



Abnormalities of Brain White Matter in Type 2 Diabetes Mellitus: A Meta-Analysis of Diffusion Tensor Imaging

Li Huang^{1,2,3†}, Qingqing Zhang^{3†}, Tong Tang^{1,2,3}, Minguang Yang^{1,2,4}, Cong Chen^{1,2,4}, Jing Tao^{1,4} and Shengxiang Liang^{1,2,4*}

¹ National-Local Joint Engineering Research Center of Rehabilitation Medicine Technology, Fujian University of Traditional Chinese Medicine, Fuzhou, China, ² Rehabilitation Industry Institute, Fujian University of Traditional Chinese Medicine, Fuzhou, China, ³ College of Rehabilitation Medicine, Fujian University of Traditional Chinese Medicine, Fuzhou, China, ⁴ Traditional Chinese Medicine Rehabilitation Research Center of State Administration of Traditional Chinese Medicine, Fujian University of Traditional Chinese Medicine, Fuzhou, China

Aims: The study aimed to conduct a meta-analysis to determine the abnormalities of white matter in patients with type 2 diabetes mellitus (T2DM) by identifying the consistency of diffusion tensor imaging (DTI).

OPEN ACCESS

Edited by:

Ramesh Kandimalla, Indian Institute of Chemical Technology (CSIR), India

Reviewed by:

Michael Malek-Ahmadi, Banner Alzheimer's Institute, United States Wenbin Li, Sichuan University, China Venkata Ajay Narendra Talabattula, Delaware State University, United States

> *Correspondence: Shengxiang Liang sxliang@fjtcm.edu.cn

[†]These authors have contributed equally to this work

> Received: 12 April 2021 Accepted: 05 July 2021 Published: 06 August 2021

Citation:

Huang L, Zhang Q, Tang T, Yang M, Chen C, Tao J and Liang S (2021) Abnormalities of Brain White Matter in Type 2 Diabetes Mellitus: A Meta-Analysis of Diffusion Tensor Imaging. Front. Aging Neurosci. 13:693890.

Front. Aging Neurosci. 13:693890. doi: 10.3389/fnagi.2021.693890 **Method:** The literature for DTI comparing patients with T2DM with controls published before October 30, 2020, were reviewed in PubMed, Web of Science, Embase, CNKI, and Wan Fang databases. The meta-analysis was performed using the activation likelihood estimation (ALE) method, including 12 reports and 381 patients with T2DM.

Results: The meta-analysis identified 10 white matter regions that showed a consistent reduction of fractional anisotropy (FA) in patients with T2DM, including genu of the corpus callosum, the body of corpus callosum, bilateral anterior corona radiata, bilateral superior corona radiata, bilateral cingulum, and bilateral superior fronto-occipital fasciculus.

Conclusion: This study revealed the abnormal characteristics of white matter in T2DM, which would be helpful to understand the underlying neuropathological and physiological mechanisms of T2DM and provide evidence for clinical diagnosis and treatment.

Keywords: type 2 diabetes mellitus, diffusion tensor imaging, white matter, meta-analysis, activation likelihood estimation

INTRODUCTION

Type 2 diabetes mellitus (T2DM) is a chronic metabolic disease in which blood glucose is abnormal due to insufficient insulin secretion or the inability of the body to effectively use insulin (Arnold et al., 2018). According to the International Diabetes Federation (IDF), the incidence of type 2 diabetes worldwide is estimated to be 8.18%, with 415 million adults living with the disease and the number still on the rise (Zheng et al., 2018). T2DM is recognized as an independent risk factor for cognitive impairment (Biessels et al., 2006; Cukierman-Yaffe et al., 2020). The risk of developing Alzheimer's disease (AD) in patients with T2DM is two times as high as that in healthy individuals (Peila et al., 2002; Crane et al., 2013). Studies have shown that cognitive decline is related to abnormal brain microstructures (Alfaro et al., 2018; Reas et al., 2020).

Clarifying these brain abnormalities is helpful for the clinical diagnosis and treatment of cognitive decline in T2DM.

White matter makes up about 50% of the volume in the human brain (Filley, 1998). In the human forebrain, there are about 135,000 km of myelinated fibers (Saver, 2006) connecting the gray matter regions. Information processing in gray matter, manifested by synaptic events such as neurotransmitter release and long-term enhancement, is supplemented by information transmission in white matter, both of which are essential for the highly evolved behavior of the human brain (Filley, 2010). However, several studies have found abnormalities in the white matter of T2DM (Gao et al., 2019; Zhuo et al., 2019; Cui et al., 2020).

As a highly sensitive method for detecting the microstructure of white matter, diffusion tensor imaging (DTI) is an important technique for assessing white matter integrity in vivo (Catani, 2006; Mori et al., 2009). Based on the diffusion velocity of water molecules in different tissues, white matter integrity is usually estimated via fractional anisotropy (FA), and diffusion coefficient parameters are calculated by DTI. Previous DTI studies have observed widespread white matter abnormalities in patients with type 2 diabetes, but the results of these studies were mixed. For example, a reduction in FA was found mainly in the left temporal lobe (Yau et al., 2009), as opposed to a reduction in the right temporal lobe (Xiong et al., 2019b). Other researchers have also found a decrease in FA in the frontal lobes and corpus callosum (Ruan et al., 2017; Xie et al., 2017; Su et al., 2020). Differences in the results of these studies might be due to the limited sample sizes, different acquisition parameters, and analysis methods. Several systematic reviews have summarized the brain microstructural alterations in T2DM using DTI, which suggests the detrimental effects of T2DM on cognitive functions that might be associated with the alterations in brain microstructure (Sanjari Moghaddam et al., 2019; Alotaibi et al., 2021). Although the specific abnormal regions of white matter in T2DM were summarized in the systematic reviews, the locations and sizes of the alterations were not reported. Therefore, it is urgently needed to conduct a meta-analysis to identify the locations of the alterations in white matter in patients with T2DM.

Activation likelihood estimation (ALE) is a brain imaging meta-analysis method based on coordinates, which integrates many brain imaging results to obtain locations of the stable differences in brain regions through automated statistical analysis (Turkeltaub et al., 2002; Laird et al., 2005; Eickhoff et al., 2009, 2012). To date, ALE has been widely used in neuroimaging studies of neurological and psychological disorders (Tahmasian et al., 2017; Xia et al., 2017; Roberts et al., 2020). To identify the consistency of white matter changes in T2DM, we applied ALE to conduct a meta-analysis of DTI in patients with T2DM.

