) is shown in Supplementary Table 1. These results show that the R6/1 model depicts changes as already described in other HD mouse models and was further analyzed.

3.2. Enlargements of the R6/1 mice ventricular system can be detected by UHFD-MRI

An increase in ventricular volumes was observed in the R6/1 mice compared to the WT control group (Fig. 3). These changes were substantially larger for the lateral (WT-11wks = 1.49 ± 0.07 mm³, vs. R6/1 (11wks) = 1.62 ± 0.05 mm³ vs. R6/1 (30wks) = 5.18 ± 0.28 mm³) and third ventricular systems (WT-11wks = 2.23 ± 0.08 mm³ vs. R6/1 (11wks) 2.33 ± 0.09 mm³, vs. R6/1 (30wks) = 3.13 ± 0.13 mm³) than in the fourth ventricle (WT-11wks = 1.09 ± 0.05 mm³ vs. R6/1 (11wks) = 1.30 ± 0.11 mm³ vs. R6/1 (30wks) = 1.45 ± 0.17 mm³). However, statistical significance was only attained between control and R6/1 groups at later stages (lateral ventricle: WT-11wks vs. R6/1 (30wks), p < 0.0001), (third ventricle; WT-11wks vs. R6/1 (30wks), p < 0.001) and (fourth ventricle).



Fig. 2. UHD-MRI MRI can capture early volumetric changes in specific GM cortical areas of the R6/1 mice. (a) MRI Three-dimensional (3D) reconstruction of three representative mouse brains from the control wild type (WT) and Huntington disease mouse (R6/1) at the presymptomatic (11wks) and symptomatic stage (30wks). (b, c) Manual segmentation and 3D reconstruction of susceptible WM and GM areas of transgenic HD mouse brain R6/1. (d) Quantitative volumetric measurements of the entire brain and areas segmented in the WM deep GM. Note that none of these areas show presymptomatic changes. (e) Regions from superficial GM areas of the brain. Except for the frontal accessory cortical area, no significant difference between control and R6/1 mouse was detected in the presymptomatic stage of the disease. Abbreviations: FrAc, Frontal Accessory area; Cing, Cingular cortex area; M1, Primary Motor area; M2, Supplementary Motor area; HC, Hippocampus; STR, Striatal Area; CC, corpus callosum. Scale bar = 1mm. (n = 4 per group), (# = p < 0.05).

tricle; WT-11wks vs. R6/1 (30wks), p < 0.01). These findings agree with those described in other HD mouse models and thus led us to detect additional changes via contributing MRI techniques.

3.3. DTI can detect early microstructural anomalies in the R6/1 mice

In WM tracts particularly affected in the context of HD (like the CC). the averaged FA reached a significant reduction in

presymptomatic animals (11 weeks), specifically in its anterior portion or genu (Figs. 4 and 5a). However, no other significant changes in FA were detected in the CC body or splenium regions. Analysis of the striatum showed a spread increase of FA across the entire region as the disease progressed towards the symptomatic stage (Figs. 4 and 5b). Moreover, the hippocampus showed an early initial localized FA decrease in the middle (labeled as "HC plane 2" in Fig. 1) spreading to its entire region at the symptomatic stage



Fig. 3. Enlargement of the R6/1 mice ventricular area cannot be detected in the early R6/1 mice model of HD. (a) MRI coronal sections from control (WT) and transgenic HD mouse brains (R6/1). Lateral ventricles are masked (red), third, and fourth ventricular systems are marked in orange and yellow respectively (arrows). (b) Diagram of the ventricular system in the mouse brain showing the location of the coronal planes (a', b' and c') used in (c) Three-dimensional reconstruction of the entire ventricular system in a WT mouse. (d) Quantitative volumetric analytics of the ventricular system. Note that no statistical changes in ventricular volumes can be detected between control and R6/1 mouse at the early stages. Abbreviation: L vent, lateral ventricles; III vent, third ventricle, and IV, fourth ventricle. Scale bar = 1mm. (# = p < 0.05, ## = p < 0.01, ### = p < 0.001). (n = 4 per group).



Fig. 4. Fractional anisotropy (FA) dMRI sections show progressive microstructural changes in brain regions of interest. Changes in FAs in the anterior part of the corpus callosum (a), striatum (b), Hippocampus (c), and frontal accessory areas (d). One- way ANOVA and Tukey post-hoc comparison was established between WT-11wks & WT-30wk and their corresponding R6/1 mice groups at 11wks and 30wks). (# = p < 0.05, ## = p < 0.01, ### = p < 0.001). (n = 4 per group).

(Figs. 4 and 5c). Cortical FrAc and Cing regions (Figs. 4 and 5d), presented a broader change between normal and disease groups at the very early presymptomatic period fundamentally located to the more anterior portions of the brain. Cortical areas, as the M1 and M2, showed more modest and localized early reductions in FA as well. A more detailed account of DTI derivatives from the segmented ROIs, including axial (AD = λ_1), radial (RD = $\lambda_2 + \lambda_3$)/2, and mean diffusivity (MD = $\lambda_1 + \lambda_2 + \lambda_3$)/3) are described in Table 1. Analysis of brain diffusion tensor maps, as well as deterministic tractography reconstructions, have shown a significant and progressive loss in terms of axonal connectivity in the R6/1 mice groups, compared to the WT control (Figs. 6 and 7).

3.4. Changes in brain tissue compartments can be detected by NODDI parameters

Using NODDI as a multi-shell technique in our R6/1 mouse model, we were able to determine changes in intra-neurite, extraneurite, and free water compartments (Zhang et al., 2012). In the genu portion of the CC, we observed a significant decrease in the intracellular volume fraction (ICVF) at the HD presymptomatic and symptomatic stages, as well as an increase in the orientation dispersion index (ODI), and the isotropic volume fraction (IsoVF) (Fig. 8a). In cortical FrAc and Cing areas, a progressive increase in symptomatic values of IsoVF was detected, as well as in subcortical GM structures. However, in the HC, IsoVF changes were seen more prevalent in the presymptomatic than the symptomatic phase of the disease (Fig. 8b). A complete recollection of the changes in NODDI parameters across each segmented ROI is shown in Table 2.

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Fig. 5. Analysis of consecutive spatial trends in fractional anisotropy (FA) dMRI sections shows progressive microstructural changes in different brain regions of the R6/1 mice. (a) FA changes along with the CC. Note that the anterior section (Genu) shows early changes. (b) Contrary to other regions, FA changes in STR show an increasing trend as the disease progresses, predominantly in the posterior regions. (c) An increase in the FA occurs in the medial section of the HC (see MRI section number 2 in Fig. 1a). (d) An earlier (11wks) and more anterior decrease in FA can be located in the FrAc and Cingular GM areas compared to M1 and M2 cortical areas of the R6/1 mouse. Abbreviations: FrAc, Frontal Accessory area; Cing, Cingular cortex area; M1, primary motor area; M2, Supplementary motor area; HC, hippocampus; STR, Striatal Area; CC, corpus callosum. Upper brackets = (WT-11wks vs. R6/1-11wks, groups), lower brackets (WT-11wks vs. R6/1 30 wks. groups) (# = p < 0.05, # = p < 0.01, ## = p < 0.001). (n = 4 per group).

3.5. Early cellular neuroinflammatory structural biomarkers in the R6/1 mouse brain

The presence of early glia anomalies in HD has been documented in previous work. In this preclinical model, we evaluated the presence of GFAP as a maker to track astrocytes. Our studies found increased levels of this protein as the disease progressed, in susceptible structures, such as the CCX, CC, and STR of the presymptomatic R6/1 mice compared to the WT control (Fig. 9a-c). These results were larger in the middle HC region, particularly in early stages (Fig. 9d).

A more thorough examination of the different regions of the HCs is shown in Fig. 10. The HC was subdivided into different areas, and ROIs in each were established to determine the intensity of the GFAP signal, the number of astrocytes, and morphological alterations (Fig. 10a-c). The plots show that the intensity of the GFAP signal was higher in all areas (CA1, CA3, and hilus) for the symptomatic animals. Interestingly, the CA3 area showed the highest GFAP signal for presymptomatic animals, similar to what was observed for FA values in HC middle portions (Figs. 5 and 6). We focused on the CA3 area for further analysis. A higher density of astrocytes was evidenced in the R6/1 presymptomatic mice compared to the WT control or symptomatic counterparts (Fig. 10b, bottom). Additionally, we performed a structural analysis using skeletonizations and fractals to compare the cell morphologies (Fig. 10c, explained in materials and methods). The parameters obtained, the fractal dimension (FD) and lacunarity (L), offer a way of quantifying and comparing geometrical complexity (Di Leva, 2016), and have been used extensively in the field of neuroscience as well as many others (Kapitulnik and Deutscher, 1984). In general terms, a higher FD reflects a higher irregularity, while

L complements the analysis yielding a measure of heterogeneity. The representative astrocyte images show a higher FD (fractal dimension) and L (lacunarity) in the R6/1 compared to WT-11wks mice evidencing a higher level of complexity, even at the presymptomatic stage.

