Title

Drumming motor sequence training induces apparent myelin remodelling in Huntington's disease: a longitudinal diffusion MRI and quantitative magnetization transfer study

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Key words

24   Huntington's disease, drumming training, white matter, myelin, diffusion MRI

List of abbreviations

ANOVA: analysis of variance

CC: corpus callosum

CHARMED: Composite Hindered and Restricted Model of Diffusion

CI: confidence intervals

DT MRI: diffusion tensor magnetic resonance imaging

DTI: diffusion tensor imaging

EPI: echo planar imaging

FA: fractional anisotropy

FAS: functional assessment score

FDR: false discovery rate

fODF: fiber orientation density function

FoV: field of view

FR: restricted fraction

FSPGR: fast spoiled gradient echo

FWE: family-wise error

GM: grey matter

HD: Huntington's disease

MPF: macromolecular proton fraction

MRI: magnetic resonance imaging

MT: magnetization transfer

MT-w: MT-weighted

PCA: principal component analysis

qMT: quantitative magnetization transfer

RD: radial diffusivity

ROI: region of interest

SD: standard deviation

SE: spin-echo

SMA: supplementary motor area

SPGR: spoiled gradient recalled-echo

T1-w: T1-weighted

TBSS: tract-based spatial statistics

TE: echo time

TMS: total motor score

TR: repetition time

UHDRS: Unified Huntington's disease rating scale

WM: white matter

Declarations

Ethics approval and consent to participate

66 The study was approved by the local National Health Service (NHS) Research Ethics

67 Committee (Wales REC 1 13/WA/0326) and all participants provided written informed

68 consent.

Availability of data and materials

70 The datasets analysed during the current study are available from the corresponding

71 author on reasonable request.

Competing interests

74 The authors declare that they have no competing interests.

Authors' contributions

CMB designed the study, obtained the funding which made the study possible; helped

with the set-up and completion of the study; contributed to the interpretation of results.

JBT was responsible for data collection, helped with MRI data processing, carried out

tractography of the tracts investigated, and helped with scoring the drumming

assessment. SB and GP helped with MRI data processing. CC processed some of the

data, performed statistical analysis on the data, interpreted results and wrote the

paper. GDP helped with the processing pipeline of the diffusion data. DKJ helped with

the interpretation of results. AR and EC helped with patients' recruitment and

undertook UHDRS assessments.

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1. Abstract

Background: Impaired myelination may contribute to Huntington's disease (HD) pathogenesis. This study assessed differences in white matter (WM) microstructure between HD patients and controls, and tested whether drumming training stimulates WM remodelling in HD. Furthermore, it examined whether training-induced microstructural changes are related to improvements in motor and cognitive function. **Methods**: Participants undertook two months of drumming exercises. Working memory and executive function were assessed before and after training. Changes in WM microstructure were investigated with diffusion tensor magnetic resonance imaging (DT-MRI)-based metrics, the restricted diffusion signal fraction (Fr) from the composite hindered and restricted model of diffusion (CHARMED) and the macromolecular proton fraction (MPF) from quantitative magnetization transfer (qMT) imaging. WM pathways linking the putamen and the supplementary motor area (SMA-Putamen), and three segments of the corpus callosum (CCI, CCII, CCIII) were studied using deterministic tractography. Baseline MPF differences between patients and controls were assessed with tract-based spatial statistics (TBSS). **Results:** MPF was reduced in HD patients compared to controls in the mid-section of the CC in HD subjects at baseline, while a significantly greater change in MPF was detected in HD patients relative to controls in the CCII, CCIII, and the right SMA-putamen post-training. Further, although patients improved their drumming and executive function performance**,** such improvements did not correlate with microstructural changes. **Conclusions:** Increased MPF suggests training-induced myelin changes in HD. Tailored behavioural stimulation may lead to neural benefits in early HD that could be exploited for delaying disease progression.

2. Introduction

2.1 The Pathology of Huntington's disease

Huntington's disease (HD) is a genetic, neurodegenerative disorder caused by an expansion of the CAG repeat within the *huntingtin* gene, leading to debilitating cognitive, psychiatric, and motor symptoms. In addition to striatal grey matter (GM) degeneration (Weaver et al., 2009), HD pathology has also been linked to white matter (WM) changes (Bardile et al., 2018; Bartzokis et al., 2007; Beglinger et al., 2007; Ciarmiello et al., 2006; Gregory et al., 2018; Paulsen et al., 2008; Rosas et al., 2018; Wang & Yang, 2019). WM is composed of axons and non-neuronal glia cells, such as myelin-producing oligodendrocytes, and it is unclear whether axons, myelin, or both are predominantly responsible for the WM abnormalities observed in HD (Gregory et al., 2018).

Some studies indicate that such changes are a result of Wallerian degeneration (Di Paola et al., 2014; Rosas et al., 2018; Weaver et al., 2009). However, an increasing body of research suggests that myelin-associated biological processes at the cellular and molecular level contribute to WM changes (Gómez-Tortosa et al., 2001; Huang et al., 2015; Jin et al., 2015; Myers et al., 1991; Simmons et al., 2007; Teo et al., 2016). Myelin is a multi-layered membrane sheath wrapping axons and is produced by oligodendrocytes. Axon myelination is vital during brain development and critical for healthy brain function, as it plays a fundamental role in the efficiency and speed of action potential propagation (Martenson, 1992). Oligodendrocyte/myelin dysfunction can slow down or stop otherwise fast axonal transport, which in turn can result in synaptic loss and eventually axonal degeneration (Han et al., 2010).

