

## Original Article

**Cite this article:** Meinert S *et al* (2020). White matter fiber microstructure is associated with prior hospitalizations rather than acute symptomatology in major depressive disorder. *Psychological Medicine* 1–9. <https://doi.org/10.1017/S0033291720002950>

Received: 18 March 2020

Revised: 29 June 2020

Accepted: 3 August 2020

**Key words:**

Acute symptomatology; course of illness; diffusion tensor imaging; hospitalization; major depressive disorder; remitted; superior longitudinal fasciculus

**Author for correspondence:**

Udo Dannlowski,

E-mail: [dannlow@uni-muenster.de](mailto:dannlow@uni-muenster.de)

# White matter fiber microstructure is associated with prior hospitalizations rather than acute symptomatology in major depressive disorder

Susanne Meinert<sup>1</sup>, Elisabeth J. Leehr<sup>1</sup>, Dominik Grotegerd<sup>1</sup>, Jonathan Repple<sup>1</sup>, Katharina Förster<sup>1,2</sup>, Nils R. Winter<sup>1</sup>, Verena Enneking<sup>1</sup>, Stella M. Fingas<sup>1</sup>, Hannah Lemke<sup>1</sup>, Lena Waltemate<sup>1</sup>, Frederike Stein<sup>3</sup>, Katharina Brosch<sup>3</sup>, Simon Schmitt<sup>3</sup>, Tina Meller<sup>3</sup>, Anna Linge<sup>1</sup>, Axel Krug<sup>3,8</sup>, Igor Nenadić<sup>3</sup>, Andreas Jansen<sup>3,4</sup>, Tim Hahn<sup>1</sup>, Ronny Redlich<sup>1</sup>, Nils Opel<sup>1,9</sup>, Ricarda I. Schubotz<sup>5</sup>, Bernhard T. Baune<sup>1,6,7</sup>, Tilo Kircher<sup>3</sup> and Udo Dannlowski<sup>1</sup> 

<sup>1</sup>Department of Psychiatry, University of Münster, Münster, Germany; <sup>2</sup>Clinical Psychology and Behavioral Neuroscience, Faculty of Psychology, Technische Universität Dresden, Dresden, Germany; <sup>3</sup>Department of Psychiatry and Psychotherapy, University of Marburg, Marburg, Germany; <sup>4</sup>Core-Unit Brainimaging, Faculty of Medicine, University of Marburg, Marburg, Germany; <sup>5</sup>Department of Psychology, University of Münster, Münster, Germany; <sup>6</sup>Department of Psychiatry, Melbourne Medical School, The University of Melbourne, Melbourne, VIC, Australia; <sup>7</sup>The Florey Institute of Neuroscience and Mental Health, The University of Melbourne, Parkville, VIC, Australia; <sup>8</sup>Department of Psychiatry and Psychotherapy, University of Bonn, Bonn, Germany and <sup>9</sup>Interdisciplinary Centre for Clinical Research (IZKF) Münster, University of Münster, Münster, Germany

## Abstract

**Background.** Eighty percent of all patients suffering from major depressive disorder (MDD) relapse at least once in their lifetime. Thus, understanding the neurobiological underpinnings of the course of MDD is of utmost importance. A detrimental course of illness in MDD was most consistently associated with superior longitudinal fasciculus (SLF) fiber integrity. As similar associations were, however, found between SLF fiber integrity and acute symptomatology, this study attempts to disentangle associations attributed to current depression from long-term course of illness.

**Methods.** A total of 531 patients suffering from acute ( $N = 250$ ) or remitted ( $N = 281$ ) MDD from the FOR2107-cohort were analyzed in this cross-sectional study using tract-based spatial statistics for diffusion tensor imaging. First, the effects of disease state (acute *v.* remitted), current symptom severity (BDI-score) and course of illness (number of hospitalizations) on fractional anisotropy (FA), mean diffusivity (MD), radial diffusivity (RD), and axial diffusivity were analyzed separately. Second, disease state and BDI-scores were analyzed in conjunction with the number of hospitalizations to disentangle their effects.

**Results.** Disease state ( $p_{\text{FWE}} < 0.042$ ) and number of hospitalizations ( $p_{\text{FWE}} < 0.032$ ) were associated with decreased FA and increased MD and RD in the bilateral SLF. A trend was found for the BDI-score ( $p_{\text{FWE}} > 0.067$ ). When analyzed simultaneously only the effect of course of illness remained significant ( $p_{\text{FWE}} < 0.040$ ) mapping to the right SLF.

**Conclusions.** Decreased FA and increased MD and RD values in the SLF are associated with more hospitalizations when controlling for current psychopathology. SLF fiber integrity could reflect cumulative illness burden at a neurobiological level and should be targeted in future longitudinal analyses.

## Introduction

Half of all patients suffering from major depressive disorder (MDD) relapse within 2 years after recovery (Kanai *et al.*, 2003) and about 80% of patients in remission experience at least one recurrence in their lifetime (Vos *et al.*, 2004). Rates of relapse increase with every subsequent episode (Ferrari *et al.*, 2013; Keller & Boland, 1998), the time remaining in recovery decreases (Solomon *et al.*, 2000) and the likelihood of recovery diminishes with longer durations of previous episodes (de Carlo, Calati, & Serretti, 2016; Otte *et al.*, 2016; Spijker *et al.*, 2004). Therefore, understanding the pathophysiology of course of illness is of utmost clinical importance.