MATERIALS AND METHODS

Search Strategy

A meta-analysis of the literature on alterations in the white matter of T2DM was performed according to PRISMA guidelines (Moher et al., 2015). All the articles published before October 30, 2020, were retrieved from PubMed, Web of Science, Embase, CNKI, and Wan Fang database. Two researchers independently used the combined keywords to complete the search: (1) "White matter," "Diffusion tensor imaging," or "fractional anisotropy," and (2) "Type 2 diabetes mellitus" or "T2DM." The FA differences were estimated in most studies *via* voxel-based analysis (VBA) or tract-based special statistics (TBSS), while the comparison of other diffusion measures was very limited, such as mean diffusivity (MD), radial diffusivity (RD), and axial diffusivity (AD). We comprehensively searched the studies using MD, RD, or AD for DTI analysis in T2DM. There was one study that estimated MD and one that estimated RD which met the inclusion criteria. Therefore, this study conducted an imaging meta-analysis to investigate the FA differences.

Study Inclusion and Exclusion Criteria

Studies were included according to the criteria: (1) the original research articles were published in peer-reviewed journals, (2) patients with T2DM were diagnosed, (3) significant differences in FA values were examined in the study, and (4) the coordinates in Montreal Neurological Institute (MNI) or Talairach (Tal) space were reported.

Studies were excluded in the following situations: (1) the study included patients with undiagnosed T2DM in the experimental group, (2) coordinates were not reported in the study or were not obtained after contacting the first author, (3) the results were limited to a particular region of interest, not the whole brain, and (4) the articles or non-research articles were unpublished.

Literature Quality Assessment

The literature quality assessment form was developed based on the previous study (Iwabuchi et al., 2015). The total score of the form was 20, with two parts of sample information (10 points) and imaging methods (10 points).

Sample information was scored as follows: (1) patients were diagnosed with specified standardized diagnostic criteria (1 point), (2) age and educational background were reported with mean (or median) and SDs (or range) (2 points), (3) healthy comparison subjects were evaluated to exclude psychiatric and medical illnesses, and demographic data were reported (1 point), (4) important clinical variables [e.g., illness duration, fasting plasma glucose (FPG) level, HbA1c, and cognitive level] were reported with mean (or median) and SDs (or range) (4 points), and (5) sample size per group was ≥ 10 (2 points).

Evaluation of imaging methods was scored in the following aspects: (1) whole-brain analysis was automated without *a priori* regional selection (3 points), (2) magnet strength was at least 1.5 T (1 point), (3) the study had at least six of diffusion direction (1 point), (4) whole-brain coverage of scans was used (1 point), (5) the acquisition and preprocessing techniques were clearly described so that they could be reproduced (1 point), (6) coordinates were reported in standard space (1 point), (7) significant results were reported after correction for multiple testing using a standard statistical procedure [false discovery rate (FDR), family-wise error (FWE), or permutation-based methods] (1 point), and (8) conclusions were consistent with the results obtained and the limitations were discussed (1 point).

Data Extraction

Information extraction from the included study was done independently by two researchers. We extracted the baseline information, DTI characteristics, and coordinates in the study. Baseline information included the name of the first author, publication year, study group, and sample size of each group, age, years of education, FPG, HbA1c, disease course, body mass index (BMI), and cognitive level. DTI characteristics included the following information: magnet, image processing, analytical method, coordinates type, direction, b values, and statistical threshold.

Coordinate-Based Meta-Analysis

To analyze the consistency of abnormal FA clusters reported in all studies, a meta-analysis of ALE was performed using the GingerALE2.3.3 software package (Laird et al., 2005; Eickhoff et al., 2009; Research Imaging Institute of the University of Texas Health Science Center, San Antonio, TX, United States). The voxel coordinates of each study were regarded as probability distributions to create ALE distribution maps (Turkeltaub et al., 2002). The x, y, and z peak activation coordinates of all the white matter fiber clusters were included as the input for the meta-analysis. FA measures from all included studies were used in the meta-analysis. The analysis was performed in the MNI space. The ALE meta-analysis was estimated using a cluster-level inference threshold of P < 0.05 (FWE correction) with 5,000 permutations and P < 0.05. Then, the Johns Hopkins University (JHU) DTI-based white-matter atlases were applied to label the resulting clusters. Besides, Colin brain template in MNI space was used to visualize the results using DPABI software (Yan et al., 2016).

RESULTS

Search Results and Study Characteristics

Two researchers independently searched the literature according to the search terms. A total of 2,100 articles were retrieved using



the search strategy (CNKI n = 327; Web of Science n = 638; Wan Fang n = 448; PubMed n = 307; Embase n = 380). About 871 duplicate papers were deleted, 291 theses, 13 conference papers, and 5 books were excluded according to literature types, and 920 journal articles were selected. The titles or abstracts of journal papers were screened according to the inclusion and exclusion criteria, and 48 journal articles were selected to meet the criteria. Two researchers independently reviewed the full text of 48 articles. About 36 articles were excluded for the following reasons: region of interest, coordinates not reported, imaging rather than DTI, fiber tracking, axial diffusion, animal, and full text not found. Finally, a total of 12 studies were included for the meta-analysis (Yau et al., 2009, 2010, 2013; Chen et al., 2013; Kim et al., 2016; Nouwen et al., 2017; Ruan et al., 2017; Xie et al., 2017; Yoon et al., 2017; Liang et al., 2019; Xiong et al., 2019b; Su et al., 2020; Figure 1).

The T2DM group and the control group in the 12 included studies were usually described by their characteristics, such as age, gender, education, FPG, HbA1C, duration, BMI, and cognitive level (**Table 1**). In the included studies, one study (Kim et al., 2016) showed a higher level of HbA1C than the others. The DTI parameters and analysis methods in the studies were used to examine magnetic field intensity, toolbox, analysis method, coordinate space, the number of diffusion direction, b0-value, and statistical threshold (**Table 2**). Literature quality assessments with a total score of 20 were conducted. All the scores of the 12 studies were higher than 16, which indicated that the quality of the included studies was acceptable (**Figure 2**).