The microglia (stained with Iba1) has been documented as playing a role at the later stages of HD (Palpagama et al., 2019). Results from our analysis confirmed this with a significant increase in the levels of the marker located in the medial areas of the HC at the presymptomatic stage of the R6/1 compared to the WT-11wks control mice (Fig. 11).

4. Discussion

This work has revealed early changes in dMRI microstructural parameters in the R6/1 HD mouse model and associated the findings with cellular biomarkers of neuroinflammation. At the macrostructural level, changes in volumes detected in this work (Fig. 2) are in line with those determined in HD patients, and documented in other HD mouse models, like the zQ175 (Heikkinen et al., 2012) and previous work in the R6/1 mice (Hansson et al., 1999). Other HD mouse models show volumetric changes as well. In the case of the R6/2 model, changes were detected via MRI in symptomatic animals, CCX, STR, and HC, but not in presymptomatic animals (Cepeda-Prado et al., 2012). Volume loss was also detected in the HdH150Q mouse brain at preand post-symptomatic stages (Steventon et al., 2016). In this study, R6/1 mice mimic changes described in clinical HD, with a volume loss of about 23% in total cortical GM (Halliday et al., 2009), and highest in STR nuclei (Ruocco et al., 2006).

An increase in HD ventricular size is another main feature described in HD (Hobbs et al., 2010). Likewise, preclinical studies in

Table 1				
DTI outputs from	segmented	R6/1	mice	regions

Brain structure	FA [†]			AD (10 ⁻⁴ mm/s ⁻¹)*			RD (10 ⁻⁴ mm/s ⁻¹)*			MD (10 ⁻⁴ mm/s ⁻¹)*			
	WT (11wks)	R6/1 (11wks)	WT (30 wks)	R6/1 (30wks)	WT (11wks)	R6/1 (11wks)	R6/1 (30wks)	WT (11wks)	R6/1 (11wks)	R6/1 (30wks)	WT (11wks)	R6/1 (11wks)	R6/1 (30wks)
CC (Genu)	0.59 ± 0.01	0.51 ± 0.01 ▼14% (##)	0.58 ± 0.01	0.45 ± 0.01 ▼22% (###)	4.61 ± 0.02	5.46 ± 0.002 ▲18% (##)	5.85 ± 0.01 ▲27% (###)	2.94 ± 0.01	3.26 ± 0.01 ▲11% (#)	3.52 ± 0.01 ▲20% (###)	2.01 ± 0.01	2.32 ± 0.002 ▲15% (##)	2.95 ± 0.01 ▲47% (###)
CC (Body)	0.56 ± 0.02	0.53 ± 0.02 ▼5%	0.56 ± 0.01	0.50 ± 0.03 ▼11% (#)	4.44 ± 0.01	4.46 ± 0.02 ▲<1%	4.68 ± 0.03 ▲5%	3.01 ± 0.01	3.07 ± 0.02 ▲2%	3.41 ± 0.13 ▲13% (##)	2.20 ± 0.01	2.28 ± 0.06 ▲4%	2.40 ± 0.13 ▲9% (#)
CC (Splenium)	0.57 ± 0.03	0.55 ± 0.03 ▼4%	0.53 ± 0.02	0.50 ± 0.06 ▼6%	4.68 ± 001	4.52 ± 0.02 ▲3%	4.98 ± 0,03 ▲6%	2.98 ± 0.11	3.19 ± 0.16 ▲7%	3.44 ± 0.54 ▲15% (##)	$2.21{\pm}~001$	2.23 ± 0.01 ▲<1%	2.45 ± 0,01 ▲11% (##)
CCX (FrAc)	0.34 ± 0.01	0.26 ± 0.02 ▼24% (###)	0.31 ± 0.07	0.24 ± 0.01 ▼23% (###)	3.69 ± 0.01	4.18 ± 0.01 ▲13% (##)	4.79 ± 0.02 ▲30% (###)	2.85 ± 0.01	3.27±0.01 ▲15% (##)	3.88 ± 0.02 $\blacktriangle 36\% (\# \# \#)$	2.47 ± 001	2.95 ± 0.01 ▲19% (##)	3.10 ± 0,01 ▲26% (###)
CCX (Cing)	0.35 ± 0.01	0.30 ± 0.01 ▼14% (##)	0.32 ± 0.06	0.28 ± 0.01 ▼13% (##)	4.36 ± 0.01	4.69 ± 0.01 ▲8% (#)	4.83 ± 0.02 ▲12% (##)	3.32 ± 0.01	3.67 ± 0.01 $\blacktriangle 11\% (\# \# \#)$	4.07 ± 0.02 $\blacktriangle 23\% (\# \# \#)$	2.36 ± 001	2.69 ± 0.01 $\blacktriangle 14\% (\#\#)$	3.18 ± 0,01 ▲35% (###)
CCX (M2)	0.34 ± 0.01	0.30 ± 0.02 $\checkmark 12\% (##)$	0.33 ± 0.09	0.27 ± 0.01 ▼18% (##)	4.06 ± 0.01	4.37 ± 0.01 ▲8% (#)	5.13 ± 0.02 $\blacktriangle 26\% (\# \# \#)$	3.02 ± 0.01	3.36 ± 0.01 ▲11% (##)	3.83 ± 0.01 ▲27% (###)	2.48 ± 001	2.86 ± 0.01 $\blacktriangle 15\% (\#\#)$	3.19 ± 0,01 ▲29% (###)
CCX (M1)	0.34 ± 0.01	0.30 ± 0.01 $\checkmark 12\% (\#\#)$	0.32 ± 0.06	0.26 ± 0.01 ▼19% (##)	3.88 ± 0.01	4.34 ± 0.01 ▲12% (##)	4.95 ± 0.01 ▲28% (###)	3.16 ± 0.01	3.40 ± 0.01 ▲8% (##)	3.86 ± 0.02 $\blacktriangle 22\% (\# \# \#)$	2.78 ± 001	3.19 ± 0.01 ▲15% (##)	3.56 ± 0,01 ▲28% (###)
НС	0.36 ± 0.01	0.31 ± 0.01 $\checkmark 14\% (\#\#)$	0.34 ± 0.01	0.28 ± 0.01 $\checkmark 18\% (##)$	4.36 ± 0.02	4.76 ± 0.01 ▲9% (#)	5.18 ± 0.01	3.10 ± 0.02	3.57 ± 0.01 ▲15% (##)	3.71 ± 0.02	2.66 ± 001	3.08 ± 0.01	3.31 ± 0,01 ▲24% (###)
STR	0.29 ± 0.01	0.42 ± 0.01 ▲45% (###)	0.31 ± 0.01	0.49 ± 0.02 ▲58% (###)	4.88 ± 0.01	4.51 ± 0.02 ▼13% (##)	440 ± 0.01 ▼10% (##)	3.69 ± 0.01	2.75 ± 0.01 ▼25% (##)	2.61 ± 0.01 ▼29% (###)	3.09 ± 001	2.62 ± 0.01 ▼15% (##)	2.44 ± 0,01 ▼21% (###)

AD, axial diffusivity; CC, corpus callosum; CCX, cortex; Cing, cingulate cortical area; DTI, diffusion tensor imaging; FA, fractional anisotropy; FrA, frontal accessory area; HC, hippocampus; M1, primary motor area; M2, secondary motor area; MD, mean diffusivity; R6/1, a transgenic mouse model of HD; RD, radial diffusivity; Splen., splenium; STR, striatum; wks., weeks; WT, wild type. FA values are compared to their WT (11wks) and WT (30wks) respective controls. AD, RD, and MD values are compared to WT (11wks). (n = 4 per group). \bigstar , percentage of increase. \forall , percentage of decrease. *R6/1 (11wks) and R6/1 (30wks). †R6/1 (11wks) and R6/1 (30wks). #p < 0.05. # # p < 0.01.

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Fig. 6. Anisotropic directional brain maps from coronal sections from diffusion tensor imaging (DTI) calculations in WT and R6/1 mice. Differences between control and R6/1 mice brain's maps are due to changes in tissue microstructure changing the color encoded information eigenvectors' direction. Scale bar = 5mm.

rodent models of NDDs have shown that the detection of changes in ventricular size measured by UHF-MRI is feasible. In clinical longitudinal MRI studies, Hobbs et al. estimated that HD ventricular volumes (adjusted for normal aging, sex, and total intracranial volume) increased more than in control groups, only one year before the onset of motor deficits (Hobbs et al., 2010). This short window of presymptomatic detection agrees with our findings (Fig. 3), where no significant changes in the R6/1 mouse were seen at an earlier stage of the disease. This finding confirms the fact that ventricular size is not a reliable bioimaging marker to predict HD at earlier stages.