2.2 Interventions and Brain Plasticity

Currently, there are no disease-modifying treatments for HD. However, environmental stimulation and behavioural interventions may have the potential to reduce disease progression and delay disease onset (van Dellen et al., 2000; Yhnell et al., 2016). Converging evidence implicates myelin plasticity as one of the routes by which experience shapes brain structure and function (Caeyenberghs et al., 2018; Gibson et al., 2014; Lakhani et al., 2016; Mensch et al., 2015; Sampaio-Baptista et al., 2013; Scholz et al., 2009). Accordingly, plastic changes in myelination may be implicated in early adaptation and longer-term consolidation and improvement in motor tasks (Costa et al., 2004; Shmuelof & Krakauer, 2011; Steele et al., 2013; Yin et al., 2009). Changes in myelin-producing oligodendrocytes have been reported within the first hours of skill acquisition (Xiao et al., 2016); furthermore, alterations in grey and WM microstructure have been detected within 2 hours in people learning to play an action video game (Sagi et al., 2012), suggesting that experience can be quickly translated into adaptive changes in the brain.

This study assessed whether two months of drumming training could trigger WM microstructure changes, and potentially myelin remodelling, in individuals with HD. The present intervention was designed to exercise cognitive and motor functions including sequence and reversal learning, response speed and multi-tasking (Metzler-Baddeley et al., 2014), all of which rely on healthy functioning of cortico-basal ganglia loops that are known to be impaired in HD (Papoutsi et al., 2014). In brief, the intervention involves practising drumming patterns in ascending order of difficulty over a two-month period, and was previously found to induce WM changes in HD (Metzler-Baddeley et al. 2014). Based on reports of larger training-associated changes in

structural MRI metrics in patient populations than in healthy subjects (Caeyenberghs et al., 2018), we hypothesised that changes in *microstructural* metrics would be more marked in patients than in healthy subjects.

Previous WM plasticity neuroimaging studies (Giacosa et al., 2016; Scholz et al., 2009) have predominantly employed indices from diffusion tensor MRI (DT-MRI) (Pierpaoli & Basser, 1996). However, while sensitive, such measures are not specific to changes in specific sub-compartments of WM microstructure, challenging the interpretation of any observed change in DT-MRI indices (De Santis et al., 2014; Wheeler-Kingshott & Cercignani, 2009). To improve compartmental specificity beyond DT-MRI, the present study explored changes in the macromolecular proton fraction (MPF) from quantitative magnetization transfer (qMT) (Sled, 2018) imaging and the restricted diffusion signal fraction (Fr) from the composite hindered and restricted model of diffusion (CHARMED) (Assaf & Basser, 2005). Fractional anisotropy (FA) and radial diffusivity (RD) from DT-MRI (Pierpaoli & Basser, 1996) were included for comparability with previous training studies (Lövdén et al., 2010; Scholz et al., 2009; Zatorre et al., 2012).

The MPF has been proposed as a proxy MRI marker of myelin (Serres et al., 2009). Accordingly, histology studies show that this measure reflects demyelination accurately in Shiverer moice (Ou et al., 2009), is sensitive to de-myelination in multiple sclerosis patients (Levesque et al., 2010) and reflects WM myelin content in post-mortem studies of multiple sclerosis brains (Schmierer et al., 2007). Fr, on the other hand, represents the fraction of signal-attenuation that can be attributed to restricted diffusion, which is presumed to be predominantly intra-axonal, and therefore provides a proxy measure of axonal density (Barazany et al., 2009).

Training effects were investigated in WM pathways linking the putamen and the supplementary motor area (SMA-Putamen), and within three segments of the corpus callosum (CCI, CCII, CCIII). The SMA has efferent and afferent projections to the primary motor cortex and is involved in movement execution, and previous work has reported altered DT-MRI metrics in the putamen-motor tracts of symptomatic HD patients (Poudel et al., 2014). The anterior and anterior-mid sections of the corpus callosum contain fibres connecting the motor, premotor and supplementary motor areas in each hemisphere (Hofer & Frahm, 2006). Previous work has demonstrated a thinning of the corpus callosum in post-mortem HD brains (Vonsattel & Difiglia, 1998), altered callosal DT-MRI metrics in both pre-symptomatic and symptomatic HD patients (Rosas et al., 2010; Phillips et al., 2013), and a correlation between these metrics and performance on motor function tests (Dumas et al., 2012). Given previous reports of an effect of motor learning on myelin plasticity (Lakhani et al., 2016;), we expected changes following training to be more marked in MPF, as compared to the other non-myelin sensitive metrics assessed in this study. We also investigated the relationship between training-associated changes in MRI measures, and changes in drumming performance and cognitive/executive function. Finally, as previous evidence has shown widespread reductions in MPF in premanifest and manifest HD patients (Bourbon-Teles et al., 2017), we used tract-based spatial statistics (TBSS) (Smith et al., 2006) to investigate patient-control differences in MPF before training, across the whole brain.

3. Materials and Methods

3.1 Participants

The study was approved by the local National Health Service (NHS) Research

Ethics Committee (Wales REC 1 13/WA/0326) and all participants provided written

informed consent. All subjects were drumming novices and none had taken part in our

previously-reported pilot study (Metzler-Baddeley et al., 2014). Fifteen HD patients

were recruited from HD clinics in Cardiff and Bristol. Genetic testing confirmed the

presence of the mutant huntingtin allele. Thirteen age, sex, and education-matched

healthy controls were recruited from the School of Psychology community panel at

Cardiff University and from patients' spouses, carers or family members. The inclusion

criteria were the following: no history of head injury, stroke, cerebral haemorrhages or

any other neurological condition; eligible for MRI scanning; stable medication for at

least four weeks prior to the study.