Meta-Analysis

Activation likelihood estimation meta-analysis was performed on the FA values of 381 patients with T2DM from 12 included studies. Compared to controls, patients with T2DM had shown consistent FA reductions (**Figure 3**; **Table 3**). The largest cluster (4,432 mm³) was located in the body of corpus callosum, and the others included left superior corona radiata (1,944 mm³), right superior corona radiata (1,400 mm³), right anterior corona radiata (1,336 mm³), genu of corpus callosum (448 mm³), right cingulum (120 mm³), left anterior corona radiata (88 mm³), left cingulum (24 mm³), left superior frontooccipital fasciculus (16 mm³), and right superior fronto-occipital fasciculus (8 mm³).

DISCUSSION

In the present study, a coordinate-based meta-analysis of DTI was conducted to investigate the abnormal white matter in patients with T2DM. Twelve eligible studies with a total of 381 patients with T2DM were analyzed. The consistent abnormal white matters were identified, namely, the commissural fibers genu of corpus callosum and body of corpus callosum, the association fibers cingulum and superior fronto-occipital fasciculus, and the projection fibers anterior corona radiata and superior corona radiata.

Commissural fibers connected the left and right hemispheres, such as corpus callosum, anterior commissure, and fornix. As

the largest commissural fiber in the telencephalon, the corpus callosum was an important connective fiber that played a key role in transmitting, integrating, and coordinating information between the left and right hemispheres, whether excitatory or inhibitory (Bloom and Hynd, 2005; Caillé et al., 2005; Roland et al., 2017). Consistent with previous reports (Chen et al., 2013; Yau et al., 2013; Yoon et al., 2017), abnormalities in the corpus callosum were also found in patients with T2DM in this study. Furthermore, based on the discovery of abnormalities in the corpus callosum, we made a more detailed localization and found abnormalities in the genu and body of the corpus callosum. The body of the corpus callosum, which connects the parietal, temporal, and occipital lobes, contains important structures for processing memory, emotions, and execution (Musiek, 1986; Matsukawa et al., 2011; Goldstein et al., 2021). Damage to the body of the corpus callosum would result in dysfunction or abnormal memory, mood, and executive function (Peltier et al., 2012). Impairment in the body of the corpus callosum might play an important role in cognitive impairment in T2DM. Besides, the genu of the corpus callosum contains fibers that connect the left and right prefrontal lobes, the premotor cortex, and the accessory motor cortex (Matsukawa et al., 2011). Abnormalities in the genu and body of the corpus callosum may contribute to sensory, cognitive, mental, or visual impairments in patients with T2DM (Yang et al., 2014). Therefore, these regions may become new targets for the treatment of patients with T2DM who also have cognitive impairment.

Projection fibers are the connecting fibers between the cerebral cortex and the subcortical center. The anterior corona contains projection fibers from the internal capsule to the cerebral cortex (Wakana et al., 2004), which primarily projects to the prefrontal cortex and plays a role in the neural circuitry for emotion regulation (Sanjuan et al., 2013; Goodkind et al., 2015). Consistent with the results of this meta-analysis, several studies on DTI have found that the anterior corona radiata and superior corona radiata were abnormal in patients with T2DM (Xiong et al., 2016, 2019a; Sun et al., 2018). Patients with T2DM are often accompanied by a variety of emotional disorders, such as depression and anxiety. For example, a crosssectional study found that the prevalence of depression and anxiety in patients with type 2 diabetes in western Saudi Arabia was 33.8 and 38.3%, respectively (Alzahrani et al., 2019). It was reported that the prevalence of anxiety in people with T2DM was significantly higher than in those who were healthy (Tu et al., 2017). At the same time, a meta-analysis showed that almost a quarter of patients with T2DM were depressed (Khaledi et al., 2019). Further studies are needed to explore whether the occurrence of depression and anxiety in patients with T2DM may be related to white matter abnormalities such as the corona radiata.

The abnormal connecting fibers that connect the adjacent gyrus of the ipsilateral cerebral hemisphere were found in T2DM, including the cingulum and superior fronto-occipital fasciculus. Consistent with the previous studies, significant abnormalities in the cingulum of patients with T2DM were found in the study (Liang et al., 2019; Cui et al., 2020). As a core member of the limbic system, the cingulate serves as a bridge