Current models of recurrence focus on risk factors associated with a less favorable course. These include demographic variables (e.g. gender and socioeconomic status) (Burcusa & Iacono, 2007) and clinical characteristics (higher symptom severity and longer duration especially of the first episode, psychiatric comorbidities, etc.) (Kraus, Kadriu, Lanzenberger, Jr., Carlos, Jr., & Kasper, 2019; Penninx *et al.*, 2011). Also, psychological factors such as the

level of psychosocial functioning or stressful life events are associated with recurrent MDD (Burcusa & Iacono, 2007; Hovens et al., 2012; Penninx et al., 2011). ‘Scar theories’ try to explain the increased rates of relapse in the course of MDD (Burcusa & Iacono, 2007). They postulate that depression itself increases the vulnerability to become depressed again, for example as sensitivity to future stressful life events increases (Post, Leverich, Xing, & Weiss, 2001). At a biological level, this could be translated to an elevated hypothalamic–pituitary–adrenal (HPA) axis (Oberlander et al., 2008; Targum, 1984; Varghese & Brown, 2001) that becomes increasingly dysregulated in the course of MDD (McEwen, 2003; Zaremba et al., 2018a, 2018b). As a consequence, increased glucocorticoid secretion might potentially reduce myelination (Jauregui-Huerta et al., 2010), inhibit a brain-derived neurotrophic factor important for the formation of neural connections and dendritic branching (Campbell & MacQueen, 2004), and thus, affect fiber microstructure. A non-invasive technique to measure fiber structure is diffusion tensor imaging (DTI). DTI quantifies water diffusion to reflect myelination, axon density, axon diameter, and number of fibers (Le Bihan, 2003; Winston, 2012). Its interpretation is limited, as DTI estimations are also influenced by fiber crossings and general fiber orientation in healthy fiber structure (Jones, Knösche, & Turner, 2013). Nonetheless, reduced fiber integrity in MDD patients compared to healthy controls has already been meta-analytically confirmed (Wise et al., 2016).

In DTI studies, results investigating course of illness are contradictory (de Diego-Adeliño et al., 2014) found a negative association of frontal white matter integrity and course of illness using the duration of illness and number of previous episodes. Similarly, Abe et al. (2010) found a trend of a negative association with total days depressed in the anterior cingulate cortex and the left frontal white matter. Other studies failed to find any association between white matter and course of illness using illness duration and number of depressive episodes (Guo et al., 2012a; Li et al., 2007).

A meta-analysis by Murphy and Frodl (2011) illustrated that course of illness measured with the mean illness duration per study was associated with reduced fiber integrity in the superior longitudinal fasciculus (SLF). They were, however, unable to disentangle the effects of course of illness from the effects of acute symptomatology, as they overlapped anatomically. As residual symptoms are common and persistent after severe episodes (Kennedy, Abbott, & Paykel, 2004) and are a risk factor for relapse (Burcusa & Iacono, 2007; Paykel, 2008), it is possible that the course of illness is a confounder driving the significant effect of acute symptomatology or vice versa. However, for a potential theory of the neurobiological underpinnings of MDD, it is important to distinguish state (e.g. acute symptomatology) from prolonged (e.g. course of illness) effects. The former, on the one hand, would indicate present psychopathological processes which might for example be used to guide treatment evaluations. The latter, on the other hand, could help to understand the devastating effects of MDD chronicity and might help to identify patients with higher need for more intensive care.

Thus, to separate course of illness from acute symptomatology, two facets of the latter will be differentiated: first, the categorical distinction of patients suffering from an acute depressive episode compared to remitted MDD (disease state, acute v. remitted MDD) and second, the self-reported continuum of the number and severity of depressive symptoms (symptom severity). On the one hand, disease state was associated with decreased fiber integrity in the ventromedial prefrontal region in treatment-resistant MDD

compared with remitted recurrent MDD (de Diego-Adeliño et al., 2014), whereas no distinction of patients with an acute MDD episode compared with remitted MDD patients was found in other studies (Bae et al., 2006; Harada et al., 2018). Symptom severity in MDD patients, on the other hand, has been linked to reduced fiber integrity in the SLF, the corticospinal tract, the uncinate fasciculus and frontal brain regions in general (Bergamino, Kuplicki, Victor, Cha, & Paulus, 2017; Dalby et al., 2010; de Diego-Adeliño et al., 2014; Nobuhara et al., 2006; Repple et al., 2020). Yet again, other studies found no association (Abe et al., 2010; Bae et al., 2006; Guo et al., 2012a, 2012b; Jia et al., 2010; Li et al., 2007).

The contradictory results concerning the systematic association between course of illness, disease state, and symptom severity in MDD patients and white matter microstructure could in part be due to small and heterogeneous samples (Burcusa & Iacono, 2007). Furthermore, the methods used to quantify these three factors differ considerably. Zaremba et al. (2018b) demonstrated reduced hippocampal gray matter volume in patients suffering from MDD using the duration and number of hospitalizations, whereas no association was found with the duration and frequency of depressive episodes. These results highlight that different operationalizations of the same construct can lead to divergent results. Patients might be able to specify hospitalizations more reliably, as they constitute a decisive event in their everyday life. In addition, hospitalizations can be verified using medical records. The number of depressive episodes, on the other hand, are a less clear cut construct for patients and can be prone to autobiographical memory biases (Williams et al., 2007).

Understanding the biological basis of white matter fiber structure alterations in depressed patients is a prerequisite to improve clinical care. The aim of this study was, thus, to replicate the meta-analytical results of Murphy and Frodl (2011) and to distinguish acute symptomatology (disease state and symptom severity) and course of illness. This distinction between state (acute symptomatology) and prolonged (course of illness) effects is important, to further the understanding of the neurobiological underpinnings of MDD. To this end, three concepts were analyzed: (1) disease state, defined as the presence of clinically relevant depressive symptoms in acute MDD patients in comparison with remitted MDD patients; (2) current symptom severity, defined as the dimensional measurement of depressive symptom severity (BDI-scores); and (3) course of illness, defined as the number of hospitalizations in the past. We expect that all three concepts should be associated with reduced fiber integrity in the SLF replicating meta-analytical results (Murphy & Frodl, 2011). Furthermore, the effects of disease state and symptom severity are analyzed in conjunction with the number of hospitalizations to distinguish their respective effects.