Of the recruited sample, two patients were not MRI compatible, four withdrew

during the study and one patient's MRI data had to be excluded due to excessive

motion. Therefore, while drumming performance and cognitive data from 11 patients

were assessed, only 8 patients had a complete MRI dataset . One control participant

was excluded due to an incidental MRI finding, two participants dropped out of the

study and a fourth participant was not eligible for MRI. In total, we assessed drumming

and cognitive tests performance in 8 controls, while MRI data from nine controls were

available for analyses. Table 1 summarizes patients demographic and background

clinical characteristics. Most patients were at early disease stages, however two were

more advanced, as shown by their Total Motor Score (TMS; 69 and 40, respectively)

and Functional Assessment Score (FAS; 18 and 17, respectively). Table 2

summarizes demographic variables and performance in the Montreal Cognitive

Assessment (MoCA) (Nasreddine et al., 2005) and in the revised National Adult

Reading Test (NART-R) (Nelson, 1991) for patients and controls. While the groups did not differ significantly in age, controls were on average slightly older, performed significantly better on the MoCA, and had a significantly higher NART-IQ than patients.

Figure 1. White matter pathway regions of interest. Sagittal views of the reconstructed WM pathways displayed on a T$_1$-weighted image for one control participant. (A) CCI, CCII, and CCIII (Hofer and Frahm, 2006): CCI is the most anterior portion of the CC and maintains prefrontal connections between both hemispheres; CCII is the portion that maintains connections between premotor and supplementary motor areas of both hemispheres. CCIII maintains connections between primary motor cortices of both hemispheres. (B) SMA-putamen pathway: this pathway has efferent and afferent projections to the primary motor cortex and is involved in movement execution.

Table 1. Demographics and background clinical information of the patients' cohort. Based on TMS and FAS, most of the patients were at early disease stages, however two of them were more advanced.

	Age	Length of CAG repeats	TMS	FAS
Mean (Range)	48.5 (22-68)	43.6 (40-51)	18.7 (0-69)	22.6 (17-25)
SD	15.6	3.5	24.7	3.3

Abbreviations: CAG = cytosine-adenine-guanine; TMS = Total Motor Score out of 124 (the higher the scores the more impaired the performance); FAS = Functional Assessment Score out of 25 (the higher the scores the better the performance); SD = Standard Deviation.

3.2 Training intervention: Drumming-based rhythm exercises

The rhythm exercise and drumming training described previously in Metzler-Baddeley et al. (2014) was applied. Participants were provided with twenty-two 15 min training sessions on CDs, a pair of Bongo drums and a drumming diary and could practise at home. They were asked to exercise for 15 min per day, 5 times per week, for 2 months and to record the date and time of each exercise in their diary. Each training session introduced a drumming pattern based on one of the following rhythms: Brazilian samba, Spanish rumba, West-African kuku and Cuban son. After a brief warm up, trainees were encouraged to drum along with the instructor, initially with each hand separately and then with both hands alternating, starting with the dominant hand first and then reversing the order of the hands. The first exercises were based

on very simple, slow, and regular patterns but the level of complexity and speed
increased over the training sessions. Importantly, each individual progressed through
the training adaptively at their own pace i.e., as long as they exercised for the specified
time, they could repeat each session as often as they felt necessary to master it. To
maintain engagement and motivation, the training incorporated pieces of music based
on rhythms participants had learned and could drum along to. The researcher (JBT)
supervised the first training sessions and then remained in regular telephone contact
(at least once a week) with each participant throughout the intervention. Whenever
possible, carers and/or spouses were involved in the study to support the training.
Control participants started with Session 3 since the first two exercises were built on
a very low level of complexity, with slow, regular patterns of movement required, and
were therefore designed for patients.

3.3 Drumming assessment

Progress in drumming ability was assessed by digitally recording participants'
drumming performance for three patterns of ascending levels of difficulty (easy,
medium and hard), which were not part of the training sessions, at baseline and after
the training. Each recording was judged by an independent rater, blind to group and
time, according to an adopted version of the Trinity College London marking criteria
for percussion (2016) (www.trinitycollege.com; see Supplementary Figure 1 for
details).

3.4 Cognitive assessments

Different aspects of cognition and executive function were assessed before and after the training as described in Metzler-Baddeley et al. (2014). Multi-tasking was assessed with a dual task requiring simultaneous box crossing and digit sequences repetition (Baddeley, 1996). Attention switching was assessed with the trails test (VT) requiring the verbal generation of letter and digit sequences in alternate order relative to a baseline condition of generating letter or digit sequences only (Baddeley, 1996). Distractor suppression was tested with the Stroop task involving the naming of incongruent ink colours of colour words (Trenerry et al., 1989). Verbal and category fluency were tested using the letter cues "F", "A", "S" and "M", "C", "R" as well as the categories of "animals" and "boys' names" and "supermarket items" and "girls' names" respectively (Baldo et al., 2001). In total, we assessed 7 outcome variables, and percentage change scores in performance were computed for each of these variables (Table 3).