References	Groups	Sample	Age (years)	Education (years)	FPG (mmol/L)	HbA1C (%)	Duration (years)	BMI (kg/m²)	Full Scale IQ	MMSE	MoCA
Yau et al. (2009)	T2DM	24	57.21 ± 8.05	15.29 ± 2.76	NA	7.83 ± 1.88	7.94 ± 5.64	32.13 ± 5.96	106.74 ± 13.00	NA	NA
	Control	17	56.44 ± 6.94	16.06 ± 1.75	NA	5.37 ± 0.42	_	24.09 ± 3.69	112.78 ± 12.28	NA	NA
Yau et al. (2010)	T2DM	18	16.46 ± 1.89	10.75 ± 1.53	8.37 ± 4.79	8.32 ± 2.87	NA	37.70 ± 6.36	87.83 ± 12.55	IQ MMSE 3.00 NA 2.28 NA 2.55 NA 1.75 NA NA <td>NA</td>	NA
	Control	18	17.16 ± 1.45	11.15 ± 1.66	4.16 ± 0.43	5.29 ± 0.33	_	36.80 ± 7.22	103.63 ± 11.75		NA
Yau et al. (2013)	HTN/T2DM	22	58.36 ± 7.91	NA	8.01 ± 3.26	7.89 ± 1.90	NA	32.89 ± 7.03	NA	NA	NA
	HTN/non-T2DM	11	62.38 ± 7.32	NA	4.40 ± 0.62	5.64 ± 0.19	_	25.48 ± 5.08	NA	NA	NA
Chen et al. (2013)	T2DM	16	61.00 ± 12.00	NA	NA	NA	13.00 ± 4.00	NA	NA	NA	NA
	Control	16	60.00 ± 8.00	NA	NA	NA	_	NA	NA	NA	NA
Kim et al. (2016)	T2DM	20	54.60 ± 2.30	11.90 ± 2.30	10.00 ± 1.04	10.70 ± 0.30	12.10 ± 6.50	24.70 ± 0.60	NA	NA	NA
	Control	20	54.30 ± 2.40	10.00 ± 3.60	5.19 ± 0.13	5.90 ± 0.10	_	23.60 ± 0.40	NA	NA	NA
Nouwen et al. (2017)	T2DM	13	16.00 ± 1.60	NA	8.87 ± 3.87	7.80 ± 1.97	2.57 ± 1.92	NA	NA	NA	NA
	Control	20	16.10 ± 1.90	NA	4.78 ± 0.49	5.29 ± 0.33	_	NA	NA	NA	NA
Xie et al. (2017)	T2DM	58	56.09 ± 8.16	11.72 ± 3.31	8.06 ± 2.81	8.35 ± 2.10	7.60 ± 5.82	25.57 ± 2.19	NA	29.21 ± 0.89	NA
	Control	58	54.66 ± 7.03	11.07 ± 2.64	5.13 ± 0.65	5.56 ± 0.33	_	24.64 ± 2.99	NA	NA 29.21 \pm 0.89 29.50 \pm 0.94	NA
Yoon et al. (2017)	T2DM	100	49.20 ± 7.70	NA	7.67 ± 2.24	7.12 ± 1.43	1.83 ± 1.53	25.50 ± 3.40	NA	NA	NA
	Control	50	49.00 ± 7.80	NA	5.24 ± 0.21	5.29 ± 0.15	_	22.70 ± 1.80	NA	NA	NA
Ruan et al. (2017)	T2DM	30	55.83 ± 6.00	7.10 ± 1.30	12.39 ± 2.44	NA	8.77 ± 2.46	NA	NA	NA	24.77 ± 1.04
	Control	30	56.33 ± 5.27	7.23 ± 1.33	4.88 ± 0.53	NA	_	NA	NA	NA	28.47 ± 1.31
Liang et al. (2019)	T2DM	34	58.29 ± 4.19	10.50 ± 1.21	6.64 ± 2.00	7.85 ± 1.34	6.85(3-14)	24.36 ± 1.89	NA	NA	NA
	Control	32	56.31 ± 4.46	9.75 ± 1.32	5.28 ± 0.23	NA	_	20.58 ± 1.47	NA	NA	NA
Xiong et al. (2019b)	T2DM-MCI	20	63.55 ± 5.81	11.30 ± 3.61	9.39 ± 2.16	8.15 ± 1.63	9.09 ± 8.14	24.37 ± 3.32	NA	25.40 ± 2.09	25.25 ± 1.25
	Control	28	59.65 ± 5.98	10.18 ± 3.07	5.19 ± 0.68	5.23 ± 0.38	_	24.03 ± 2.02	NA	28.36 ± 0.99	28.64 ± 1.22
Su et al. (2020)	T2DM	26	55.58 ± 5.93	14.54 ± 1.61	9.25 ± 3.38	8.13 ± 1.76	11.12 ± 5.56	23.97 ± 2.22	NA	28.69 ± 1.29	27.50 ± 1.24
	Control	25	53.12 ± 4.79	15.56 ± 2.00	5.31 ± 0.85	5.58 ± 0.51	_	24.00 ± 3.45	NA	28.56 ± 1.42	27.54 ± 1.10

TABLE 1 | Study and subject information of the studies included in the current meta-analysis.

Data are mean ± SD. BMI, body mass index; FPG, fasting plasma glucose; HTN, hypertension; MCI, mild cognitive impairment; MMSE, mini-mental state examination; MoCA, montreal cognitive assessment; T2DM, type 2 diabetes mellitus.

TABLE 2 | Diffusion tensor imaging (DTI) parameters and analysis methods in the studies of the meta-analysis.

References	Magnet	Image processing	Analytical method	Coordinates	Direction	b0 (s/mm²)	Statistical threshold
Yau et al. (2009)	1.5T	MIDAS, ART2	VBA	Tal	6	1,000	P < 0.005
Yau et al. (2010)	1.5T	SPM	VBA	Tal	6	1,000	<i>P</i> < 0.005
Yau et al. (2013)	NA	ART2	VBA	Tal	6	1,000	<i>P</i> < 0.01
Chen et al. (2013)	3.0T	SPM	VBA	MNI	15	1,000	<i>P</i> < 0.01
Kim et al. (2016)	3.0T	FSL	TBSS	MNI	30	1,000	P < 0.05 (FWE)
Nouwen et al. (2017)	3.0T	FSL	TBSS	MNI	61	1,500	P < 0.05 (TFCE)
Xie et al. (2017)	3.0T	SPM, FSL, DKE	VBA	MNI	25	1,000, 2,000	P < 0.05 (AlphaSim)
Yoon et al. (2017)	1.5T	FSL	TBSS	MNI	54	1,000	P < 0.05 (Cluster-based thresholding)
Ruan et al. (2017)	3.0T	SPM, FSL	VBA	MNI	64	1,000	P < 0.05 (corrected for multiple comparisons)
Liang et al. (2019)	3.0T	PANDA	VBA	MNI	25	1,000	P < 0.05 (AlphaSim)
Xiong et al. (2019b)	3.0T	FSL	TBSS	MNI	25	1,250, 2,500	P < 0.05 (FWE)
Su et al. (2020)	3.0T	FSL	TBSS	MNI	64	1,000	P < 0.05 (TFCE)

ART2, automated registration toolkit 2; DKE, diffusional kurtosis estimator; FA, fractional anisotropy; FWE, family-wise error; FSL, FMRIB software library; MIDAS, multi-modal imaging data analysis system; MNI, Montreal neurological Institute; PANDA, pipeline for analyzing brain diffusion images; SPM, statistical parametric mapping; Tal, talairach; TBSS, tract based spatial statistics; TFCE, threshold-free cluster enhancement; VBA, voxel-based analysis.



clearly described in the study. The red circle means the information is not described in the study.

connecting the various lobes of the brain and the cingulate gyrus (Dalgleish, 2004; Bubb et al., 2018). The anterior cingulate cortex is related to cognitive control and decision-making, and

the posterior cingulate cortex is involved in the adjustment of working memory, visual space, and spatial orientation. In particular, the anterior cingulate cortex plays a key role in



FIGURE 3 Abnormal white matter estimated by fractional anisotropy (FA) in type 2 diabetes mellitus (T2DM). (A) The results from the activation likelihood estimation (ALE) meta-analyses are shown in 3D rendering. The gray region represents the outline of the brain. The green region represents the normal white matter. The blue region represents the abnormalities in the white matter. (B) The results from the ALE meta-analyses are overlaid onto a standard template in the Montreal Neurological Institute (MNI) space. The color bar represents the ALE value. Cluster-level inference threshold of P < 0.05 [family-wise error (FWE) correction] with 5,000 permutations and P < 0.05.