## Materials and methods

### Participants

A total of 531 participants were selected from the FOR2107-cohort (see online Supplementary S1) for this cross-sectional study (Kircher et al., 2019; Vogelbacher et al., 2018);  $N = 250$  currently depressed patients ( $N = 149$  female,  $M_{age} = 37.28$ ,  $s.D._{age} = 13.24$ ) compared with  $N = 281$  remitted depressed patients ( $N = 141$  in partial remission,  $N = 140$  in complete remission;  $N = 189$  female;  $M_{age} = 36.57$ ,  $s.D._{age} = 13.41$ ). Data were collected at two scanning sites (Marburg and Münster). As the

body-coil had to be exchanged in Marburg mid recruitment, three different scanner settings had to be addressed in all analyses (Marburg pre body-coil, Marburg post body-coil and Münster).

The Structural Clinical Interview for DSM-IV-TR (SCID-I) (Wittchen, Wunderlich, Gruschwitz, & Zaudig, 1997) was used by trained personnel to verify clinical diagnoses. Exclusion criteria comprised age below 18 or above 65 years, history of substance dependence, bipolar disorder, schizophrenia, schizoaffective disorder, neurological abnormalities, history of seizures, longer periods of unconsciousness or severe head trauma, severe physical impairment, claustrophobia, color blindness, and general magnetic resonance imaging (MRI) contradictions. To measure medication intake, the Medication Load Index (Redlich et al., 2015) was used. The Medication Load Index is a composite measure reflecting dose and number of prescriptions irrespective of active components. The sum of all psychopharmacological medication per patient was calculated with dosage categories of absent (= 0), low (= 1, a dosage equal or lower than average), or high (= 2, a dosage greater than average) relative to the midpoint of the daily dose range recommended by Physician's-Desk-Reference (2017). Patients were asked to report the number of previous hospitalizations. All patients underwent inpatient treatment predominantly for a severe depressive episode. To assess current depressive symptoms the Beck Depression Inventory (BDI) (Beck, Ward, Mendelson, Mock, & Erbaugh, 1961) was used. Socioeconomic status was operationalized as the monthly income per household.

The authors assert that all procedures contributing to this work comply with the ethical standards of the relevant national and institutional committees on human experimentation and with the Helsinki Declaration of 1975, as revised in 2008. All procedures involving human patients were approved by the ethics committees of the Medical Faculties of the Universities of Münster (2014-422-b-S) and Marburg (AZ: 07/14). All experiments were performed in accordance with the ethical guidelines and regulations. Written informed consent was obtained from all participants prior to examination. They received financial compensation (50€) for participation after the testing session.

### DTI data acquisition

The following methods were already described in greater detail in previous analyses (Meinert et al., accepted). Data were acquired using a 3T whole body MRI scanner (Marburg: Tim Trio, Siemens, Erlangen, Germany; Münster: Prisma, Siemens, Erlangen, Germany) using a GRAPPA acceleration factor of 2. Fifty-six axial slices, 2.5 mm with no gap, were measured with an isotropic voxel size of  $2.5 \times 2.5 \times 2.5 \text{ mm}^3$  (TE = 90 ms, TR = 7300 ms). Five non-diffusion-weighted (DW) images ( $b_0 = 0$ ) and 60 DW images with a  $b$ -value of  $1000 \text{ s/mm}^2$  were measured. To ensure data quality open-source software DTIPrep (Oguz et al., 2014) was used with default options correcting for slice-wise and gradient-wise inconsistencies plus eddy-current, head motion, bed vibration, and pulsation or venetian blind artifacts. DW images affected by artifacts were excluded with more than 20% of excluded images resulting in the complete drop-out of that participant before matching. The included participants had 64.27 images on average (s.d. = 1.42, range: [54–65]).

### Image processing

FSL5.0.10 (FMRIB Image Analysis Group, Oxford, UK, <http://fsl.fmrib.ox.ac.uk/fsl/fslwiki/>) (Jenkinson, Beckmann, Behrens,

Woolrich, & Smith, 2012; Smith et al., 2004; Woolrich et al., 2009) was used for preprocessing and analysis. The DW images were corrected for eddy and motion artifacts using FSL's 'eddy' (Andersson & Sotiroopoulos, 2016), with  $b$ -vectors being rotated after eddy correction. As reference for alignment, the first  $b_0$  image was used after automated skull stripping with the Brain Extraction Tool (Smith, 2002) in FSL. For diffusion tensor estimation the 'DTIFIT' within FMRIB's Diffusion Toolbox (FDT) (Behrens et al., 2003) was used to generate tensor-derived maps, resulting in an estimation of fractional anisotropy (FA), mean diffusivity (MD), radial diffusivity (RD), and axial diffusivity (AD) for each voxel per participant. FA is defined as the normalized variance of the three eigenvalues about their mean. FA quantifies directional diffusion, ranging between 0 and 1, reflecting isotropic and anisotropic diffusion, respectively. AD reflects diffusivity along the primary diffusion direction, which should represent tract orientation and RD perpendicular to the tract. MD is calculated as the average of all three dimensions. In general, increased MD and decreased FA can hint at neuronal injury, increased AD was associated with damage to the axon, and increased RD with demyelination (Feldman, Yeatman, Lee, Barde, & Gaman-Bean, 2010). However, any interpretation has to be drawn with caution, as the number of fibers, fiber crossings, and general fiber orientation can also influence diffusion metrics in healthy fiber structure (Jones et al., 2013).