3.5 MRI data acquisition

MRI data were acquired on a 3 Tesla General Electric HDx MRI system (GE Medical Systems, Milwaukee) using an eight channel receive-only head RF coil at the Cardiff University Brain Research Imaging Centre (CUBRIC). The MRI protocol comprised the following images sequences: a high-resolution fast spoiled gradient echo (FSPGR) T_1-weighted (T_1-w) sequence for registration; a diffusion-weighted spin-echo echo-planar sequence (SE\EPI) with 60 uniformly distributed directions ($b = 1200$ s/mm^2), according to an optimized gradient vector scheme (Jones et al., 1999); a CHARMED acquisition with 45 gradient orientations distributed on 8 shells (maximum b-value = 8700s/mm^2) (Assaf and Basser, 2005); and a 3D MT-weighted

fast spoiled gradient recalled-echo (FSPGR) sequence (Cercignani and Alexander, 2006). The acquisition parameters of all scan sequences are reported in Table 4. Diffusion data acquisition was peripherally gated to the cardiac cycle. The off-resonance irradiation frequencies (Θ) and their corresponding saturation pulse amplitude (ΔSAT) for the 11 Magnetization transfer (MT) weighted images were optimized using Cramer-Rao lower bound optimization (Cercignani & Alexander, 2006).

Table 2. Demographics and general cognitive profile of patients and controls. *Both groups were matched for age, sex and years of education but the patient group had a lower* NART-IQ *and performed less well than the control group in the MoCA.*

Mean (SD, range)	Patients (n = 8)	Controls (n = 9)	Mann-Whitney U (p-value)
Age	48.5 (15.62, 22-68)	52.6 (14.56, 22-68)	U = 31 (p = 0.673)
NART-IQ	106.3 (13.13, 94-123)	121.22 (4.32, 117-128)	U = 8 (p = 0.006)
MoCa	23 (5.6, 14-29)	27.67 (1, 26-29)	U = 14 (p = 0.036)

Abbreviations: NART-IQ = verbal IQ estimate based on the National Adult Reading Test; MoCA = Montreal Cognitive Assessment score out of 30.

Table 3. Cognitive outcome variables assessed in this study. Tests were carried out before and after the training, and a percentage change score was computed for each variable.

Task	Outcome variables	Description
Simultaneous box crossing and digit sequences repetition (Baddeley, 1996)	Correct digits recalled under single task condition; correct digits recalled under dual task conditions; boxes identified under dual task condition.	Correct number of recalled digits in a standard digit span test; correct number of recalled digits in the dual condition; number of boxes identified in the dual condition
Stroop test (Trenerry et al., 1989)	Stroop interference score	Calculated by subtracting the number of errors from the total number of items presented in the test
Trials test (Baddeley, 1996)	Trail test switching	Performance accuracy: reflects the ability of moving flexibly from one set of rules to another in response to changing task requirements
Verbal and category fluency test (Delis et al., 2001)	Verbal and category fluency	Number of generated words starting with the following letters: "F", "A", "S" and "M", "C", "R"; number of generated words belonging to the following categories: "animals" and "boys'

names" and "supermarket items"
and "girls' names"

3.6 MRI data processing

The diffusion-weighted data were corrected for distortions induced by the diffusion-weighted gradients, artefacts due to head motion and EPI-induced geometrical distortions by registering each image volume to the T_1-w anatomical images (Irfanoglu et al., 2012), with appropriate reorientation of the encoding vectors (Alexander Leemans & Jones, 2009), all done in ExploreDTI (Version 4.8.3) (Leemans, Jeurissen, Sijbers, & Jones, 2009). A two-compartment model was fitted to derive maps of FA and RD in each voxel (Pasternak et al., 2009). CHARMED data were corrected for motion and distortion artefacts according to the extrapolation method of Ben-Amitay, Jones, and Assaf (2012). The number of distinct fiber populations (1, 2, or 3) in each voxel was obtained using a model selection approach (De Santis et al., 2014) and Fr was calculated per voxel with an in-house software (De Santis et al., 2014) coded in MATLAB (The MathWorks, Natick, MA). MT-weighted SPGR volumes for each participant were co-registered to the MT-volume with the most contrast using an affine (12 degrees of freedom, mutual information) registration to correct for inter-scan motion using Elastix (Klein et al., 2010). The 11 MT-weighted SPGR images and T_1 map were modelled by the two pool Ramani's pulsed MT approximation (Henkelman et al., 1993; Ramani et al., 2002), which included corrections for amplitude of B_0 field inhomogeneities. This approximation provided MPF maps, which were nonlinearly warped to the T_1-w images using the MT-volume

with the most contrast as a reference using Elastix (normalized mutual information

cost function) (Klein et al., 2010).

Table 4. Scan parameters. All sequences were acquired at 3T. For each of the
sequences, the main acquisition parameters are provided. T_1-w: T_1-weighted; MT-w:
MT-weighted; FSPGR: fast spoiled gradient echo; SE: spin-echo; EPI: echo-planar
imaging; SPGR: spoiled gradient recalled-echo; FoV: field of view; TE: echo time; TR:
repetition time.

	T_1-w	DTI	CHARMED	T1 map	MT-w	B0 map
Pulse sequence	FSPGR	SE\EPI	SE\EPI	SPGR (3D)	FSPGR (3D)	SPGR (3D)
Matrix size	256×256	96×96	96×96	96×96×60	96×96×60	128×128
FoV (mm)	230	230	230	240	240	220
Slices	172	60	60	-	-	-
Slice thickness (mm)	1	2.4	2.4	-	-	-
TE,TR (ms)	7.8, 2.9	87, 16000	126, 17000	6.85, 1.2	2.18,25.82	TE: 9 & 7 TR: 20

Off-resonance pulses (Hz/°)	-	-	-	-	1000/332, 1000/333, 12062/628, 47185/628, 56363/332, 2751/628, 1000/628, 1000/628, 2768/628, 2791/628, 2887/628	-
Flip angles (°)	20	90	90	15,7,3	5	90

3.7 Deterministic Tractography

Training-related changes in FA, RD, Fr, and MPF were quantified using a tractography approach in pathways interconnecting the putamen and the supplementary motor area bilaterally (SMA-Putamen), and within three segments of the corpus callosum (CCI, CCII and CCIII) (Hofer & Frahm, 2006) (Figure 1).