At the microstructural level, (among the effects associated with the presence of mutant HTT), the loss of axonal connectivity (AC) is one of the earliest and main pathological events described in HD (Cowan and Raymond, 2006; Sieradzan and Mann, 2001). This AC loss can be visualized further with the use of MRI techniques in preclinical models, like the R6/1 mice (Table 3). dMRI has shown that DTI can evaluate brain injury and reveal changes in brain microstructure (Gupta et al., 2008; Gupta et al., 2010; Gupta et al., 2012). FA values reported in this work (Figs. 5 and 6) agree with the literature: as the progression of the disease relates to structure disorganization, we found lower FA values in WM and superficial GM structures - as well as opposite increased FA values in deep GM nuclei (STR). The unexpected increase of FA in the striatum might reflect a coherent breakdown of different grey and white matter constituents (e.g., microstructure simplification, cytoskeleton degeneration, myelin breakdown, loss of cortical axons (Fazio et al., 2018). Also, as Douaud et al suggested, an atypical increase of FA could be the result of the selective loss of some of the subcortical connections targeted by the degenerative processi.e., mainly the striatopallidal projections at an early stage of the disease that may turn the striatum into a seemingly more organized structure (Douaud et al., 2009) and reviewed before (Gatto and Weissmann, 2019). This reinforces the validity of the R6/1 as a model to further evaluate changes in HD. Moreover, specific delineation of anterior to posterior regions, within the structures of interest detected the longitudinal progression of the disease. Changes in neuronal connectivity can be reflected in WM orientations, probably related to a balance of several factors such as neurodegeneration and microstructure reorganization, as well as neuroplastic events (Gatto, 2020b). This is also detected in this study by the significant changes in DTI parameters (FA) and changes in the histological compartments determined by NODDI (ICVF and particularly ODI). This new information can improve our understanding of the topographic evolution of the disease and network alterations at the different stages (Fig. 7), as described in other preclinical models of HD (Chang et al., 2018).

Anomalies described by DTI can be connected to AC changes (Table 1). However, the study of neuroinflammation requires a dual intra and extracellular analysis. In the NODDI model, ODI parameters (and by extension ICVF and IsoVF scalars) have been directly related to the loss of connectivity and neuroinflammation phenomena. As an example, Yi et al. showed that the increased presence of microglia in the extra-neurite space contributed to alterations in water diffusivity, particularly detected in ODI values (Yi et al., 2019). An extra-neuronal compartment upsurge (based on an increase in ODI values) in the context of microglia infiltration, would be somehow difficult to explain. In this work, NODDI results reproduced the compartmental redistribution measured by ICVF and ODI in other NDDs (Gatto et al., 2018c) (Table 2). The alteration of other parameters associated with an increase of free water in the interstitial space (IsoVF) could more accurately represent the increase in interstitial water associated with inflammatory processes. This has been described in other neurodegenerative diseases (Gatto et al., 2018c) and depicted in (Fig. 8b).



Fig. 7. Global changes in axonal connectivity demonstrated by deterministic tractography reconstructions in the brain of the control and R6/1 mice. (a) Coronal dMRI section mapping local fiber directions in the control and transgenic presymptomatic (R6/1 11wks) and symptomatic HD mice (R6/1 30wks). (b) Directionally color encoded tractography reconstructions showing a higher number of diffusion tracts in the WT mice compared to the R6/1 mice groups with a loss of neuronal connectivity as the disease progresses. Abbreviations: WT, wild type; R6/1, transgenic mice with Huntington's disease mutation; S, Superior; L, lateral; P, Posterior. Scale bar = 1mm.



Fig. 8. Neurite Orientation Dispersion and Density Imaging model outputs from segmented brain ROIs in the R6/1 mice illustrate changes in different tissue compartments. (a) A combined decrease in intracellular ventricular fraction (ICVF) and a rise in orientation dispersion index (ODI), as well as an isovolumetric fraction (IsoVF), can be visualized in the anterior portion of the corpus callosum R6/1 mice (a main affected WM structure in HD). (b) Additional compartmental analysis in superficial (FrAc & Cing cortex) and deep (STR & HC) GM structures show a progressive increase in IsoVF. Note the broader changes in this parameter between WT and the R6/1 mice at the early stages of the disease. Abbreviations: WT, wild type; R6/1, transgenic mice with Huntington's disease mutation; WM, white matter; GM, grey matter; NODDI, Neurite Orientation Dispersion and Density Imaging; ICVF, Intracellular Ventricular Fraction; ODI, Orientation Dispersion Index; IsoVF, Isovolumic Fraction; FrAc, Frontal Accessory area; Cing, Cingular cortex area; HC, hippocampus; STR, Striatal Area; CC, corpus callosum. (# = p < 0.05, ## = p < 0.01, ### = p < 0.001). (n = 4 per group).

Table 2						
NODDI output vari	ables from	n segmented	R6/1	mice	brain	regions

Brain structure	ICVF [†]				ODI [†]				IsoVF [†]			
	WT (11wks)	R6/1 (11wks)	WT (30wks)	R6/1 (30wks)	WT (11wks)	R6/1 (11wks)	WT (30wks)	R6/1 (30wks)	WT (11wks)	R6/1 (11wks)	WT (30wks)	R6/1 (30wks)
CC (Genu)	0.45 ± 0.01	0.38 ± 0.01 ▼16% (##)	0.44 ± 0.07	0.34 ± 0.01 ▼23% (###)	0.16 ± 0.01	0.19 ± 0.01 ▲19% (##)	0.17 ± 0.05	0.21 ± 0.01 ▲24% (###)	0.0096 ± 0.0025	0.0172 ± 0.0022 ▲79% (###)	0.0108 ± 0.0014	0.0212 ± 0.0043 ▲96% (###)
CC (Body)	0.48 ± 0.01	0.41 ± 0.01 ▼15% (##)	0.48 ± 0.01	0.42 ± 0.01 ▼13% (###)	0.29 ± 0.02	0.40 ± 0.01 ▲38% (###)	0.31 ± 0.02	0.42 ± 0.01 ▲35% (###)	0.0031 ± 0.0015	0.0048 ± 0.0010 ▲55% (###)	0.0032 ± 0.0052	0.0067 ± 0.0013 ▲109% (###)
CC (Splenium)	0.48 ± 0.01	0.42 ± 0.01 ▼13% (#)	0.48 ± 0.01	0.40 ± 0.01 ▼18% (##)	0.15 ± 0.01	0.16 ± 0.01 ▲6%	0.18 ± 0.01	0.21 ± 0.01 ▲17% (##)	0.0036 ± 0.0008	0.0041 ± 0.0010 ▲14% (#)	0.0038 ± 0.0024	0.0073 ± 0.0011 ▲92% (###)
CCX (FrAc)	0.43 ± 0.01	0.36 ± 0.01 ▼16% (##)	0.41 ± 0.07	0.32 ± 0.02 ▼22% (###)	0.28 ± 0.01	0.40 ± 0.01 ▲43% (###)	0.36 ± 0.04	0.42 ± 0.02 $\blacktriangle 17\% (##)$	0.0014 ± 0.0004	0.0059 ± 0.0019 ▲321% (###)	0.0016 ± 0.0008	0.0081 ± 0.0019 ▲406% (###)
CCX (Cing)	0.40 ± 0.01	0.33 ± 0.01 ▼18% (##)	0.39 ± 0.05	0.29 ± 0.02 $\checkmark 26\% (###)$	0.29 ± 0.01	0.39 ± 0.01 $\blacktriangle 34 \% (\# \# \#)$	0.30 ± 0.02	0.43 ± 0.01 43% (###)	0.0019 ± 0.0006	0.0062 ± 0.0018 ▲226% (###)	0.0019 ± 0.0006	0.0078 ± 0.0021 11% (###)
CCX (M2)	0.34 ± 0.01	0.32 ± 0.01 ▼6%	0.33 ± 0.05	0.25 ± 0.01 $\checkmark 24\%$ (###)	0.27 ± 0.01	0.37 ± 0.01 ▲37% (###)	0.27 ± 0.01	0.35 ± 0.01 ▲30% (###)	0.0030 ± 0.0004	0.0037 ± 0.0007 ▲23% (###)	0.0032 ± 0.006	0.0056 ± 0.0015 ▲75% (###)
CCX (M1)	0.30 ± 0.01	0.27 ± 0.01 ▼10% (#)	0.32 ± 0.08	0.22 ± 0.01 $\checkmark 31\% (###)$	0.25 ± 0.01	0.34 ± 0.01 $\blacktriangle 36 \% (\# \# \#)$	0.25 ± 0.01	0.35 ± 0.01	0.0035 ± 0.0003	0.0048 ± 0.0011 ▲37% (###)	0.0038 ± 0.0015	0.0051 ± 0.0012 ▲34% (###)
НС	0.36 ± 0.01	0.30 ± 0.01 $\checkmark 17\% (\#\#)$	0.36 ± 0.03	0.27 ± 0.01 $\checkmark 25\% (###)$	0.24 ± 0.01	0.27 ± 0.01	0.25 ± 0.05	0.34 ± 0.01	0.0043 ± 0.0007	0.0118 ± 0.0019 ▲174% (###)	0.0046 ± 0.0005	0.0081 ± 0.0012
STR	0.33 ± 0.01	0.38 ± 0.01	0.33 ± 0.04	0.41 ± 0.01	0.35 ± 0.01	0.29 ± 0.01 17% (##)	0.34 ± 0.05	0.27 ± 0.02 $\sqrt{21\%} (\# \# \#)$	0.0056 ± 0.0018	0.0129 ± 0.0025	0.0060 ± 0.0010	0.0197 ± 0.0041

CC, corpus callosum; CCX, cortex; Cing, cingulate cortical area; FrA, frontal accessory area; HC, hippocampus; ICVF; intracellular volume fraction; IsoVF; isotropic volume fraction; M1, primary motor area; M2, secondary motor area; NODDI, neurite orientation dispersion and density imaging; ODI; orientation dispersion index; R6/1, a transgenic mouse model of Huntington's disease; STR, striatum; WT, wild type. A%, percentage of increase compared to the WT group. \mathbf{v} %, percentage of decrease compared to the WT group. n = 4 per group). #p < 0.05. ##p < 0.01. ##p < 0.001. \dag Note that FA values from R6/1 (11wks) and R6/1 (30wks) are compared to their respective controls WT-11wks and WT-30wks.