### Analysis

Analyses on demographic data were conducted using IBM SPSS Statistics 25 (SPSS Inc., Chicago, IL, USA). Tract-based spatial statistics (TBSS) (Smith et al., 2006) was used to reduce partial volume effects and registration misalignments. First, FA images were aligned to the FMRIB58\_FA template [ $1 \times 1 \times 1 \text{ mm}^3$  Montreal Neurological Institute (MNI) standard space] using FMRIB's non-linear image registration tool. Second, the mean of all aligned FA images was skeletonized with a threshold of 0.2. Third, the mean FA skeleton was filled with the maximum weighted for distance FA values orthogonal to the skeleton from each participant's aligned FA image. The same registration steps were used on MD, AD, and RD maps. Voxel-wise statistical analyses were performed in skeleton space, correcting for multiple comparisons with a threshold-free cluster enhancement (TFCE) (Smith & Nichols, 2009) approach using the non-parametric permutation testing implemented in FDT's 'randomise' (Nichols & Holmes, 2002) (5000 permutations) with default values provided by  $-T2$  option optimized for TBSS. The null distribution of permuted input data of the maximum TFCE scores was used to determine significance, correcting for the family-wise error (FWE) at  $p < 0.05$ . In case of significant results peak voxel MNI coordinates and cluster sizes were retrieved with the 'cluster' tool implemented in FSL and tract labels were extracted with FSL's 'atlasquery' using the 'JHU White-Matter Tractography Atlas' (Hua et al., 2008; Mori, Wakana, van Zijl, & Nagae-Poetscher, 2005; Wakana et al., 2007). FSL's 'fslstats' was used to extract mean diffusion metric values of significant clusters. The total intracranial volume (TIV) was extracted from T1 images using the Computational Anatomy Toolbox (CAT-12, <http://www.neuro.uni-jena.de/cat, v933>). For the T1 sequence see Vogelbacher et al. (2018).

To correct for scanner differences two dummy coded variables (Marburg pre body-coil: yes  $v.$  no, Marburg post body-coil: yes  $v.$  no) with Münster as reference category were defined. The influence of disease state (acute MDD  $v.$  remitted MDD),

**Table 1.** Demographic and clinical characteristics of the sample

Characteristics	Current MDD (N = 250)	Remitted MDD (N = 281)	p value
Sociodemographic			
Sex, m/f	101/149	92/189	0.067 <sup>a</sup>
Age, years, <i>M</i> ± s.d.	37.28 ± 13.24	36.57 ± 13.42	0.540 <sup>b</sup>
Clinical characteristics			
BDI, <i>M</i> ± s.d.	24.59 ± 10.19	12.85 ± 9.41	0.001 <sup>b</sup>
Medication Load Index, <i>M</i> ± s.d.	1.68 ± 1.43	0.91 ± 1.20	0.001 <sup>b</sup>
Socioeconomic status, <i>M</i> ± s.d. <sup>c</sup>	1711 ± 1378	1808 ± 1377	0.446 <sup>b</sup>
Comorbid disorders, yes/no	121/129	111/170	0.039 <sup>a</sup>
Course of illness			
Number of hospitalization, <i>M</i> ± s.d.	2.11 ± 2.39	1.24 ± 1.75	0.001 <sup>b</sup>
Time in inpatient treatment, month, <i>M</i> ± s.d. <sup>c</sup>	14.65 ± 19.23	8.87 ± 13.58	<0.001 <sup>b</sup>
Number of depressive episodes, <i>M</i> ± s.d. <sup>c</sup>	4.51 ± 5.81	3.88 ± 6.19	0.247 <sup>b</sup>
Months since first psychiatric treatment, <i>M</i> ± s.d. <sup>c</sup>	82.07 ± 92.56	99.96 ± 104.71	0.058 <sup>b</sup>

BDI, Beck Depression Inventory; f, female; m, male; *M*, mean; s.d., standard deviation.

Notes: <sup>a</sup> $\chi^2$  test, <sup>b</sup>t test, <sup>c</sup>not all participants provided the necessary information, *N*<sub>min</sub> = 197 current MDD and *N*<sub>min</sub> = 231 remitted MDD.

BDI-values and number of hospitalizations were analyzed in separate analysis of covariances (ANCOVAs), with FA, AD, RD, and MD as dependent variables and age, sex, TIV, Marburg pre body-coil and Marburg post body-coil as nuisance variables. Furthermore, either disease state or symptom severity were analyzed simultaneously with number of hospitalizations to estimate the main effect of disease state/BDI-score, the main effect of number of hospitalizations and their interaction corrected for age, sex, TIV, Marburg pre body-coil, and Marburg post body-coil. The interactions of symptom severity and the number of hospitalizations were included, as additive and interactive effects, e.g. current psychopathological process being intensified in chronic patients, could not be ruled out *a priori* due to the lack of similar analyses in the literature.

In case of significant results, mean FA, AD, RD, and MD values per participant from the significant cluster were extracted. The same ANCOVA as mentioned above was repeated with additional variables for medication intake (Medication Load Index), socioeconomic status (monthly income), month since first psychiatric treatment, and presence of comorbid diagnosis (yes v. no) in SPSS to correct for their respective influence. For these analyses 23 participants had to excluded, as they did not complete all the necessary questionnaires.

As the number of hospitalizations were skewed, Cook's distance of the regression of disease state/BDI-score, number of hospitalizations, their interaction and age, sex, TIV, Marburg pre body-coil, and Marburg post body-coil on mean extracted FA, AD, RD, and MD values were calculated in SPSS. Influential data points that could exert strong leverage on the regression slope (extreme values in Cooks' distance) were excluded to check whether the regression in SPSS is robust to outliers.

## Results

### Descriptive statistics

Acutely depressed patients had more prior hospitalizations compared with remitted depressed patients ( $t_{(529)} = 4.82$ ,  $p < 0.001$ ;

**Table 1**). Furthermore, a significant positive correlation between higher BDI scores and a higher number of hospitalizations was found ( $r = 0.198$ ,  $p < 0.001$ ).