Whole brain tractography was performed for each participant in their native space using the damped Richardson-Lucy algorithm (Dell'acqua et al., 2010), which allows the recovery of multiple fiber orientations within each voxel including those affected by partial volume. The tracking algorithm estimated peaks in the fiber orientation density function (fODF) by selecting seed points at the vertices of a 2×2×2 mm grid superimposed over the image and propagated in 0.5-mm steps along these

axes re-estimating the fODF peaks at each new location (Jeurissen et al., 2011).

Tracks were terminated if the fODF threshold fell below 0.05 or the direction of

pathways changed through an angle greater than 45° between successive 0.5 mm

steps. This procedure was then repeated by tracking in the opposite direction from the

initial seed-points.

The WM tracts of interest were extracted from the whole-brain tractograms by

applying way-point regions of interest (ROI) (Catani et al., 2002). These were drawn

manually by one operator (JBT) blind to the identity of each dataset on color-coded

fiber orientation maps in native space guided by the following anatomical landmark

protocols (Figure 2).

3.7.1 Corpus callosum

Reconstruction of the CC segments followed the protocol of Hofer and Frahm

(Hofer & Frahm, 2006) as illustrated in Figure 2A. Segment reconstructions were

visually inspected and, if necessary, additional gates were placed to exclude

streamlines inconsistent with the known anatomy of the CC.

3.7.2 SMA-putamen pathway

One axial way-point ROI was placed around the putamen and one axial ROI

around the supplementary motor cortex (Leh et al., 2007) (Figure 2B). A way-point

gate to exclude fibres projecting to the brain stem was placed inferior to the putamen.

Figure 2. Sagittal views of the tractography protocols. *(A) CCI, CCII and CCIII (B)*
SMA - putamen pathway. Booleian logic OR waypoint regions of interest gates are
illustrated in blue; AND gates in green. M = Midline.

3.8 Statistical analyses

Statistical analyses were carried out in R Statistical Software (Foundation for
Statistical Computing, Vienna, Austria).

3.8.1 Assessment of training effects on drumming performance

Improvements in drumming performance were analysed with a two-way mixed
analysis of variance (ANOVA) testing for the effects of group (HD/controls), time of
assessment (before/after the training) and group by time interaction effects.
Furthermore, we compared our results to the ones obtained by running a robust mixed
ANOVA, using the bwtrim R function from the WRS2 package (Mair & Wilcox, 2020).
This implements robust methods for statistical estimation and therefore provides a
good option to deal with data presenting small sample sizes, skewed distributions and
outliers (Wilcox, 2011). Significant effects were further explored with post-hoc paired

and independent t-tests. The reliability of the post-hoc analyses was assessed with

bootstrap analysis based on 1000 samples and the 95% confidence interval (CI) of the

mean difference is provided for each significant comparison.

3.8.2 Assessment of group differences in the effect of training on cognitive

performance

Performance measures in executive function tasks have been shown to share

underlying cognitive structures (Testa et al., 2012). Therefore, PCA was employed to

reduce the complexity of the cognitive data and hence the problem of multiple

comparisons as well as to increase experimental power. PCA was run on change

scores for all participants across both groups. Due to the relatively small sample size,

we first confirmed with the Kaiser-Meyer-Olkin (KMO) test that our data was suited for

PCA. Subsequently, we followed guidelines to limit the number of extracted

components (Preacher & MacCallum, 2002; Winter, Dodou, & Wieringa, 2009), as

follows: first, we employed the Kaiser criterion of including all components with an

eigenvalue greater than 1; second, we inspected the Cattell scree plot (Cattell, 1966)

to identify the minimal number of components that accounted for most variability in the

data; third, we assessed each component's interpretability. A PCA procedure with

orthogonal Varimax rotation of the component matrix was used. Loadings that

exceeded a value of 0.5 were considered as significant.

Next, we assessed group differences in the component scores with permutation

analyses, to understand whether the training had differentially affected HD patients as

compared to controls. Significant group differences were tested using 5,000

permutations. Permutation testing relies only on minimal assumptions and can

therefore be applied when the assumptions of a parametric approach are untenable

such as in the case of small sample sizes. Multiple comparison correction was based

on a 5% false discovery rate (FDR) using the Benjamini-Hochberg procedure

(Benjamini & Hochberg, 1995).

3.8.3 Training effects on WM microstructure

Median measures of FA, RD, Fr and MPF were derived for each of the

reconstructed tracts in ExploreDTI (Leemans, Jeurissen, Sijbers, & Jones, 2009). A

percentage change score in these measures between baseline and post-training was

calculated in each tract (CCI, CCII, CCIII, left and right SMA-Putamen).

Previous research has shown that variation in the microstructural properties of

WM may represent a global effect, rather than being specific to individual tracts, and

that WM measures are highly correlated across WM areas (Lövdén et al., 2010; Penke

et al., 2010; Wahl et al., 2010). Therefore, we inspected the inter-tract correlation for

each of microstructural metric and found that MPF values were highly correlated,

whereas this was not true for the other metrics (Figure 3). Hence, percentage change

scores in MPF across the different tracts were transformed with PCA to extract

meaningful anatomical properties, following the procedure described above for the

PCA of cognitive change scores. PC scores for each participant were then used as

dependent variables in a permutation-based analysis using 5,000 permutations to

assess group differences in training associated changes in MPF. Finally, as a post-

hoc exploration, we looked for between-groups differences in MPF changes in the

individual tracts using 5000 permutations.