Fig. 9. Astrocytic activated markers showed neuroinflammatory processes from the early stage of the disease in the R6/1 mice. (a) Glial Fibrillary Acidic Protein (GFAP) histochemical staining for the frontal accessory (FrAc) cortical area showing a statistically significant difference in the number of astrocytes (arrows) from presymptomatic to symptomatic stages in the R6/1 mice. (b) Progressive increase in astrocyte GFAP in the anterior portion (genu) of the corpus callosum (CC) particularly in the R6/1 at 11 and 30weeks. (c) Surge in striatal astrocytes labeled with GFAP (arrows) detected at the early and late stages of the disease. (d) Quantitative GFAP analysis centered in the hippocampal dental gyrus showed a greater increase in the presymptomatic R6/1 mice compared to the symptomatic mice group, marking a stronger and earlier inflammatory response of this deep GM structure. All scale bars = 10 microns. Abbreviations: FrAc, Frontal Accessory cortical area; ccx. cortex area; HC, hippocampus; STR, Striatal Area; CC, corpus callosum; hil., hilus; WT, wild type; a.u., arbitrary units; R6/1, transgenic mice with Huntington's disease mutation; GFAP, Glial Fibrillary Acid Protein; PI, Propidium Iodine. (# = p < 0.05, ## = p < 0.001).

Our work detected a marked change in the diffusion change in the middle portion of the HC at a presymptomatic stage. The same area showed changes as detected in our histological studies using neuroinflammatory markers and centered in the HC middle portion, particularly in the CA3 region (Fig. 9). Glial (and diffusion) changes in HC could relate to cognitive alterations documented in clinical and rodent models, showing an increase of GFAP levels in CA1, CA3, and DG regions (Bondan et al., 2019). Additionally, studies have shown that structures such as the HC are particularly sensitive to the harmful effects of neuroinflammatory mechanisms in synaptic plasticity (Mancini et al., 2017). Detailed histological analysis in HC showed changes in astrocyte morphology, a common pathological feature preceding structural deficits that may drive disease progression. Astrocytes undergo a series of morphological and functional changes (overall described as astrogliosis) mainly characterized by swelling of the cell body, presence of larger and thicker processes, as well as increasing GFAPexpression (Pirici et al., 2009). Also, this cell type expresses AOP- 4 channels and thus their role in water uptake as has been described for other NDDs (Gatto et al., 2018a). Specifically, in HD, astrocytes show inclusions of mutant HTT (mHTT) proteins that have been related to alterations of its components (Cisbani and Cicchetti, 2012; Shin et al., 2005). This alone was shown to trigger HD pathology in mice that overexpress mHTT exclusively in astrocytes (Bradford et al., 2010).

In our MRI experiments, isoVF showed alterations in HC middle portions at the early stages; interestingly, as mentioned before, histology showed that GFAP increased levels were also prominent particularly in CA3 areas of the HC together with an increase in astrocyte complexity. Likewise, differences in GFAP levels in different areas of HC were also described for Alzheimer's disease (AD) transgenic mice compared to the WT counterpart (Ugolini et al., 2018), and a change in astrocytic morphology was reported also in some brain structures of the R6/2 mice (Vorisek et al., 2017). In previous work, astrocytes have been analyzed via the fractal dimension (FD) technique in different preclinical (Reichenbach et al.,



Fig. 10. Summary of changes observed in the hippocampus of the transgenic R6/1 mice model determined by GFAP staining of astrocytes. (a) Longitudinal sections stained against GFAP and using PI as a nuclear marker showing ROIs measured in a WT (A), R6/1 presymptomatic (B), and R6/1 symptomatic mouse at 10x (upper panel) and 20x (bottom panel) for proper visualization of the ROIs analyzed. (b) Representative plots of GFAP mean intensity of CA1, hilus, and CA3 as described in a. and quantification of astrocytes per ROIs and branches/cell in CA3 (bottom). (c) Representative images of astrocytes from CA3 str Rad of WT (A), R6/1 presymptomatic (B), and R6/1 symptomatic mouse analyzed to obtain a binary image, and outline, skeleton, and the convex hull and minimum bounding circle diagram with parameters (D, fractal dimension, and L, lacunarity) obtained via the plugin Fractac through image J as described in materials and methods. The plot shows the mean of Fractal Dimension (D) obtained for each condition. Notice the increase in D (complexity) in R6/1 animals compared to WT. Abbreviations: CA, Cornu Ammonis; pyr, pyramidalis; sr, strata radiatum; sp strata pyramidalis; D, fractal dimension; L, lacunarity; AU, Arbitrary Units; Scale bar = 1mm.

1992), and clinical studies (Pirici et al., 2009). In our studies, we found that different conditions (control, presymptomatic and symptomatic mice) bore differences in astrocytic morphological patterns (Fig. 10).

PFA fixation has an impact on diffusion parameters. As an example, clinical studies have shown that formalin fixation promotes a decrease in MRI relaxometry (T1 and T2), and a decrease in myelin water fraction (MWF) measurements, but accounting for minimal changes in diffusion tensor imaging (DTI) parameters such FA and mean diffusivity, (MD) (Shatil et al., 2018). Specifically, to mouse brain DTI and tractography studies, there are some changes in the scalar metrics one year after fixation, but these changes are sufficiently small, particularly in WM to support reproducible connectomes over a period ranging from two weeks to one-year postfixation (Xiao et al., 2020). Also, it is a well-recognized fact that formalin fixation alters water diffusion coefficients in the context of pathological conditions, such as the infarcted brain (Sun et al., 2005) as well as in DTI preclinical studies with fresh *ex vivo* and fixed brains tissue (D'Arceuil et al., 2007), which have to be accounted for in the ex vivo pipeline of high-quality images (Dyrby et al., 2011) to improve the validation strategies for the interpretation of microstructure (Dyrby et al., 2018). Particular to our *ex vivo* study, the chosen lower b-value (1000 s/mm²) for acquisition has been used in previous preclinical studies (Crombe et al., 2018; Wang et al., 2019). However, it is also probable that this particular lower b value maynot be enough (in the case of the NODDI model) to completely distinguish the restricted diffusion for the intra-neurite compartment from other compartments, thus a lower initial b value might



Fig. 11. Iba1 marker showed a progressive increase in microglia activiation in the deep regions of the R6/1 mouse hippocampus. Quantitative analysis of the R6/1 mouse hippocampal dental gyrus (DG) illustrates a progressive increase in microglia infiltration (arrows) with a maximum level at the symptomatic stage (30 weeks). Scale bar = 10 microns. Abbreviations: WT, wild type; a.u., arbitrary units; R6/1, transgenic mice with Huntington's disease mutation; Iba1, microglia cell markers; DG, Dental gyrus; Ili., hippocampus's ilium; PI, Propidium Iodine. (# = p < 0.05, ## = p < 0.01, ### = p < 0.001).

Table 3

Representative high-field MRI diffusion studies in R6 mice models of Huntington's disease

Study/year (type)	Mouse model	Diffusion technique	Field strength	b value/s & gradient directions	Brain structures analyzed	MRI diffusion parameters analyzed
Gatto et al 2019a (ex vivo)	R6/2 (160CAG)	DTI & CTRW	17.6T	b=1500 (s/mm ²), 12dir & 400, 800, 1500, 3000, 5000, 7500, 10,000, 12,000, 15,000, 20,000 (s/mm ²) .3dir	WM (CC genu)	DTI FA \checkmark , AD \blacktriangle , RD \blacktriangle , MD \blacktriangle CTRW $\alpha \checkmark$
(Gatto et al., 2019)	M&F, P60 ($n = 3$ per group)					
Vorisek et al 2017 (in vivo)	R6/2 (160CAG) F, 10, 12 wks. $(n = 4 \& n = 6)$	ADC _W	4.7T	b= 136, 329, 675, 1035, 1481,1825 (s/mm²).	GM (CCX, pall)	ccx (ADC _w -x \checkmark , ADC _w -y \checkmark , ADC _w -z \checkmark) pall (ADC _w -x \land , ADC _w -y \land , ADC _w -z \land)
(Vorisek et al., 2017)						· · · · · · · · · · · · · · · · · · ·
Gatto et al 2015b (ex vivo)	R6/2 (160CAG) M&F, P30 (n = 3 per group)	DTI	9.4T	b=1000 (s/mm²) 64dir	WM (CC genu)	FA 🔻
(Gatto et al., 2015) Xiang et al 2011 (ex vivo) (Xiang et al., 2011)	R6/2 (160CAG) M&F 12wks (n = 3 per group)	DTI	11.7T	b=1500 (s/mm ²)6dir	WM (CC genu & Splenium)	FA 🔻

AD, axial diffusivity; ADC_w, apparent diffusion coefficient of water; α , complexity coefficient; cc. corpus callosum; CTWR, continuous random walk model; ccx, cortex; dir, gradient direction; DTI, diffusion tensor imaging; FA fractional anisotropy; GM, gray matter values; MD; mean diffusivity; P, post-natal day; pall, pallidum; RD; radial diffusivity; T, Tesla; WM, white matter; wks., weeks; M, male; F, female; \blacktriangle , increased compared to WT. \checkmark , decreased compared to WT.