### Separate effects of disease state, symptom severity, and course of illness

#### Disease state

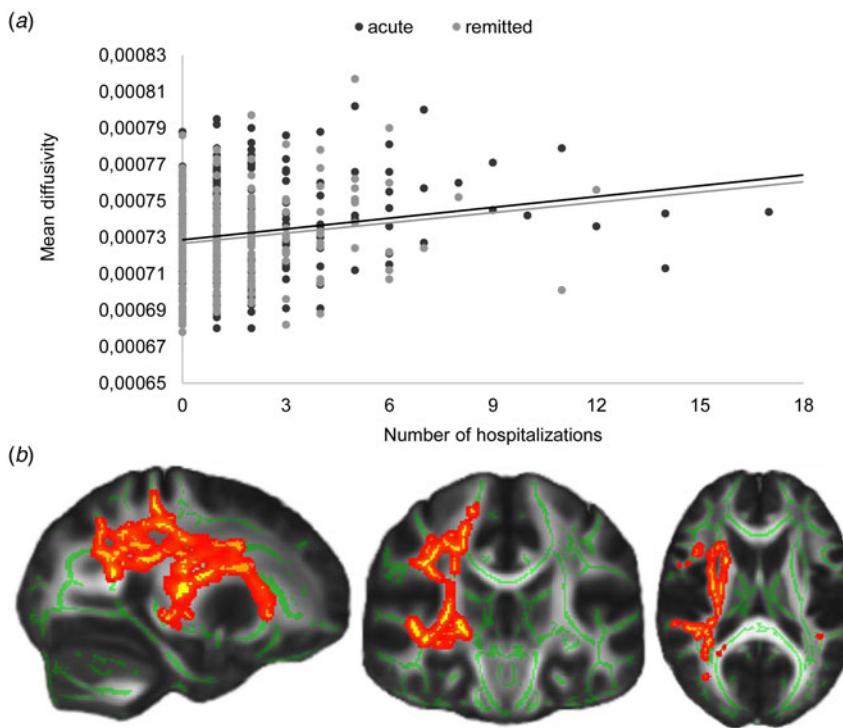
FA was lower although MD and RD were higher in acute MDD patients compared to remitted MDD patients (FA:  $p_{\text{FWE}} = 0.031$ , total  $k = 5109$  voxel in 11 clusters, peak voxel of largest cluster:  $x = 26$ ,  $y = -40$ ,  $z = 32$ ; MD:  $p_{\text{FWE}} = 0.042$ , total  $k = 3535$  voxel in 3 clusters, peak voxel of largest cluster:  $x = 38$ ,  $y = -46$ ,  $z = 25$ ; RD:  $p_{\text{FWE}} = 0.027$ , total  $k = 5633$  voxel in 10 clusters, peak voxel of largest cluster:  $x = 33$ ,  $y = -40$ ,  $z = 29$ ; online Supplementary Table S1). Effects had the highest probability to be part of the bilateral SLF. No significant effects were found in AD (all  $p_{\text{FWE}} > 0.226$ ).

#### Symptom severity

No significant effects were found in FA, RD, AD, and MD with symptom severity (FA:  $p_{\text{FWE}} = 0.218$ , RD:  $p_{\text{FWE}} = 0.127$ , AD:  $p_{\text{FWE}} = 0.109$ ). However, a statistical trend was present in MD with higher values of BDI-scores being associated with higher MD values in the bilateral SLF ( $p_{\text{FWE}} = 0.067$ ).

#### Course of illness

Lower FA and higher MD and RD were associated with more hospitalizations (FA:  $p_{\text{FWE}} = 0.032$ , total  $k = 3874$  voxel in 13 clusters, peak voxel of largest cluster:  $x = 37$ ,  $y = 6$ ,  $z = 22$ ; MD:  $p_{\text{FWE}} = 0.002$ , total  $k = 7876$  voxel in 2 clusters, peak voxel of largest cluster:  $x = 40$ ,  $y = -44$ ,  $z = 23$ ; RD:  $p_{\text{FWE}} = 0.019$ , total  $k = 15\,274$  voxel in 17 clusters, peak voxel of largest cluster:  $x = 26$ ,  $y = -29$ ,  $z = 42$ ; online Supplementary Table S1, Fig. 1). Again, effects had the highest probability to be part of the bilateral SLF. No significant effects were found in AD ( $p_{\text{FWE}} = 0.094$ ).



**Fig. 1.** The effect of number of hospitalizations. (a) Correlation between course number of hospitalizations and mean extracted MD values in acutely and remitted depressed patients. (b) Effect displayed on the FMRIB58 template, MNI coordinates of peak voxel:  $x = 26$ ,  $y = -24$ ,  $z = 15$ .

This association could also be found using a combination of different course of illness variables (online Supplementary S2).

#### ANCOVA with disease state, course of illness, and their interaction

No significant effects were found in FA (all  $p_{\text{FWE}} > 0.122$ ) or AD (all  $p_{\text{FWE}} > 0.080$ ). In MD and RD neither a main effect of disease state (MD:  $p_{\text{FWE}} = 0.194$ , RD:  $p_{\text{FWE}} = 0.110$ ) nor an interaction of disease state  $\times$  number of hospitalizations (MD:  $p_{\text{FWE}} = 0.387$ ; RD:  $p_{\text{FWE}} = 0.383$ ) reached level of significance. However, the previously observed association of MD and RD with previous hospitalizations remained significant in this model (MD:  $p_{\text{FWE}} = 0.029$ ,  $k = 5262$  voxels in 1 cluster, peak voxel:  $x = 38$ ,  $y = -46$ ,  $z = 23$ ; RD:  $p_{\text{FWE}} = 0.040$ , total  $k = 2244$  voxels in 3 cluster, peak voxel of largest cluster:  $x = 25$ ,  $y = -29$ ,  $z = 42$ ; online Supplementary Table S1). Effects were located in the right SLF with the highest probability. This relationship was robust to outliers (online Supplementary S3). The relationship between the mean extracted MD and RD voxels from the significant cluster and number of hospitalizations remained significant even after correcting for medication intake, socioeconomic status, month since first psychiatric treatment, and presence of comorbid diagnoses (online Supplementary Table S2).