TABLE 3 | Regions that showed consistent fractional anisotropy (FA) reduction in patients with T2DM.

No.	ROI	х	Y	Z	Extrema	Volume (mm ³)
1	Genu of corpus callosum	16	20	24	0.01434	448
2	Body of corpus callosum	18	12	32	0.015065	4,432
3	Anterior corona radiata R	18	16	32	0.013598	1,336
4	Anterior corona radiata L	-16	16	30	0.0034261	88
5	Superior corona radiata R	18	12	34	0.017433	1,400
6	Superior corona radiata L	-22	-8	36	0.0094876	1,944
7	Cingulum R	10	20	24	0.0042591	120
8	Cingulum L	-6	-2	32	0.0025344	24
9	Superior fronto-occipital fasciculus R	22	14	22	0.0021478	8
10	Superior fronto-occipital fasciculus L	-20	10	24	0.002947	16

error handling (Leech and Sharp, 2014; Bliss et al., 2016; Maldonado et al., 2020). Cingulate impairment in neurological and psychiatric disorders has been widely reported. Based on these studies, we speculate that the abnormal cingulate may be related to cognitive decline in T2DM. In addition, some studies have previously reported the abnormality of the postdefault mode network connection in patients with T2DM (Cui et al., 2015; Ishibashi et al., 2018; Liu et al., 2019), and the cingulate as the structural basis of this default network (van den Heuvel et al., 2008). The findings in this study may help to explain these network changes. Interestingly, abnormalities in the fronto-occipital fasciculus were also found in the study, which is consistent with the reduction of FA in bilateral frontal occipital tracts of T2DM reported by Xiong et al. (2019a). The fronto-occipital fasciculus played an important role in visual processing and spatial awareness (Bar et al., 2006; Meola et al., 2015). Therefore, the visual impairment in patients with T2DM might not only be due to retinal atherosclerosis but may also be related to abnormalities of the cingulate and frontooccipital fasciculus.

This meta-analysis has determined the abnormalities of white matter in patients with T2DM by identifying the consistency of DTI. However, some limitations need to be pointed out. First, the number of included studies was relatively small. Second, few studies have reported the other indicators via DTI; therefore, only FA was used to estimate the abnormalities in white matter. Further studies based on MD, AD, and RD should be conducted to comprehensively assess the abnormal white matter in T2DM. Third, on the one hand, the MRI machine, acquisition parameters, and analysis method in each of the studies were different; on the other hand, few studies share the original data because it is too large. It is difficult to estimate the heterogeneous using FA images. Further studies should be done to establish a shared neuroimaging dataset about T2DM, which would make it possible to do the meta-analysis based on the original images.

In summary, the meta-analysis of DTI demonstrated the abnormalities in commissural fibers, association fibers, and projection fibers in T2DM, which might contribute to the neurobehavioral disorders in T2DM. These findings would promote the understanding of the neuropathological mechanisms of T2DM.

DATA AVAILABILITY STATEMENT

The original contributions presented in the study are included in the article/supplementary material, further inquiries can be directed to the corresponding author/s.

AUTHOR CONTRIBUTIONS

SL and JT designed the whole study. LH, TT, MY, and CC searched and selected the studies, analyzed the data, prepared figures, and drafted the article. LH and SL undertook the

REFERENCES

- Alfaro, F. J., Gavrieli, A., Saade-Lemus, P., Lioutas, V. A., Upadhyay, J., and Novak, V. (2018). White matter microstructure and cognitive decline in metabolic syndrome: a review of diffusion tensor imaging. *Metab. Clin. Exp.* 78, 52–68. doi: 10.1016/j.metabol.2017. 08.009
- Alotaibi, A., Tench, C., Stevenson, R., Felmban, G., Altokhis, A., Aldhebaib, A., et al. (2021). Investigating brain microstructural alterations in type 1 and type 2 diabetes using diffusion tensor imaging: a systematic review. *Brain Sci.* 11:140. doi: 10.3390/brainsci11020140
- Alzahrani, A., Alghamdi, A., Alqarni, T., Alshareef, R., and Alzahrani, A. (2019). Prevalence and predictors of depression, anxiety, and stress symptoms among patients with type ii diabetes attending primary healthcare centers in the western region of saudi arabia: a cross-sectional study. *Int. J. Ment. Health Syst.* 13:48. doi: 10.1186/s13033-019-0307-6
- Arnold, S. E., Arvanitakis, Z., Macauley-Rambach, S. L., Koenig, A. M., Wang, H. Y., Ahima, R. S., et al. (2018). Brain insulin resistance in type 2 diabetes and alzheimer disease: concepts and conundrums. *Nat. Rev. Neurol.* 14, 168–181. doi: 10.1038/nrneurol.2017.185
- Bar, M., Kassam, K. S., Ghuman, A. S., Boshyan, J., Schmid, A. M., Dale, A. M., et al. (2006). Top-down facilitation of visual recognition. *Proc. Natl. Acad. Sci.* U.S.A. 103, 449–454. doi: 10.1073/pnas.0507062103
- Biessels, G. J., Staekenborg, S., Brunner, E., Brayne, C., and Scheltens, P. (2006). Risk of dementia in diabetes mellitus: a systematic review. *Lancet Neurol.* 5, 64–74. doi: 10.1016/S1474-4422(05)70284-2
- Bliss, T. V., Collingridge, G. L., Kaang, B. K., and Zhuo, M. (2016). Synaptic plasticity in the anterior cingulate cortex in acute and chronic pain. *Nat. Rev. Neurosci.* 17, 485–496. doi: 10.1038/nrn.2016.68
- Bloom, J. S., and Hynd, G. W. (2005). The role of the corpus callosum in interhemispheric transfer of information: excitation or inhibition? *Neuropsychol. Rev.* 15, 59–71. doi: 10.1007/s11065-005-6252-y
- Bubb, E. J., Metzler-Baddeley, C., and Aggleton, J. P. (2018). The cingulum bundle: anatomy, function, and dysfunction. *Neurosci. Biobehav. Rev.* 92, 104–127. doi: 10.1016/j.neubiorev.2018.05.008
- Caillé, S., Sauerwein, H. C., Schiavetto, A., Villemure, J. G., and Lassonde, M. (2005). Sensory and motor interhemispheric integration after section of different portions of the anterior corpus callosum in nonepileptic patients. *Neurosurgery* 57, 50–9; discussion: 50–59. doi: 10.1227/01.NEU.00001 63089.31657.08
- Catani, M. (2006). Diffusion tensor magnetic resonance imaging tractography in cognitive disorders. *Curr. Opin. Neurol.* 19, 599–606. doi: 10.1097/01.wco.0000247610.44106.3f
- Chen, Z., Li, J., Liu, M., and Ma, L. (2013). Voxel-based analysis of diffusion tensor indices in the brain of the patients with type ii diabetes mellitus. *Chin. J. Radiol.* 47, 811–815. doi: 10.3760/cma.j.issn.1005-1201.2013. 09.012