Fig. 12. Summary of changes observed in brain tissue of the transgenic R6/1 mice model captured by diffusion MRI and histopathological markers of neuroinflammation. (a) Macrostructural biomarkers derived from MRI scans showed a significant reduction in volume across regions, particularly susceptible to the mHTT expression. Significant changes in whole brain volume and segmented ROIs can only be detected at the symptomatic stage. (b) Microstructural changes can be measured in specific regions of the R6/1 mouse neuroanatomy. In such regions, early changes in GM and WM neuronal microstructure can be analyzed by applying diffusion tensor imaging (DTI). In WM tracts (represented as axonal cross-sections in Fig. 11d) and cortical GM, a significant decrease in fractional anisotropy (FA) can be associated with a neuronal injury. Neurodegenerative microstructural changes detected with the Neurite Orientation Dispersion and Density Imaging (NODDI) model using the cell degeneration and neuronal mass intracellular compartment (IC) and the intracellular volume fraction (ICVF). (c) Glial cells like astrocytes and microglia have a major role in the regulation of the brain tissue microenvironment. (d) Astrocytes (by their regulation of the membrane Aquaporin receptor expression) regulate the hydric environment. Water balance mechanisms could be altered or enhanced in the early stages of HD. Thus, the combination of neurite volume decreases and a potential increase in water content could explain the expansion of the extracellular compartment (EC), increasing the overall mean diffusion (MD) within the voxel, as well as the orientation diffusion index (ODI). Also, microglial cells, (one of the main cells involved in inflammatory processes) increases their secretion of cytokine mediators, also augmenting the tissue water permeability from the brain-blood barrier (BBB) at later stages of the disease. The combination of all previously mentioned effects (decrease in the IC, increase in EC, and small vessel permeability) may lead to an early and local (interstitial) raise in free-water content, as detected by an increase in the levels of isotropic volume fraction (IsoVF). Abbreviations; WT, wild type; R6/1, a transgenic mouse with Huntington's mutation HD, Huntington's Disease; FrA, frontal accessory area; Cing, Cingular cortical area; M1, primary motor area; M2, secondary motor area, CC, Corpus Callosum; STR, Striatum, HC hippocampus; Cereb., cerebellum; STN, o subthalamic Nucleus Th, thalamus; LV, lateral ventricle; IV, Fourth ventricular system, BBB, Brain blood Barrier; AQP; aquaporin receptor; Pre-HD; Presymptomatic stage of Huntington's disease; HD, Huntington's disease; GM, grey Matter; WM, white matter; IC, intracellular compartment; EC, extracellular compartment; GFAP, astrocyte marker; Iba1, microglial marker; NODDI, Neurite Orientation Dispersion and Density Imaging; ICVF, Intracellular Ventricular Fraction; ODI, Orientation Dispersion Index; IsoVF, Isovolumetric Fraction.

be required (Colon-Perez et al., 2019) and implemented in our future studies.

In terms of the initial diffusion parameters considered for the NODDI model, previous clinical studies explored the neurite compartment, setting the intrinsic parallel diffusivity d// to $1.7 \ \mu m^2 \ ms^{-1}$ evaluating the residuals concerning the parallel diffusivity in a range from 0.5 to $3.0 \ \mu m^2 \ ms^{-1}$ over tissue type and age (Guerrero et al., 2019). Interestingly, that study showed the optimum parallel diffusivity is significantly lower for GM relative to WM, and a significantly decrease in optimum parallel diffusivity in GM & WM in the infant group. Moreover, in recent clinical studies,

researchers have optimized the value of d// for the cerebral cortex to $1.1 \times 10^{-3} \text{ mm}^2/\text{s}$ (from the default value of $1.7 \times 10^{-3} \text{ mm}^2/\text{s}$) in the study of GM regions (Fukutomi et al., 2019). Although our studies used an empirical diffusivity value equal for GM and WM brain structures, further studies are necessary to define specific GM diffusivity values in murine brain tissues.

Conceptual limitations of the dMRI models must always be considered due to the complexity of the biological materials under analysis. Different mechanisms can compensate for one another and lead to misinterpretations of the underlying events. As an example, immune cell infiltration, which results in a decrease in the apparent diffusion coefficient (ADC) (Sevick et al., 1992), can be counteracted by the vasogenic edema induced by neuroinflammation (Schlaug et al., 1997). Thus, the use of DTI derived parameters may not be specific enough to selectively quantify restricted diffusion in inflammatory tissues, where several interrelated cellular events are taking place simultaneously (Glass et al., 2010; Komine and Yamanaka, 2015; Ransohoff et al., 2015; Sochocka et al., 2017). FA could potentially fall short in the GM assessment due to its isotropic nature and particularly important in the short distance exchange of information. Thus, it is possible that the collection of cell bodies and apical dendrites are not represented by most of the isotropic features (Gatto, R. G. et al., 2020). Therefore, the use of more complex models as non-Gaussian diffusion models, can improve our quantitative approach and interpretation of the microstructural events in preclinical models of HD (Gatto et al., 2015) and other NDDs (Gatto, 2020a; Gatto et al., 2019).

Even though DWI methods might not be specific enough to reflect all the biological changes occurring in HD, these methods can be influenced by axonal pathology as well as changes in myelin content.Advanced DWI models appear to improve correlations with histology (Albrecht et al., 2016). More and more aspects of neuroinflammation become accessible to probing by MRI/MRS, and an increasing number of studies are carried out taking advantage of the complementarity of the information provided by this technique (Quarantelli, 2015). The structural metrics obtained with these techniques are not direct measures of cell number or density and are likely to reflect a combination of many cellular processes. Nonetheless, these methods are sensitive to anatomical changes that can occur in the context of pathology such as atrophy, which has been linked to inflammation. MRI metrics are important for the characterization of neuroinflammatory disorders, and some have already proven their usefulness for diagnosis, and follow-up of patients affected by neuroinflammation (Albrecht et al., 2016). Advanced diffusion models, based on neurite orientation dispersion and density imaging (NODDI) in MRI, have proved to be sensitive to microglial density and the cellular changes associated with microglial activation in a preclinical setting (Yi et al., 2019). Astrocytes, the most abundant cell type in the brain, (Miller, 2018) show changes in neuroinflammation, but evidence derived by MRI on astrocytic activation are still confined to few groups, and often on limited samples (Cavaliere et al., 2020).

Overall, UHFD-MRI is an exceptional tool to study structural neurodegenerative features in preclinical HD models (Gatto et al., 2018a) (Weissmann et al., 2020) (Fig. 12). However, the link between multiple macrostructural, microstructural, and neurochemical events are complex and still not well-understood. Thus, our approach encourages the combination of multiple dMRI models and analytical histological methods. This holistic approach could improve the timing of potential therapeutic approaches.

5. Conclusions

Diffusion MRI can be used to extract information on early neuropathological events in a preclinical model of HD. Using a multiple dMRI technique approach, it is possible to non-invasively evaluate microstructural changes at early stages in different brain tissue compartments. Taken together, our studies showed that parameters from dMRI techniques, such as DTI and NODDI and complex histological analysis can lead to a better understanding of the disease. This study highlights the potential of UHFD-MRI to find early bioimaging markers, and use them to examine patients affected by HD.

Credit author statement

All authors had full access to all the data in the study and take responsibility for the integrity of the data and the accuracy of the data analysis. RG conceptualized and visualized the experiments. RG, CW, and MA collected the data, conducted the experiments, and analyzed the MRI and histological data. MA and DD assisted with the MRI data collection and curation. JS provided the biological samples for these studies. QA, LG, and LCSC processed the mouse brains. CW conducted the histological studies. RG and CW analyzed and curated the data. RG and CW wrote the manuscript. OU, TM, and RM added their opinions and validated the manuscript.

Conflict of Interest

The authors certify that they have no affiliations with or involvement in any organization or entity that may have had any financial or nonfinancial interest in the subject matter.

Data accessibility

The datasets generated and/or analyzed during the current study are available from either of the corresponding authors on reasonable request.

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Supplementary material

Supplementary material associated with this article can be found, in the online version, at doi:10.1016/j.neurobiolaging.2021. 02.006.