#### ANCOVA with symptom severity, course of illness, and their interaction

No significant effects were found in AD (all  $p_{\text{FWE}} > 0.131$ ). In FA, MD, and RD, neither a main effect of BDI-scores (FA:  $p_{\text{FWE}} = 0.369$ , MD:  $p_{\text{FWE}} = 0.160$ , RD:  $p_{\text{FWE}} = 0.210$ ) nor an interaction of BDI-scores  $\times$  number of hospitalizations (FA:  $p_{\text{FWE}} = 0.547$ , MD:  $p_{\text{FWE}} = 0.674$ , RD:  $p_{\text{FWE}} = 0.827$ ) were significant. However, again a negative association of FA and a positive association of MD and RD with a more prior hospitalizations was confirmed

(FA:  $p_{\text{FWE}} = 0.038$ , total  $k = 1331$  voxels in 2 clusters, peak voxel of largest cluster  $x = 37$ ,  $y = -5$ ,  $z = 22$ ; MD:  $p_{\text{FWE}} = 0.024$ ,  $k = 6718$  voxels in 1 cluster, peak voxel of largest cluster:  $x = 39$ ,  $y = -41$ ,  $z = 24$ ; RD:  $p_{\text{FWE}} = 0.029$ , total  $k = 4393$  voxels in 1 cluster, peak voxel of largest cluster:  $x = 26$ ,  $y = -31$ ,  $z = 41$ ; online Supplementary Table S1), again with the highest probability to be located in the right SLF. This relationship was also robust to outliers (online Supplementary S4). The relationship between the mean extracted FA, MD, and RD voxels from the significant cluster and number of hospitalizations remained significant even after correcting for medication intake, socioeconomic status, month since first psychiatric treatment, and presence of comorbid diagnoses in SPSS (online Supplementary Table S3).

#### Discussion

The present cross-sectional study aimed to differentiate the effects of acute depression (disease state and BDI-score) from the influence of chronic disease burden (number of hospitalizations) on brain microstructural integrity in a large and well characterized sample of depressed patients.

More hospitalizations were associated with decreased fiber microstructure in the SLF. Similar results were found for disease state, whereas the association with the BDI-score only approached significance. The associations found between course of illness and acute symptomatology and SLF fiber structure, respectively, are in concordance with the meta-analysis by Murphy and Frodl (2011) and confirmed the number of hospitalizations as a reliable measure to approximate course of illness in MDD consistent with results from Zaremba et al. (2018b).

Moreover, we were able to expand upon meta-analytical results and disentangle the effects of course of illness and acute symptomatology: when analyzing disease state or BDI-scores in conjunction with the number of hospitalizations, associations with fiber microstructure attributed to acute depression were nullified.

In return, the positive relationship between more hospitalizations and decreased fiber microstructure in the right SLF remained significant. Notably, the association between number of hospitalizations and diffusion metrics in the SLF does not seem to be explained by differences in medication intake, socioeconomic status, time since first psychiatric treatment, or the presence of comorbid diagnoses, and was robust to the exclusion of outliers.

These results could be explained at a biological level by prolonged periods of elevated stress reflected in more frequent hospitalizations. This could set off a cascade of endocrinological, epigenetic, and immunological changes resulting in an increasingly dysregulated HPA axis (Bao & Swaab, 2019) and neurobiological changes in SLF fiber microstructure. As the SLF is a late-maturing tract and associated with healthy aging (Madhavan, McQueeney, Howe, Shear, & Szaflarski, 2014), it might be more vulnerable to neurotoxic effects over the course of a lifespan.

The role of SLF fiber integrity in the psychopathology of MDD is unclear. One possible hypothesis comes from studies of healthy controls, in which SLF microstructure was associated with cognitive deficits (Kerchner et al., 2012; Salami, Eriksson, Nilsson, & Nyberg, 2012; Turken et al., 2008). Interestingly, cognitive deficits in MDD persist after remission (Bora, Harrison, Yücel, & Pantelis, 2013; Ormel, Oldehinkel, Nolen, & Vollebergh, 2004; Rock, Roiser, Riedel, & Blackwell, 2014) and a worse course of illness is associated with increased cognitive deficits (Gorwood, Corruble, Falissard, & Goodwin, 2008). Furthermore, cognitive functioning is one prominent risk factor for disadvantageous occupational and psychosocial functioning (Bora et al., 2013; Evans, Iverson, Yatham, & Lam, 2014) that might in turn increase the risk for a more detrimental course of illness. Future studies need to investigate the possible link between course of illness, SLF fiber integrity, and cognitive dysfunction, to confirm this hypothesis.

Yet, SLF integrity differences in MDD cannot be explained solely by course of illness: reduced DTI metrics in the SLF compared to healthy controls were demonstrated early in the course of MDD (Wu et al., 2010) and even in first-episode MDD patients (Guo et al., 2012b). It is possible that risk factors such as childhood maltreatment experiences (Huang, Gundapuneedi, & Rao, 2012; Meinert et al., 2019) or a genetic predisposition (Barbu et al., 2019; Tozzi et al., 2016) might result in changes in diffusion metrics in the SLF prior to the development of MDD. Interestingly, both childhood maltreatment (Cross, Fani, Powers, & Bradley, 2017; Tarullo & Gunnar, 2006) and genetic predispositions (Bao & Swaab, 2019; Terenina et al., 2019) were linked to HPA axis dysfunction. Thus, similar biological mechanisms – elevated stress and a dysregulated HPA axis resulting in endocrine, immunological, epigenetic, and neurobiological changes – could lead to SLF microstructural changes prior to the onset of MDD. Future studies should try to measure HPA axis dysfunction directly and link it to fiber structure to confirm this hypothesis.