Training-associated changes in FA, Fr and RD were investigated with

permutation analyses separately for each tract. Significant group differences in these

measures were tested using 5,000 permutations. Multiple comparison correction was

based on a 5% FDR using the Benjamini-Hochberg procedure (Benjamini & Hochberg,

1995).

TBSS (Smith et al., 2006) was carried out to investigate baseline differences in

MPF between HD subjects and healthy controls. To produce significance maps, a

voxel-wise analysis was performed on the MPF projected 4D data for all voxels with

FA ≥ 0.20 to exclude peripheral tracts where significant inter-subject variability exists.

Inference based on permutations (5,000 permutations) and threshold-free-cluster-

enhancement was used. The significance level was set at $p < 0.05$ and corrected by

multiple comparisons (family-wise error, FWE).

***Figure 3.** Correlation matrices for the MRI metrics investigated across the different WM pathways. Colour intensity and the size of the circles are proportional to the strength of the correlation. * p < 0.05, ** p < 0.01, *** p < 0.001. The absolute correlation coefficient is plotted. MPF values were highly correlated across tracts, whereas this was not true for the other metrics*

<u>3.8.4 Relationship between changes in MRI measures and changes in drumming and</u>

<u>cognitive performance</u>

We computed percentage change scores for the drumming performance, in the

same way cognitive change scores were calculated. Scores were computed for the

easy test pattern in patients and for the medium test pattern in controls, as these

training patterns showed a significant improvement in the two groups, respectively.

Spearman correlation coefficients were calculated between drumming and cognitive

performance, and microstructural components that showed significant group

differences, to assess whether microstructural changes were related to any drumming

and/or cognitive benefits of the training.

4. Results

4.1 Training effects on drumming performance

The mixed ANOVA of drumming performance for the easy and medium test

pattern showed a significant effect of group [easy: $F(1,17) = 22.3$, $p < 0.001$; medium:

$F(1,17) = 13.1$, $p = 0.002$] and time [easy: $F(1,17) = 12.83$, $p = 0.004$; medium: $F(1,17)$

$= 13.4$, $p = 0.002$] but no interaction (easy: $p = 0.8$; medium: $p = 0.3$). For the hard test

pattern there was only a significant effect of group [$F(1,17) = 9.95$, $p = 0.006$] but not

of time ($p = 0.1$) and there was no interaction ($p = 0.4$). Results from the robust mixed

ANOVA were largely consistent with the above. Specifically, the easy and medium test

patterns showed a significant effect of group (easy: $p = 0.002$; medium: $p = 0.02$) and

time (easy: $p = 0.04$; medium: $p = 0.049$) but no interaction (easy: $p = 0.45$; medium:

$p = 0.69$). The hard test pattern showed a significant effect of group ($p = 0.02$) but not

of time ($p = 0.22$) and no interaction ($p = 0.8$). Figure 4 summarises the average

drumming performance per group and time point. Overall patients' drumming

performance was poorer than controls. Patients improved their drumming

performance significantly for the easy pattern [$t(10) = 2.7$, $p = 0.02$; 95% CI of mean

difference: $1.5 – 7.8$] and controls for the medium pattern [$t(7) = 3.8$, $p = 0.01$; 95% CI

of mean difference: $2.8 – 8.5$].

4.2 Group differences in the effect of training on cognitive performance

Three components, accounting for 79% of the variance in performance improvement in the cognitive benchmark tests, were extracted. The first component loaded highly on performance changes in the dual task (total number of boxes identified under dual task condition), the Stroop task (Stroop interference score), and the trails making task (Trail test switching). Since these variables all measure executive functions including focused attention and distractor suppression, the first component was labelled "executive" component. The second component loaded on variables reflecting the ability to correctly recall digits sequences (i.e. number of correct digits recalled under single and dual task condition) and was therefore labelled "working memory capacity" component. Finally, the third extracted component loaded highly on verbal and category fluency and was therefore labelled "fluency" component (Table 5).

We tested whether the two groups differed in terms of post-training cognition changes, by running permutation analyses on the individual scores for the three extracted components. The two groups differed in the executive component, $t = -1.03$, $p = 0.008$, FDR-corrected $p = 0.024$. However, no significant group differences were detected in the other two components (Working Memory capacity: $t = -0.22$, $p = 0.3296$, FDR-corrected $p = 0.3296$; Fluency: $t = -0.39$, $p = 0.242$ FDR corrected $p = 0.3296$).

4.3 Training effects on WM microstructure

Table 6 reports a summary of the training associated changes in FA, RD, Fr and MPF, across the different tracts.

Figure 4. Mean ratings for drumming performance *according to the Trinity College*
London marking criteria for percussion (2016) as a function of group and time point.
Patients improved their drumming performance significantly for the easy test pattern
*and controls for the medium difficult test pattern. * p < 0.05, ** p < 0.01, bootstrapping*
based on 1000 samples.

4.3.1 Training-associated group differences in FA

Permutation analyses of FA changes across the different tracts revealed no
significant differences between HD and control groups [CCI: t = 1.22, p = 0.91 (FDR-
corrected); CCII: t = 2.65, p = 0.91 (FDR-corrected); CCIII: t = 0.325, p = 0.13 (FDR-
corrected); right SMA-Putamen: t = -9.54, p = 0.10 (FDR-corrected); left SMA-
Putamen: t = 5.16, p = 0.77 (FDR-corrected).