References

- Albrecht, D.S., Granziera, C., Hooker, J.M., Loggia, M.L., 2016. In vivo imaging of human neuroinflammation. ACS Chem. Neurosci. 7 (4), 470–483.
- Andersson, J.L., Skare, S., Ashburner, J., 2003. How to correct susceptibility distortions in spin-echo echo-planar images: application to diffusion tensor imaging. Neuroimage 20 (2), 870–888.
- Andersson, J.L.R., Sotiropoulos, S.N., 2016. An integrated approach to correction for off-resonance effects and subject movement in diffusion MR imaging. Neuroimage 125, 1063–1078.
- Beery, A.K., 2018. Inclusion of females does not increase variability in rodent research studies. Curr. Opin. Behav. Sci. 23, 143–149.
- Begeti, F., Schwab, L.C., Mason, S.L., Barker, R.A., 2016. Hippocampal dysfunction defines disease onset in Huntington's disease. J. Neurol. Neurosurg. Psychiatry 87 (9), 975–981.
- Benjamini, Y., 2010. Discovering the false discovery rate. J. R. Stat. Soc. Series B 72 (4), 405–416.
- Benjamini, Y., Hochberg, Y., 1995. Controlling the false discovery rate: a practical and powerful approach to multiple testing. J. R. Stat. Soc. Series B 57 (1), 289–300.

- Bondan, E.F., Cardoso, C.V., Martins, M.F.M., Otton, R., 2019. Memory impairments and increased GFAP expression in hippocampal astrocytes following hypercaloric diet in rats. Arq. Neuropsiquiatr. 77 (9), 601–608.
- Bradford, J., Shin, J.Y., Roberts, M., Wang, C.E., Sheng, G., Li, S., Li, X.J., 2010. Mutant huntingtin in glial cells exacerbates neurological symptoms of Huntington disease mice. J. Biol. Chem. 285 (14), 10653–10661.
- Castellanos, L.C.S., Rozenfeld, P., Gatto, R.G., Reisin, R.C., Uchitel, O.D., Weissmann, C., 2020. Upregulation of ASIC1a channels in an in vitro model of Fabry disease. Neurochem. Int. 140, 104824.
- Cavaliere, C., Tramontano, L., Fiorenza, D., Alfano, V., Aiello, M., Salvatore, M., 2020. Gliosis and neurodegenerative diseases: the role of PET and MR imaging. Front. Cell. Neurosci. 14 (75), 75.
- Cepeda-Prado, E., Popp, S., Khan, U., Stefanov, D., Rodriguez, J., Menalled, L.B., Dow-Edwards, D., Small, S.A., Moreno, H., 2012. R6/2 Huntington's disease mice develop early and progressive abnormal brain metabolism and seizures. J. Neurosci. 32 (19), 6456–6467.
- Chang, W.T., Puspitasari, F., Garcia-Miralles, M., Yeow, L.Y., Tay, H.C., Koh, K.B., Tan, L.J., Pouladi, M.A., Chuang, K.H., 2018. Connectomic imaging reveals Huntington-related pathological and pharmaceutical effects in a mouse model. NMR Biomed. 31 (12), e4007.
- Charan, J., Kantharia, N.D., 2013. How to calculate sample size in animal studies? J. Pharmacol. Pharmacother. 4 (4), 303–306.
- Cisbani, G., Cicchetti, F., 2012. An in vitro perspective on the molecular mechanisms underlying mutant huntingtin protein toxicity. Cell Death. Dis. 3, e382.
- Collins, T.J., 2007. ImageJ for microscopy. BioTechniques 43 (1 Suppl), 25-30.
- Colon-Perez, L.M., Ibanez, K.R., Suarez, M., Torroella, K., Acuna, K., Ofori, E., Levites, Y., Vaillancourt, D.E., Golde, T.E., Chakrabarty, P., Febo, M., 2019. Neurite orientation dispersion and density imaging reveals white matter and hippocampal microstructure changes produced by Interleukin-6 in the TgCRND8 mouse model of amyloidosis. Neuroimage 202, 116138.
- Cowan, C.M., Raymond, L.A., 2006. Selective neuronal degeneration in Huntington's disease. Curr. Top. Dev. Biol. 75, 25–71.
- Crombe, A., Planche, V., Raffard, G., Bourel, J., Dubourdieu, N., Panatier, A., Fukutomi, H., Dousset, V., Oliet, S., Hiba, B., Tourdias, T., 2018. Deciphering the microstructure of hippocampal subfields with in vivo DTI and NODDI: applications to experimental multiple sclerosis. Neuroimage 172, 357–368.
- D'Arceuil, H.E., Westmoreland, S., de Crespigny, A.J., 2007. An approach to high resolution diffusion tensor imaging in fixed primate brain. Neuroimage 35 (2), 553–565.
- Daly, J., Cohen, J., 1987. Statistical Power Analysis For The Behavioral Sciences Revised Edition.
- Dattalo, P., 2018. Determining sample size using fast and slow thinking. J. Soc. Serv. Res. 44 (2), 180–190.
- Dhital, B., Kellner, E., Kiselev, V.G., Reisert, M., 2018. The absence of restricted water pool in brain white matter. Neuroimage 182, 398–406.
- Di Leva, A., 2016. The Fractal Geometry of the Brain.
- Douaud, G., Behrens, T.E., Poupon, C., Cointepas, Y., Jbabdi, S., Gaura, V., Golestani, N., Krystkowiak, P., Verny, C., Damier, P., Bachoud-Levi, A.C., Hantraye, P., Remy, P., 2009. In vivo evidence for the selective subcortical degeneration in Huntington's disease. Neuroimage 46 (4), 958–966.
 Dyrby, T.B., Baare, W.F., Alexander, D.C., Jelsing, J., Garde, E., Sogaard, L.V., 2011. An
- Dyrby, T.B., Baare, W.F., Alexander, D.C., Jelsing, J., Garde, E., Sogaard, L.V., 2011. An ex vivo imaging pipeline for producing high-quality and high-resolution diffusion-weighted imaging datasets. Hum. Brain Mapp. 32 (4), 544–563.Dyrby, T.B., Innocenti, G.M., Bech, M., Lundell, H., 2018. Validation strategies for the
- Dyrby, T.B., Innocenti, G.M., Bech, M., Lundell, H., 2018. Validation strategies for the interpretation of microstructure imaging using diffusion MRI. Neuroimage 182, 62–79.
- Fazio, P., Paucar, M., Svenningsson, P., Varrone, A., 2018. Novel imaging biomarkers for Huntington's disease and other hereditary choreas. Curr. Neurol. Neurosci. Rep. 18 (12), 85.
- Ferrante, R.J., 2009. Mouse models of Huntington's disease and methodological considerations for therapeutic trials. Biochim. Biophys. Acta 1792 (6), 506–520.
- Figiel, M., Szlachcic, W.J., Switonski, P.M., Gabka, A., Krzyzosiak, W.J., 2012. Mouse models of polyglutamine diseases: review and data table. Part I. Mol. Neurobiol. 46 (2), 393–429.
- Finkbeiner, S., 2011. Huntington's disease. Cold Spring Harb. Perspect. Biol. 3 (6).
- Fukutomi, H., Glasser, M.F., Murata, K., Akasaka, T., Fujimoto, K., Yamamoto, T., Autio, J.A., Okada, T., Togashi, K., Zhang, H., Van Essen, D.C., Hayashi, T., 2019. Diffusion tensor model links to neurite orientation dispersion and density imaging at high b-value in cerebral cortical gray matter. Sci. Rep. 9 (1), 12246.
- Gatto, R.G., 2020a. Editorial for "evaluating the therapeutic effect of low-intensity transcranial ultrasound on traumatic brain injury with diffusion kurtosis imaging". J. Magn. Reson. Imaging: JMRI 52 (2), 532–533.
- Gatto, R.G., 2020b. Molecular and microstructural biomarkers of neuroplasticity in neurodegenerative disorders through preclinical and diffusion magnetic resonance imaging studies. J. Integr. Neurosci. 19 (3), 571–592.
- Gatto, R.G., Amin, M.Y., Deyoung, D., Hey, M., Mareci, T.H., Magin, R.L., 2018a. Ultra-high field diffusion mri reveals early axonal pathology in spinal cord of ALS mice. Transl. Neurodegen. 7, 20.
- Gatto, R.G., Chu, Y., Ye, A.Q., Price, S.D., Tavassoli, E., Buenaventura, A., Brady, S.T., Magin, R.L., Kordower, J.H., Morfini, G.A., 2015. Analysis of YFP(16)-R6/2 reporter mice and postmortem brains reveals early pathology and increased vulnerability of callosal axons in Huntington's disease. Hum. Mol. Genet. 24 (18), 5285–5298.
- Gatto, R.G., Li, W., Gao, J., Magin, R.L., 2018b. In vivo diffusion MRI detects early

spinal cord axonal pathology in a mouse model of amyotrophic lateral sclerosis. NMR Biomed. 31 (8), e3954.