statistical analysis. LH, QZ, CC, and SL participated in the interpretation of data. LH and QZ wrote the manuscript. SL revised the manuscript. All authors read and approved the final manuscript.

FUNDING

This work was supported by the Key Research and Development Project funded by the Ministry of Science and Technology of the People's Republic of China (2019YFC1710301) and Scientific Research Foundation for the High-level Talents funded by Fujian University of Traditional Chinese Medicine (X2019002-talents).

- Crane, P. K., Walker, R., Hubbard, R. A., Li, G., Nathan, D., Zheng, H., et al. (2013). Glucose levels and risk of dementia. *N. Engl. J. Med.* 369, 540–548. doi: 10.1056/NEJMoa1215740
- Cui, Y., Jiao, Y., Chen, H. J., Ding, J., Luo, B., Peng, C. Y., et al. (2015). Aberrant functional connectivity of default-mode network in type 2 diabetes patients. *Eur. Radiol.* 25, 3238–3246. doi: 10.1007/s00330-015-3746-8
- Cui, Y., Tang, T. Y., Lu, C. Q., Cai, Y., Lu, T., Wang, Y. C., et al. (2020). Abnormal cingulum bundle induced by type 2 diabetes mellitus: a diffusion tensor tractography study. *Front. Aging Neurosci.* 12:594198. doi:10.3389/fnagi.2020.594198
- Cukierman-Yaffe, T., Gerstein, H. C., Colhoun, H. M., Diaz, R., García-Pérez, L. E., Lakshmanan, M., et al. (2020). Effect of dulaglutide on cognitive impairment in type 2 diabetes: an exploratory analysis of the rewind trial. *Lancet Neurol.* 19, 582–590. doi: 10.1016/S1474-4422(20)30173-3
- Dalgleish, T. (2004). The emotional brain. Nat. Rev. Neurosci. 5, 583-589. doi: 10.1038/nrn1432
- Eickhoff, S. B., Bzdok, D., Laird, A. R., Kurth, F., and Fox, P. (2012). Activation likelihood estimation meta-analysis revisited. *Neuroimage* 59, 2349–2361. doi: 10.1016/j.neuroimage.2011.09.017
- Eickhoff, S. B., Laird, A. R., Grefkes, C., Wang, L. E., Zilles, K., and Fox, P. T. (2009). Coordinate-based activation likelihood estimation meta-analysis of neuroimaging data: a random-effects approach based on empirical estimates of spatial uncertainty. *Hum. Brain Mapp.* 30, 2907–2926. doi: 10.1002/hbm.20718
- Filley, C. M. (1998). The behavioral neurology of cerebral white matter. *Neurology* 50, 1535–1540. doi: 10.1212/WNL.50.6.1535
- Filley, C. M. (2010). White matter: organization and functional relevance. *Neuropsychol. Rev.* 20, 158–173. doi: 10.1007/s11065-010-9127-9
- Gao, S., Chen, Y., Sang, F., Yang, Y., Xia, J., Li, X., et al. (2019). White matter microstructural change contributes to worse cognitive function in patients with type 2 diabetes. *Diabetes* 68, 2085–2094. doi: 10.2337/db19-0233
- Goldstein, A., Covington, B. P., Mahabadi, N., and Mesfin, F. B. (2021). "Neuroanatomy, corpus callosum," in *StatPearls [Internet]* (Treasure Island, FL: StatPearls Publishing).
- Goodkind, M., Eickhoff, S. B., Oathes, D. J., Jiang, Y., Chang, A., Jones-Hagata, L. B., et al. (2015). Identification of a common neurobiological substrate for mental illness. *JAMA Psychiatry* 72, 305–315. doi: 10.1001/jamapsychiatry.2014.2206
- Ishibashi, K., Sakurai, K., Shimoji, K., Tokumaru, A. M., and Ishii, K. (2018). Altered functional connectivity of the default mode network by glucose loading in young, healthy participants. *BMC Neurosci.* 19:33. doi: 10.1186/s12868-018-0433-0
- Iwabuchi, S. J., Krishnadas, R., Li, C., Auer, D. P., Radua, J., and Palaniyappan, L. (2015). Localized connectivity in depression: a meta-analysis of resting state functional imaging studies. *Neurosci. Biobehav. Rev.* 51, 77–86. doi: 10.1016/j.neubiorev.2015.01.006
- Khaledi, M., Haghighatdoost, F., Feizi, A., and Aminorroaya, A. (2019). The prevalence of comorbid depression in patients with type 2 diabetes: an updated

systematic review and meta-analysis on huge number of observational studies. *Acta Diabetol.* 56, 631–650. doi: 10.1007/s00592-019-01295-9