<u>4.3.2 Training-associated group differences in RD</u>

There were no significant differences in RD changes following training

between HD patients and controls [CCI: t = -0.48, p = 0.45 (FDR-corrected); CCII: t =

-1.29, p = 0.45 (FDM-corrected); CCIII: t = -1.04, p = 0.45 (FDR-corrected); right SMA-

Putamen, t = 4.01 , p = 0.81 (FDR-corrected); left SMA-Putamen, t = -3.68, p = 0.39

(FDR-corrected).

<u>4.3.3 Training-associated group differences in Fr</u>

Permutation analyses of Fr changes across the different tracts revealed no

significant differences between HD and control groups [CCI: t = 3.39, p = 0.82 (FDR-

corrected; CCII: t = -0.17, p = 0.82 FDR-corrected; CCIII: t = 3.08, p = 0.82 (FDR-

corrected); right SMA-Putamen: t = -5.24, p = 0.82 (FDR-corrected); left SMA-

Putamen: t = 1.05, p = 0.82 (FDR-corrected)].

Table 5. Rotated Component Loadings on Change in the Cognitive Benchmark Tests.
Significant loadings (>0.5) are highlighted in bold.

% Change	Executive	Working memory capacity	Fluency
Total box (dual)	**0.864**	0.022	0.419
Stroop interference score	**0.811**	-0.270	-0.267
Trail test switching	**0.731**	-0.470	0.162

Correct digits under single task condition	0.201	**0.904**	0.129
Correct digits under dual task condition	-0.193	**0.855**	-0.018
Category fluency	-0.070	-0.138	**0.817**
Verbal fluency	-0.026	-0.232	**-0.799**

4.3.4 Training-associated group differences in MPF

PCA of change scores in MPF revealed one single component explaining 70.2% of the variance. This component presented high loadings from all the tracts investigated. A significant group difference was found for the MPF change-score component, indicating that HD patients presented significantly greater MPF changes in response to training, as compared to controls [t(14) = -1.743, n = 17, p = 0.03].

Finally, we found a significant difference in mean MPF change scores between the two groups for CCII [t(14) = -20.72, p=0.04], CCIII [t(14) = -25.87, p=0.04], and the right SMA-putamen pathway [t(14) = -25.48, p=0.04] after FDR correction, therefore indicating that there was a differential group effect of training on MPF within these tracts (Figure 5).

4.4 Relationship between training-associated changes in MRI measures, and changes in drumming and cognitive performance.

We did not find a significant association between the 'MPF' component scores and improvement in drumming performance (PC1: rho = -0.14, p > 0.05). Moreover, although no significant correlation was observed between the 'Executive' and the 'MPF' component scores, there was a positive trend (rho = .348, p = .171)

4.5 Baseline differences in MPF

Baseline MPF was reduced in the HD group compared to controls, in the midbody of the CC (t = 3.13, p = .05, FWE corrected). Figure 6 shows the areas with reduced MPF in HD patients, in blue.

Table 6. *Summary statistics for the permutation analysis of training effects on FA, RD, Fr and MPF, across the investigated tracts.*

FA	t	p	FDR corrected p
CCI	1.220	0.909	0.91
CCII	2.650	0.91	0.91
CCIII	0.320	0.48	0.91
Left SMA-Putamen	5.160	0.77	0.91
Right SMA-Putamen	-9.54	0.02	0.11

RD

CCI	-0.48	0.35	0.45
CCII	-1.29	0.22	0.45
CCIII	-1.04	0.30	0.45
Left SMA-Putamen	-3.68	0.08	0.39
Right SMA-Putamen	4.010	0.80	0.80

Fr

CCI	0.033	0.81	0.82
CCII	-0.001	0.49	0.82
CCIII	0.03	0.82	0.82
Left SMA-Putamen	0.01	0.58	0.82
Right SMA-Putamen	-0.052	0.1996	0.817

MPF

CCI	-12.06	0.08	0.10
CCII	-20.72	0.03	**0.05**
CCIII	-25.87	0.02	**0.05**
Left SMA-Putamen	-4.34	0.38	0.38

Right SMA-Putamen	-25.48	0.02	**0.05**

5. Discussion

Based on evidence that myelin impairment contributes to WM damage in HD (Bartzokis et al., 2007), and the suggestion that myelin plasticity underlies the learning of new motor skills (Lakhani et al., 2016; Scholz et al, 2009), the present study explored whether two months of drumming training would result in changes in WM microstructure in HD patients. Specifically, we expected to detect changes in MPF, as marker of WM myelin plasticity, in HD patients relative to healthy controls.

Firstly, we demonstrated a behavioural effect of the training by showing a significant improvement in drumming performance in patients (easy test pattern) and controls (medium test pattern). We did not detect any group differences in training-associated changes in the diffusion based indices of FA, RD and Fr. However, as hypothesised, we found a group difference in training-induced changes in the MPF PCA component. Specifically, HD patients showed significantly higher increases in MPF relative to controls. Furthermore, through exploratory post-hoc investigations, we detected significantly higher training-induced MPF changes within the CCII, CCIII and the right SMA-putamen pathway between patients and controls. Additionally, TBSS analysis of baseline differences in MPF suggested partial overlap of WM areas showing significant MPF reductions at baseline with areas showing changes post-training (i.e. CCII and CCIII).