- Gatto, R.G., Mustafi, S.M., Amin, M.Y., Mareci, T.H., Wu, Y.C., Magin, R.L., 2018c. Neurite orientation dispersion and density imaging can detect presymptomatic axonal degeneration in the spinal cord of ALS mice. Funct. Neurol. 33 (3), 155–163.
- Gatto, R.G., Weissmann, C., 2019. Diffusion Tensor Imaging in Preclinical and Human Studies of Huntington's Disease: What Have we Learned so Far? Curr. Med. Imaging Rev. 15 (6), 521–542.
- Gatto, R.G., Weissmann, C., Amin, M., Finkielsztein, A., Sumagin, R., Mareci, T.H., Uchitel, O.D., Magin, R.L., 2020. Assessing neuraxial microstructural changes in a transgenic mouse model of early-stage Amyotrophic Lateral Sclerosis by ultra-high field MRI and diffusion tensor metrics. Anim. Models Exp. Med. 3 (2), 117–129.
- Gatto, R.G., Ye, A.Q., Colon-Perez, L., Mareci, T.H., Lysakowski, A., Price, S.D., Brady, S.T., Karaman, M., Morfini, G., Magin, R.L., 2019. Detection of axonal degeneration in a mouse model of Huntington's disease: comparison between diffusion tensor imaging and anomalous diffusion metrics. MAGMA 32 (4), 461–471.
- Glass, C.K., Saijo, K., Winner, B., Marchetto, M.C., Gage, F.H., 2010. Mechanisms underlying inflammation in neurodegeneration. Cell 140 (6), 918–934.
- Guerrero, J.M., Adluru, N., Bendlin, B.B., Goldsmith, H.H., Schaefer, S.M., Davidson, R.J., Kecskemeti, S.R., Zhang, H., Alexander, A.L., 2019. Optimizing the intrinsic parallel diffusivity in NODDI: An extensive empirical evaluation. PLoS One 14 (9), e0217118.
- Gupta, R.K., Nath, K., Prasad, A., Prasad, K.N., Husain, M., Rathore, R.K., Husain, N., Srivastava, C., Khetan, P., Trivedi, R., Narayana, P.A., 2008. In vivo demonstration of neuroinflammatory molecule expression in brain abscess with diffusion tensor imaging. Am. J. Neuroradiol. 29 (2), 326–332.
- Gupta, R.K., Srivastava, S., Saksena, S., Rathore, R.K., Awasthi, R., Prasad, K.N., Husain, M., Pandey, C.M., Husain, N., 2010. Correlation of DTI metrics in the wall and cavity of brain abscess with histology and immunohistochemistry. NMR Biomed. 23 (3), 262–269.
- Gupta, R.K., Trivedi, R., Awasthi, R., Paliwal, V.K., Prasad, K.N., Rathore, R.K., 2012. Understanding changes in DTI metrics in patients with different stages of neurocysticercosis. Magn. Reson. Imaging 30 (1), 104–111.
- Halliday, G., Herrero, M.T., Murphy, K., McCann, H., Ros-Bernal, F., Barcia, C., Mori, H., Blesa, F.J., Obeso, J.A., 2009. No Lewy pathology in monkeys with over 10 years of severe MPTP Parkinsonism. Mov. Disord. 24 (10), 1519–1523.
- Hansson, O., Petersén, Å., Leist, M., Nicotera, P., Castilho, R.F., Brundin, P., 1999. Transgenic mice expressing a Huntington's disease mutation are resistant to quinolinic acid-induced striatal excitotoxicity. Proc. Natl Acad. Sci. 96 (15), 8727–8732.
- Harris, K.L., Armstrong, M., Swain, R., Erzinclioglu, S., Das, T., Burgess, N., Barker, R.A., Mason, S.L., 2019. Huntington's disease patients display progressive deficits in hippocampal-dependent cognition during a task of spatial memory. Cortex 119, 417–427.
- Heikkinen, T., Lehtimaki, K., Vartiainen, N., Puolivali, J., Hendricks, S.J., Glaser, J.R., Bradaia, A., Wadel, K., Touller, C., Kontkanen, O., Yrjanheikki, J.M., Buisson, B., Howland, D., Beaumont, V., Munoz-Sanjuan, I., Park, L.C., 2012. Characterization of neurophysiological and behavioral changes, MRI brain volumetry and 1H MRS in zQ175 knock-in mouse model of Huntington's disease. PLoS One 7 (12), e50717.
- Hobbs, N.Z., Barnes, J., Frost, C., Henley, S.M., Wild, E.J., Macdonald, K., Barker, R.A., Scahill, R.I., Fox, N.C., Tabrizi, S.J., 2010. Onset and progression of pathologic atrophy in Huntington disease: a longitudinal MR imaging study. Am. J. Neuroradiol. 31 (6), 1036–1041.
- Holz, M., Heil, S.R., Sacco, A., 2000. Temperature-dependent self-diffusion coefficients of water and six selected molecular liquids for calibration in accurate 1H NMR PFG measurements. PCCP 2 (20), 4740–4742.
- Jenkinson, M., Beckmann, C.F., Behrens, T.E., Woolrich, M.W., Smith, S.M., 2012. Fsl. Neuroimage 62 (2), 782–790.
- Kamiya, K., Hori, M., Aoki, S., 2020. NODDI in clinical research. J. Neurosci. Methods 346, 108908.
- Kapitulnik, A., Deutscher, G., 1984. Percolation scale effects in metal-insulator thin films. J. Stat. Phys. 36 (5-6), 815–826.
- Komine, O., Yamanaka, K., 2015. Neuroinflammation in motor neuron disease. Nagoya J. Med. Sci. 77 (4), 537–549.
- Li, J.Y., Popovic, N., Brundin, P., 2005. The use of the R6 transgenic mouse models of Huntington's disease in attempts to develop novel therapeutic strategies. NeuroRx 2 (3), 447–464.
- Magnotta, V.A., Kim, J., Koscik, T., Beglinger, L.J., Espinso, D., Langbehn, D., Nopoulos, P., Paulsen, J.S., 2009. Diffusion tensor imaging in preclinical Huntington's disease. Brain Imaging Behav. 3 (1), 77–84.
- Mancini, A., Gaetani, L., Di Gregorio, M., Tozzi, A., Ghiglieri, V., Calabresi, P., Di Filippo, M., 2017. Hippocampal neuroplasticity and inflammation: relevance for multiple sclerosis. Multiple Sclerosis Demyelinating Disord. 2 (1), 2.
- Miller, S.J., 2018. Astrocyte heterogeneity in the adult central nervous system. Front. Cell. Neurosci. 12 (401), 401.
- Muller, H.P., Roselli, F., Rasche, V., Kassubek, J., 2020. Diffusion tensor imaging-based studies at the group-level applied to animal models of neurodegenerative diseases. Front. Neurosci. 14, 734.
- Naver, B., Stub, C., Moller, M., Fenger, K., Hansen, A.K., Hasholt, L., Sorensen, S.A., 2003. Molecular and behavioral analysis of the R6/1 Huntington's disease transgenic mouse. Neuroscience 122 (4), 1049–1057.
- Palpagama, T.H., Waldvogel, H.J., Faull, R.L.M., Kwakowsky, A., 2019. The role of mi-

croglia and astrocytes in Huntington's disease. Front. Mol. Neurosci. 12 (258), 258.