Finally, the results of this study are cross-sectional and, thus, do not allow causal inference. To completely disentangle the effects of acute symptomatology and course of illness, future longitudinal studies are necessary. In small groups of MDD patients, longitudinal studies found either associations between fiber integrity changes over the follow-up period and depressive symptom changes (Bracht, Jones, Müller, Wiest, & Walther, 2015; Doolin et al., 2019) or no association with course of illness or symptom severity, respectively (Repple et al., 2019). The largest longitudinal study to date ( $N=18\,959$ ) underscores the importance of the

longitudinal illness burden (Shen et al., 2019): Although decreased white matter microstructure was linked to depressive symptoms, the strongest association was found for the longitudinal progression of depressive symptoms in fronto-temporal association fibers. Longitudinal studies that focus on reliable measures of course of illness (e.g. number of hospitalizations) are needed.

### Limitations

A few limitations should be noted: first, in this study, symptom severity and the number of hospitalizations were correlated. It should be noted that this could be due to different recruiting strategies. Currently depressed patients were recruited in psychiatric hospitals on top of newspaper advertisements and outpatient settings. Thus, currently depressed patients had a higher chance to present with at least one if not more hospitalizations. This needs to be addressed in future studies that try to replicate the effects of a negative course of illness in an entirely outpatient setting. Nonetheless, this study includes a large, representative sample that approximates the broad spectrum of MDD.

Second, in contrast to the effects of disease state, the relationship between self-reported current symptom severity and fiber structure was not significant, though a trend was present in MD. Subjective ratings of psychopathology might be less reliable than diagnostic decisions from trained personnel. On the other hand, the lack of a strong association with symptom severity strengthens the notion that depressive symptoms are not the driving force associated with the microstructure of the SLF.

Third, we argue that the number of hospitalizations is the most reliable information provided by MDD patients to determine the course of illness. This information is, however, limited as only severely depressed patients (e.g. with loss of occupational functioning and suicidal ideation) undergo inpatient treatment in Germany. This could be confirmed in the structural clinical interview, as many patients reported severe episodes at one point in their life and it is mapped by an elevated BDI score ( $M_{current} = 24.59$ ,  $S.D._{current} = 10.19$ , Table 1) in the currently depressed sample. This point slightly limits the transferability of our results to less severe MDD patients and countries with different inpatient treatment practices.

Finally, although we correct current medication intake, previous medication and treatment history (e.g. psychotherapy and electroconvulsive therapy) could also affect the described association with course of illness. Future studies should assess these previous medication details in order to investigate this possible association in more detail.

### Conclusion

Course of illness in MDD – but not acute symptomatology – is associated with decreased structural integrity of the SLF. The major strength of this study was the use of large, well characterized patient group that allowed disentangling acute from cumulative effects of MDD on fiber structure. These results highlight that the number of hospitalizations is a feasible measure to approximate clinical course of MDD. It remains unclear, to what extent fiber microstructural changes are reversible and whether they can be influenced by treatment. The SLF is, hence, an important target for future longitudinal analyses investigating MDD recurrence.

**Supplementary material.** The supplementary material for this article can be found at <https://doi.org/10.1017/S0033291720002950>

**Acknowledgements.** This work is part of the German multicenter consortium ‘Neurobiology of Affective Disorders. A translational perspective on brain structure and function’, funded by the German Research Foundation (Deutsche Forschungsgemeinschaft DFG; Forschungsgruppe/Research Unit FOR2107). Principal investigators (PIs) with respective areas of responsibility in the FOR2107 consortium are: Work Package WP1, FOR2107 cohort and brainimaging: Tilo Kircher (speaker FOR2107; DFG grant numbers KI 588/14-1, KI 588/14-2), Udo Dannlowski (co-speaker FOR2107; DA 1151/5-1, DA 1151/5-2), Axel Krug (KR 3822/5-1, KR 3822/7-2), Igor Nenadic (NE 2254/1-2), Carsten Konrad (KO 4291/3-1). WP2, animal phenotyping: Markus Wöhr (WO 1732/4-1, WO 1732/4-2), Rainer Schwarting (SCHW 559/14-1, SCHW 559/14-2). WP3, miRNA: Gerhard Schratt (SCHR 1136/3-1, 1136/3-2). WP4, immunology, mitochondriae: Judith Alferink (AL 1145/5-2), Carsten Culmsee (CU 43/9-1, CU 43/9-2), Holger Garn (GA 545/5-1, GA 545/7-2). WP5, genetics: Marcella Rietschel (RI 908/11-1, RI 908/11-2), Markus Nöthen (NO 246/10-1, NO 246/10-2), Stephanie Witt (WI 3439/3-1, WI 3439/3-2). WP6, multi-method data analytics: Andreas Jansen (JA 1890/7-1, JA 1890/7-2), Tim Hahn (HA 7070/2-2), Bertram Müller-Myhsok (MU1315/8-2), Astrid Dempfle (DE 1614/3-1, DE 1614/3-2). CP1, biobank: Petra Pfefferle (PF 784/1-1, PF 784/1-2), Harald Renz (RE 737/20-1, 737/20-2). CP2, administration: Tilo Kircher (KI 588/15-1, KI 588/17-1), Udo Dannlowski (DA 1151/6-1), Carsten Konrad (KO 4291/4-1).