- Kim, D. J., Yu, J. H., Shin, M. S., and Shin, Y. (2016). Hyperglycemia reduces efficiency of brain networks in subjects with type 2 diabetes. *PLoS ONE* 11:e0157268. doi: 10.1371/journal.pone.0157268
- Laird, A. R., Fox, P. M., Price, C. J., Glahn, D., Uecker, A. M., Lancaster, J. L., et al. (2005). ALE meta-analysis: controlling the false discovery rate and performing statistical contrasts. *Hum. Brain Mapp.* 25, 155–164. doi: 10.1002/hbm.20136
- Leech, R., and Sharp, D. J. (2014). The role of the posterior cingulate cortex in cognition and disease. *Brain* 137(Pt 1), 12–32. doi: 10.1093/brain/awt162
- Liang, Y., Zhang, H., Tan, X., Liu, J., Qin, C., Zeng, H., et al. (2019). Local diffusion homogeneity provides supplementary information in T2dmrelated Wm microstructural abnormality detection. *Front. Neurosci.* 13:63. doi: 10.3389/fnins.2019.00063
- Liu, H., Liu, J., Peng, L., Feng, Z., Cao, L., Liu, H., et al. (2019). Changes in default mode network connectivity in different glucose metabolism status and diabetes duration. *Neuroimage Clin* 21:101629. doi: 10.1016/j.nicl.2018.101629
- Maldonado, I. L., Parente de Matos, V., Castro Cuesta, T. A., Herbet, G., and Destrieux, C. (2020). The human cingulum: from the limbic tract to the connectionist paradigm. *Neuropsychologia* 144:107487. doi: 10.1016/j.neuropsychologia.2020.107487
- Matsukawa, H., Shinoda, M., Fujii, M., Takahashi, O., Yamamoto, D., Murakata, A., et al. (2011). Genu of corpus callosum as a prognostic factor in diffuse axonal injury. J. Neurosurg. 115, 1019–1024. doi: 10.3171/2011.6.JNS11513
- Meola, A., Comert, A., Yeh, F. C., Stefaneanu, L., and Fernandez-Miranda, J. C. (2015). The controversial existence of the human superior fronto-occipital fasciculus: connectome-based tractographic study with microdissection validation. *Hum. Brain Mapp.* 36, 4964–4971. doi: 10.1002/hbm.22990
- Moher, D., Shamseer, L., Clarke, M., Ghersi, D., Liberati, A., Petticrew, M., et al. (2015). Preferred reporting items for systematic review and meta-analysis protocols (Prisma-P) 2015 statement. Syst. Rev. 4:1. doi: 10.1186/2046-4053-4-1
- Mori, S., Oishi, K., and Faria, A. V. (2009). White matter atlases based on diffusion tensor imaging. *Curr. Opin. Neurol.* 22, 362–369. doi: 10.1097/WCO.0b013e32832d954b
- Musiek, F. E. (1986). Neuroanatomy, neurophysiology, and central auditory assessment. Part III: corpus callosum and efferent pathways. *Ear Hear.* 7, 349–358. doi: 10.1097/00003446-198612000-00001
- Nouwen, A., Chambers, A., Chechlacz, M., Higgs, S., Blissett, J., Barrett, T., et al. (2017). Microstructural abnormalities in white and gray matter in obese adolescents with and without type 2 diabetes. *Neuroimage Clin.* 16, 43–51. doi: 10.1016/j.nicl.2017.07.004
- Peila, R., Rodriguez, B. L., and Launer, L. J. (2002). Type 2 diabetes, apoe gene, and the risk for dementia and related pathologies: the Honolulu-Asia aging study. *Diabetes* 51, 1256–1262. doi: 10.2337/diabetes.51.4.1256
- Peltier, J., Roussel, M., Gerard, Y., Lassonde, M., Deramond, H., Le Gars, D., et al. (2012). Functional consequences of a section of the anterior part of the body of the corpus callosum: evidence from an interhemispheric transcallosal approach. J. Neurol. 259, 1860–1867. doi: 10.1007/s00415-012-6421-x
- Reas, E. T., Hagler, D. J., Kuperman, J. M., Wierenga, C. E., Galasko, D., White, N. S., et al. (2020). Associations between microstructure, amyloid, and cognition in amnestic mild cognitive impairment and dementia. *J. Alzheimers. Dis.* 73, 347–357. doi: 10.3233/JAD-190871
- Roberts, C. A., Giesbrecht, T., Fallon, N., Thomas, A., Mela, D. J., and Kirkham, T. C. (2020). A systematic review and activation likelihood estimation metaanalysis of fMRI studies on sweet taste in humans. *J. Nutr.* 150, 1619–1630. doi: 10.1093/jn/nxaa071
- Roland, J. L., Snyder, A. Z., Hacker, C. D., Mitra, A., Shimony, J. S., Limbrick, D. D., et al. (2017). On the role of the corpus callosum in interhemispheric functional connectivity in humans. *Proc. Natl. Acad. Sci. U.S.A.* 114, 13278–13283. doi: 10.1073/pnas.1707050114
- Ruan, X., Chen, W., and Wang, J. (2017). Evaluation of microstructure damage in brain of patients with type 2 diabetes mellitus and its correlation with cognitive function by Dti. *Chin. J. Disast. Med.* 5, 446–449. doi: 10.13919/j.issn.2095-6274.2017.08.006
- Sanjari Moghaddam, H., Ghazi Sherbaf, F., and Aarabi, M. H. (2019). Brain microstructural abnormalities in type 2 diabetes mellitus: a systematic review of diffusion tensor imaging studies. *Front. Neuroendocrinol.* 55:100782. doi: 10.1016/j.yfrne.2019.100782