MPF can be affected by inflammation (Henkelman, Stanisz, & Graham, 2001) and in advanced HD it is likely that inflammation goes hand-in-hand with myelin breakdown (Rocha et al., 2016). However, a recent CSF biomarker study found no evidence of neuro-inflammation in early-manifest HD (Vinther-Jensen et al., 2016). Furthermore, although recent evidence shows that this measure may be inconsistent when investigated in relatively small WM areas, presumably because of the effect of spatial heterogeneity in myelin thickness (Wang et al., 2020), we think that assessing within-subjects changes in MPF might have helped to address this limitation. Therefore, though preliminary and based on a small sample size, our findings suggest that two months of drumming and rhythm exercises may result in myelin remodelling in patients with early HD.

It is plausible that this group difference arose due to WM microstructural differences between patients and controls before the training. Accordingly, the HD group showed a significantly lower baseline MPF, consistent with lower myelin content (Bartzokis, 2007). Furthermore, previous studies have reported that training-associated percentage changes in MRI measures tend to be higher in patients than in healthy subjects (Caeyenberghs et al., 2018). One possibility is that in the healthy brain, neural networks may be optimally myelinated and further increasing myelin may not improve performance (Chomiak & Hu, 2009; Kaller et al., 2017; Rushton, 1951). Disentangling the impact of prior WM microstructural differences on microstructural plasticity during learning is beyond the scope of the current work.

Notably, the behavioral effect of drumming training and cognition differed between patients and controls. Patients improved in the easy drumming test pattern, and control improved in the medium test pattern. Furthermore, consistent with evidence from our pilot study (Metzler-Baddeley et al., 2014), patients showed

increases in the executive function components whilst control participants did not show

improvements in their cognition. Therefore, inter-group differences in microstructural

changes might not only be due to baseline WM microstructural differences, but also to

a different behavioral effect of the task between HD subjects and controls. For

instance, control participants performed close to ceiling in the easy test pattern, and

as the training was tailored to patients' needs, some of the earlier practice sessions

may not have optimally challenged them. The fact that the training seemed more

taxing for patients than controls may also explain why improvements in executive

functions and changes in MPF were only observed for the patients.

A critical question relevant to all training studies concerns the functional

significance of any observed neural changes. While we expected significant

relationships between changes in MRI measures and changes in drumming

proficiency or performance in cognitive tests, no significant correlations were found.

This might have been due to non-specific training-related neural responses.

Specifically, while the training exercise might have triggered changes in brain

structure, training-induced changes may not necessarily co-vary with improvements in

performance. Alternatively, it might be that our study was insufficiently powered to

detect brain-function correlations. The minimum sample size required to detect a

correlation was calculated to be 64 people ($\alpha = 0.05$; 80% power; medium effect size;

GPower 3 software). Therefore, our results need replication in larger samples.

Furthermore, lack of correlation between structural and functional changes after

training has been reported in other studies (including well-powered studies) and may

suggest that these processes follow different time courses and may occur in different

brain regions (Valkanova et al., 2014).

It is important to note that our study did not include a non-intervention patient

group. Within the 12 month time period of this study it was not possible to recruit a

sufficiently large number of well-matched patient controls. Therefore, we cannot

disentangle the effects of the training on WM microstructure from HD-associated

pathological changes. However, given that HD is a progressive neurodegenerative

disease associated with demyelination (Bartzokis et al., 2007), it is unlikely that

increases in MPF observed in the patient group were due to the disease itself. Finally,

while the majority of training studies assess brain structural changes between baseline

and post-training (Caeyenberghs et al., 2018), presumably on account of cost and

participant compliance, we suggest that acquiring intermittent scans during the training

period could have helped to better capture and understand changes in WM

microstructure observed in this study. Accordingly, future studies and more advanced

statistical analyses might be able to give greater insights into the complex nonlinear

relationships between structural changes and behaviour (Thomas & Baker, 2013).

To conclude, we have demonstrated that two months of drumming and rhythm

exercises result in a significantly greater change in a proxy MRI measure of myelin in

patients with HD relative to healthy controls. Whilst the current results require

replication in a larger patient group with an appropriately matched patient control

group, they suggest that behavioural stimulation may result in neural benefits in HD

that could be exploited for future therapeutics aiming to delay disease progression.

A.

B.

C.

D.

Figure 5. MPF changes scores: *PCA scree plot (A); plot summarising how each*
variable is accounted for in every principal component - colour intensity and the size
of the circles are proportional to the loading: PC1 loads on CCI, CCII, CCIII and right
SMA-Putamen, while PC2 loads mostly on the left SMA-Putamen; the absolute
correlation coefficient is plotted (B); correlation circle, interpreted as follows: 1)
positively correlated variables are grouped together, 2) negatively correlated variables
are positioned on opposite sides of the plot origin (opposite quadrants), 3) the distance
between variables and the origin measures the quality of the variable on the factor
map. Variables that are away from the origin are well represented on the factor map
(C); Bar graph of the percentage change in MPF across the inspected tracts; Error
bars represent the standard error; training was associated with a significantly greater
*change in MPF in CCII, CCIII, and right SMA-Putamen; * (p<0.05), results corrected*
for multiple comparisons with FDR (D).

A.

B.

Figure 6. TBSS analysis of baseline MPF values (A). Light blue areas show a significant reduction of MPF in patients with HD compared to controls (p < 0.05, FWE corrected). The midbody of the CC was mostly found to be affected, which carries connections to the premotor, supplementary motor and motor areas of the brain. Tracts showing significantly greater MPF changes in HD patients post-training as compared to controls (B). Areas showing significant MPF reductions at baseline overlap with tracts showing significant changes post-training (i.e. CCII and CCIII).

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