- Paxinos, G., Franklin, K.B.J., 2004. The Mouse Brain in Stereotaxic Coordinates, Compact 2nd ed.. Elsevier Academic Press, Amsterdam; Boston.
- Phillips, O., Sanchez-Castaneda, C., Elifani, F., Maglione, V., Di Pardo, A., Caltagirone, C., Squitieri, F., Sabatini, U., Di Paola, M., 2013. Tractography of the corpus callosum in Huntington's disease. PLoS One 8 (9), e73280.
- Phillips, O., Squitieri, F., Sanchez-Castaneda, C., Elifani, F., Caltagirone, C., Sabatini, U., Di Paola, M., 2014. Deep white matter in Huntington's disease. PLoS One 9 (10), e109676.
- Pirici, D., Mogoanta, L., Margaritescu, O., Pirici, I., Tudorica, V., Coconu, M., 2009. Fractal analysis of astrocytes in stroke and dementia. Rom. J. Morphol. Embryol. 50 (3), 381–390.
- Politis, M., Lahiri, N., Niccolini, F., Su, P., Wu, K., Giannetti, P., Scahill, R.I., Turkheimer, F.E., Tabrizi, S.J., Piccini, P., 2015. Increased central microglial activation associated with peripheral cytokine levels in premanifest Huntington's disease gene carriers. Neurobiol. Dis. 83, 115–121.
- Quarantelli, M., 2015. MRI/MRS in neuroinflammation: methodology and applications. Clin. Transl. imaging 3 (6), 475–489.
- Quirion, J.G., Parsons, M.P., 2019. The onset and progression of hippocampal synaptic plasticity deficits in the Q175FDN mouse model of Huntington disease. Front. Cell. Neurosci. 13, 326.
- Ransohoff, R.M., Schafer, D., Vincent, A., Blachere, N.E., Bar-Or, A., 2015. Neuroinflammation: Ways in Which the Immune System Affects the Brain. Neurotherapeutics 12 (4), 896–909.
- Rattray, I., Smith, E.J., Crum, W.R., Walker, T.A., Gale, R., Bates, G.P., Modo, M., 2013. Correlations of behavioral deficits with brain pathology assessed through longitudinal MRI and histopathology in the R6/1 mouse model of Huntington's disease. PLoS One 8 (12), e84726.
- Reichenbach, A., Siegel, A., Senitz, D., Smith, T.G., 1992. A comparative fractal analysis of various mammalian astroglial cell types. Neuroimage 1 (1), 69–77.
- Rosas, H.D., Tuch, D.S., Hevelone, N.D., Zaleta, A.K., Vangel, M., Hersch, S.M., Salat, D.H., 2006. Diffusion tensor imaging in presymptomatic and early Huntington's disease: Selective white matter pathology and its relationship to clinical measures. Mov. Disord. 21 (9), 1317–1325.
- Rosenthal, R., Rosnow, R.L., 1984. Essentials of Behavioral Research: Methods and Data Analysis.
- Roze, E., Bonnet, C., Betuing, S., Caboche, J., 2010. Huntington's disease. Adv. Exp. Med. Biol. 685, 45–63.
- Ruocco, H.H., Lopes-Cendes, I., Li, L.M., Santos-Silva, M., Cendes, F., 2006. Striatal and extrastriatal atrophy in Huntington's disease and its relationship with length of the CAG repeat. Braz. J. Med. Biol. Res. 39 (8), 1129–1136.Schilling, K.G., Yeh, F.C., Nath, V., Hansen, C., Williams, O., Resnick, S., Ander-
- Schilling, K.G., Yeh, F.C., Nath, V., Hansen, C., Williams, O., Resnick, S., Anderson, A.W., Landman, B.A., 2019. A fiber coherence index for quality control of B-table orientation in diffusion MRI scans. Magn. Reson. Imaging 58, 82–89.
- Schlaug, G., Siewert, B., Benfield, A., Edelman, R.R., Warach, S., 1997. Time course of the apparent diffusion coefficient (ADC) abnormality in human stroke. Neurology 49 (1), 113–119.
- Schneider, C.A., Rasband, W.S., Eliceiri, K.W., 2012. NIH Image to ImageJ: 25 years of image analysis. Nat. Methods 9 (7), 671–675.
- Segovia, J., 2002. Transgenic Model for the Study of Oxidative Damage in Huntington's Disease. In: Sen, C.K., Packer, L. (Eds.), Redox Cell Biology and Genetics Part B. Academic Press, pp. 365–373.
- Sevick, R.J., Kanda, F., Mintorovitch, J., Arieff, A.I., Kucharczyk, J., Tsuruda, J.S., Norman, D., Moseley, M.E., 1992. Cytotoxic brain edema: assessment with diffusion-weighted MR imaging. Radiology 185 (3), 687–690.
- Shatil, A.S., Uddin, M.N., Matsuda, K.M., Figley, C.R., 2018. Quantitative ex vivo MRI changes due to progressive formalin fixation in whole human brain specimens: longitudinal characterization of diffusion, relaxometry, and myelin water fraction measurements at 3T. Front. Med. (Lausanne) 5, 31.
- Shim, J.H., Im, S.J., Kim, A.Y., Kim, Y.T., Kim, E.B., Baek, H.M., 2019. Generation of Mouse Basal Ganglia Diffusion Tractography Using 9.4T MRI. Exp. Neurobiol. 28 (2), 300–310.
- Shin, J.Y., Fang, Z.H., Yu, Z.X., Wang, C.E., Li, S.H., Li, X.J., 2005. Expression of mutant huntingtin in glial cells contributes to neuronal excitotoxicity. J. Cell Biol. 171 (6), 1001–1012.
- Sieradzan, K.A., Mann, D.M., 2001. The selective vulnerability of nerve cells in Huntington's disease. Neuropathol. Appl. Neurobiol. 27 (1), 1–21.

- Sochocka, M., Diniz, B.S., Leszek, J., 2017. Inflammatory response in the CNS: friend or foe? Mol. Neurobiol. 54 (10), 8071–8089.
- Steventon, J.J., Trueman, R.C., Ma, D., Yhnell, E., Bayram-Weston, Z., Modat, M., Cardoso, J., Ourselin, S., Lythgoe, M., Stewart, A., Rosser, A.E., Jones, D.K., 2016. Longitudinal in vivo MRI in a Huntington's disease mouse model: global atrophy in the absence of white matter microstructural damage. Sci. Rep. 6, 32423.
- Sun, S.W., Neil, J.J., Liang, H.F., He, Y.Y., Schmidt, R.E., Hsu, C.Y., Song, S.K., 2005. Formalin fixation alters water diffusion coefficient magnitude but not anisotropy in infarcted brain. Magn. Reson. Med. 53 (6), 1447–1451.
- Switonski, P.M., Szlachcic, W.J., Gabka, A., Krzyzosiak, W.J., Figiel, M., 2012. Mouse models of polyglutamine diseases in therapeutic approaches: review and data table. Part II. Mol. Neurobiol. 46 (2), 430–466.
- Ugolini, F., Lana, D., Nardiello, P., Nosi, D., Pantano, D., Casamenti, F., Giovannini, M.G., 2018. Different patterns of neurodegeneration and glia activation in CA1 and CA3 hippocampal regions of TgCRND8 mice. Front. Aging Neurosci. 10 (372), 372.
- Vonsattel, J.P., Myers, R.H., Stevens, T.J., Ferrante, R.J., Bird, E.D., Richardson Jr., E.P., 1985. Neuropathological classification of Huntington's disease. J. Neuropathol. Exp. Neurol. 44 (6), 559–577.
- Vorisek, I., Syka, M., Vargova, L., 2017. Brain Diffusivity and structural changes in the R6/2 mouse model of Huntington disease. J. Neurosci. Res. 95 (7), 1474–1484. Walker, F.O., 2007. Huntington's disease. Lancet 369 (9557), 218–228.
- Wang, N., Zhang, J., Cofer, G., Qi, Y., Anderson, R.J., White, L.E., Allan Johnson, G., 2019. Neurite orientation dispersion and density imaging of mouse brain microstructure. Brain Struct. Funct. 224 (5), 1797–1813.
- Weissmann, C., Amin, M., Angeles-López, A.D., García-Lara, L., Salinas Castellanos, L.C., Deyoung, D., Segovia, J., Mareci, T.H., Uchitel, O.D., Magin, R.L., Gatto, R.G., 2020. Early Changes in Structural Brain Bioimaging Markers from an R6/1 Mouse Model of Huntington's Disease Captured by Ultra-High Field MRI. Biomarkers of Paris (BMP), Paris (Virtual).
- Xiang, Z., Valenza, M., Cui, L., Leoni, V., Jeong, H.K., Brilli, E., Zhang, J., Peng, Q., Duan, W., Reeves, S.A., Cattaneo, E., Krainc, D., 2011. Peroxisome-proliferator-activated receptor gamma coactivator 1 alpha contributes to dysmyelination in experimental models of Huntington's disease. J. Neurosci. 31 (26), 9544–9553.
- Xiao, J., Hornburg, K.J., Cofer, G., Cook, J.J., Qi, Y., Johnson, G.A., 2020. A time-course study of actively stained mouse brains: DTI parameter and connectomic stability over one year. 2020.2012.2002.407338.
- Yeh, F.C., Verstynen, T.D., Wang, Y., Fernandez-Miranda, J.C., Tseng, W.Y., 2013. Deterministic diffusion fiber tracking improved by quantitative anisotropy. PLoS One 8 (11), e80713.
- Yi, S.Y., Barnett, B.R., Torres-Velazquez, M., Zhang, Y., Hurley, S.A., Rowley, P.A., Hernando, D., Yu, J.J., 2019. Detecting microglial density with quantitative multicompartment diffusion MRI. Front. Neurosci. 13 (81), 81.
- Young, K., Morrison, H., 2018. Quantifying Microglia Morphology from Photomicrographs of Immunohistochemistry Prepared Tissue using Image J. J. Vis. Exp. 136, 57648.
- Yushkevich, P.A., Piven, J., Hazlett, H.C., Smith, R.G., Ho, S., Gee, J.C., Gerig, G., 2006. User-guided 3D active contour segmentation of anatomical structures: significantly improved efficiency and reliability. Neuroimage 31 (3), 1116–1128.
- Yushkevich, P.A., Yang, G., Gerig, G., 2016. ITK-SNAP: an interactive tool for semi-automatic segmentation of multi-modality biomedical images. Annu. Int. Conf. IEEE Eng. Med. Biol. Soc. 2016, 3342–3345.
- Zhang, H., Schneider, T., Wheeler-Kingshott, C.A., Alexander, D.C., 2012. NODDI: practical in vivo neurite orientation dispersion and density imaging of the human brain. Neuroimage 61 (4), 1000–1016.
- Zheng, Z., Diamond, M.I., 2012. Huntington's disease and the huntingtin protein. Prog. Mol. Biol. Transl. Sci. 107, 189–214.