- Sanjuan, P. M., Thoma, R., Claus, E. D., Mays, N., and Caprihan, A. (2013). Reduced white matter integrity in the cingulum and anterior corona radiata in posttraumatic stress disorder in male combat veterans: a diffusion tensor imaging study. *Psychiatry Res.* 214, 260–268. doi: 10.1016/j.pscychresns.2013.09.002
- Saver, J. L. (2006). Time is brain-quantified. *Stroke* 37, 263–266. doi: 10.1161/01.STR.0000196957.55928.ab
- Su, Y., Qi, F., Wang, M., Xie, Q., Gao, J., Zhang, D., et al. (2020). Changes of white matter microstructure in middle-aged patients with type 2 diabetes mellitus and no cognitive impairment. *Diagn. Imag. Intervent. Radiol.* 29, 177–182. doi: 10.3969/j.issn.1005-8001.2020.03.004
- Sun, Q., Chen, G. Q., Wang, X. B., and Yu, Y. (2018). Alterations of white matter integrity and hippocampal functional connectivity in type 2 diabetes without mild cognitive impairment. *Front. Neuroanat.* 12:21. doi: 10.3389/fnana.2018.00021
- Tahmasian, M., Eickhoff, S. B., Giehl, K., Schwartz, F., Herz, D. M., Drzezga, A., et al. (2017). Resting-state functional reorganization in Parkinson's disease: an activation likelihood estimation meta-analysis. *Cortex* 92, 119–138. doi: 10.1016/j.cortex.2017.03.016
- Tu, H. P., Lin, C. H., Hsieh, H. M., Jiang, H. J., Wang, P. W., Huang, C. J., et al. (2017). Prevalence of anxiety disorder in patients with type 2 diabetes: a nationwide population-based study in Taiwan 2000-2010. *Psychiatr. Q.* 88, 75–91. doi: 10.1007/s11126-016-9436-0
- Turkeltaub, P. E., Eden, G. F., Jones, K. M., and Zeffiro, T. (2002). Meta-analysis of the functional neuroanatomy of single-word reading: method and validation. *Neuroimage* 16(3 Pt 1), 765–780. doi: 10.1006/nimg.2002.1131
- van den Heuvel, M., Mandl, R., Luigjes, J., and Hulshoff Pol, H. (2008). Microstructural organization of the cingulum tract and the level of default mode functional connectivity. *J. Neurosci.* 28, 10844–10851. doi: 10.1523/JNEUROSCI.2964-08.2008
- Wakana, S., Jiang, H., Nagae-Poetscher, L. M., van Zijl, P. C., and Mori, S. (2004). Fiber tract-based atlas of human white matter anatomy. *Radiology* 230, 77–87. doi: 10.1148/radiol.2301021640
- Xia, W., Chen, Y. C., and Ma, J. (2017). Resting-state brain anomalies in type 2 diabetes: a meta-analysis. *Front. Aging Neurosci.* 9:14. doi: 10.3389/fnagi.2017.00014
- Xie, Y., Zhang, Y., Qin, W., Lu, S., Ni, C., and Zhang, Q. (2017). White matter microstructural abnormalities in type 2 diabetes mellitus: a diffusional kurtosis imaging analysis. *AJNR Am. J. Neuroradiol.* 38, 617–625. doi: 10.3174/ajnr.A5042
- Xiong, Y., Sui, Y., Xu, Z., Zhang, Q., Karaman, M. M., Cai, K., et al. (2016). A diffusion tensor imaging study on white matter abnormalities in patients with type 2 diabetes using tract-based spatial statistics. *AJNR Am. J. Neuroradiol.* 37, 1462–1469. doi: 10.3174/ajnr.A4740
- Xiong, Y., Sui, Y., Zhang, S., Zhou, X. J., Yang, S., Fan, Y., et al. (2019a). Brain microstructural alterations in type 2 diabetes: diffusion kurtosis imaging provides added value to diffusion tensor imaging. *Eur. Radiol.* 29, 1997–2008. doi: 10.1007/s00330-018-5746-y
- Xiong, Y., Zhang, S., Shi, J., Fan, Y., Zhang, Q., and Zhu, W. (2019b). Application of neurite orientation dispersion and density imaging to characterize brain microstructural abnormalities in type-2 diabetics with mild cognitive impairment. J. Magn. Reson. Imaging 50, 889–898. doi: 10.1002/jmri. 26687
- Yan, C. G., Wang, X. D., Zuo, X. N., and Zang, Y. (2016). DPABI: data processing and analysis for (resting-state) brain imaging. *Neuroinformatics* 14, 339–351. doi: 10.1007/s12021-016-9299-4
- Yang, L. L., Huang, Y. N., and Cui, Z. T. (2014). Clinical features of acute corpus callosum infarction patients. *Int. J. Clin. Exp. Pathol.* 7, 5160–5164.
- Yau, P. L., Hempel, R., Tirsi, A., and Convit, A. (2013). Cerebral white matter and retinal arterial health in hypertension and type 2 diabetes mellitus. *Int. J. Hypertens*. 2013:329602. doi: 10.1155/2013/329602
- Yau, P. L., Javier, D., Tsui, W., Sweat, V., Bruehl, H., Borod, J., et al. (2009). Emotional and neutral declarative memory impairments and associated white matter microstructural abnormalities in adults with type 2 diabetes. *Psychiatry Res.* 174, 223–230. doi: 10.1016/j.pscychresns.2009.04.016
- Yau, P. L., Javier, D. C., Ryan, C. M., Tsui, W., Ardekani, B. A., Ten, S., et al. (2010). Preliminary evidence for brain complications in obese adolescents with type 2 diabetes mellitus. *Diabetologia* 53, 2298–2306. doi: 10.1007/s00125-010-1857-y

- Yoon, S., Cho, H., Kim, J., Lee, D. W., Kim, G., Hong, Y. S., et al. (2017). Brain changes in overweight/obese and normal-weight adults with type 2 diabetes mellitus. *Diabetologia* 60, 1207–1217. doi: 10.1007/s00125-017-4266-7
- Zheng, Y., Ley, S. H., and Hu, F. B. (2018). Global aetiology and epidemiology of type 2 diabetes mellitus and its complications. *Nat. Rev. Endocrinol.* 14, 88–98. doi: 10.1038/nrendo.2017.151
- Zhuo, Y., Fang, F., Lu, L., Li, T., Lian, J., Xiong, Y., et al. (2019). White matter impairment in type 2 diabetes mellitus with and without microvascular disease. *Neuroimage Clin.* 24:101945. doi: 10.1016/j.nicl.2019.101945

Conflict of Interest: The authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

Publisher's Note: All claims expressed in this article are solely those of the authors and do not necessarily represent those of their affiliated organizations, or those of the publisher, the editors and the reviewers. Any product that may be evaluated in this article, or claim that may be made by its manufacturer, is not guaranteed or endorsed by the publisher.

Copyright © 2021 Huang, Zhang, Tang, Yang, Chen, Tao and Liang. This is an open-access article distributed under the terms of the Creative Commons Attribution License (CC BY). The use, distribution or reproduction in other forums is permitted, provided the original author(s) and the copyright owner(s) are credited and that the original publication in this journal is cited, in accordance with accepted academic practice. No use, distribution or reproduction is permitted which does not comply with these terms.