Acknowledgements and members by Work Package 1: Henrike Bröhl, Bruno Dietsche, Rozbeh Elahi, Jennifer Engelen, Sabine Fischer, Jessica Heinen, Svenja Klingel, Felicitas Meier, Torsten Sauder, Annette Tittmar, Dilara Yüksel (Dept. of Psychiatry, Marburg University). Mechthild Wallnig, Rita Werner (Core-Facility Brainimaging, Marburg University). Carmen Schade-Brittinger, Maik Hahmann (Coordinating Centre for Clinical Trials, Marburg). Michael Putzke (Psychiatric Hospital, Friedberg). Rolf Speier, Lutz Lenhard (Psychiatric Hospital, Haina). Birgit Köhnlein (Psychiatric Practice, Marburg). Peter Wulf, Jürgen Kleebach, Achim Becker (Psychiatric Hospital Hephata, Schwalmstadt-Treysa). Ruth Bär (Care facility Bischoff, Neunkirchen). Matthias Müller, Michael Franz, Siegfried Scharmann, Anja Haag, Kristina Spennner, Ulrich Ohlenschläger (Psychiatric Hospital Vitos, Marburg). Matthias Müller, Michael Franz, Bernd Kundermann (Psychiatric Hospital Vitos, Gießen). Christian Bürger, Fanni Dzvonyar, Stella Fingas, Hannah Lemke, Kordula Vorspohl, Bettina Walden, Dario Zaremba (Dept. of Psychiatry, University of Münster). Harald Kugel, Jochen Bauer, Walter Heindel, Birgit Vahrenkamp (Dept. of Clinical Radiology, University of Münster). Gereon Heuft, Gudrun Schneider (Dept. of Psychosomatics and Psychotherapy, University of Münster). Thomas Reker (LWL-Hospital Münster). Gisela Bartling (IPP Münster). Ulrike Buhlmann (Dept. of Clinical Psychology, University of Münster).

We are deeply indebted to all participants of this study, the recruitment sites, and their staff.

**Financial support.** This work was supported by the German Research Foundation (Deutsche Forschungsgemeinschaft, DFG) (TK, grant numbers KI 588/14-1, KI 588/14-2, KI 588/15-1, KI 588/17-1), (UD, grant numbers DA 1151/5-1, DA 1151/5-2, DA 1151/6-1, SFB-TRR58, Projects C09 and Z02), (AK, grant numbers KR 3822/5-1, KR 3822/7-2), (IN, grant numbers NE 2254/1-2), (AJ, grant numbers JA 1890/7-1, JA 1890/7-2), (TH, grant number HA 7070/2-2); the Interdisciplinary Center for Clinical Research (IZKF) of the medical faculty of Münster (UD, grant number Dan3/012/17), (NO, grant number SEED 11/18); ‘Innovative Medizinische Forschung’ (IMF) of the medical faculty of Münster (RR, grant numbers RE111604, RE111722), (JR, grant number RE221707), (EJL, grant number LE121703), and the Deanery of the Medical Faculty of the University of Münster.

**Conflict of interest.** Tilo Kircher received unrestricted educational grants from Servier, Janssen, Recordati, Aristo, Otsuka, and neurapharm. This funding is not associated with the current work. On behalf of all other authors, the corresponding author states that there is no conflict of interest and nothing to disclose.

## References

- Abe, O., Yamasue, H., Kasai, K., Yamada, H., Aoki, S., Inoue, H., ... Ohtomo, K. (2010). Voxel-based analyses of gray/white matter volume and diffusion tensor data in major depression. *Psychiatry Research*, 181, 64–70.
- Andersson, J. L. R., & Sotiropoulos, S. N. (2016). An integrated approach to correction for off-resonance effects and subject movement in diffusion MR imaging. *NeuroImage*, 125, 1063–1078.
- Bae, J. N., MacFall, J. R., Krishnan, K. R. R., Payne, M. E., Steffens, D. C., & Taylor, W. D. (2006). Dorsolateral prefrontal cortex and anterior cingulate cortex white matter alterations in late-life depression. *Biological Psychiatry*, 60, 1356–1363.
- Bao, A.-M., & Swaab, D. F. (2019). The human hypothalamus in mood disorders: The HPA axis in the center. *IBRO Reports*, 6, 45–53.
- Barbu, M. C., Zeng, Y., Shen, X., Cox, S. R., Clarke, T.-K., Gibson, J., ... Whalley, H. C. (2019). Association of whole-genome and NETRIN1 signaling pathway-derived polygenic risk scores for major depressive disorder and white matter microstructure in the UK biobank. *Biological Psychiatry: Cognitive Neuroscience and Neuroimaging*, 4, 91–100.
- Beck, A. T., Ward, C. H., Mendelson, M., Mock, J., & Erbaugh, J. (1961). An inventory for measuring depression. *Archives of General Psychiatry*, 4, 561–571.
- Behrens, T. E. J., Woolrich, M. W., Jenkinson, M., Johansen-Berg, H., Nunes, R. G., Clare, S., ... Smith, S. M. (2003). Characterization and propagation of uncertainty in diffusion-weighted MR imaging. *Magnetic Resonance in Medicine*, 50, 1077–1088.
- Bergamino, M., Kuplicki, R., Victor, T. A., Cha, Y.-H., & Paulus, M. P. (2017). Comparison of two different analysis approaches for DTI free-water corrected and uncorrected maps in the study of white matter microstructural integrity in individuals with depression. *Human Brain Mapping*, 38, 4690–4702.
- Bora, E., Harrison, B. J., Yücel, M., & Pantelis, C. (2013). Cognitive impairment in euthymic major depressive disorder: A meta-analysis. *Psychological Medicine*, 43, 2017–2026.
- Bracht, T., Jones, D. K., Müller, T. J., Wiest, R., & Walther, S. (2015). Limbic white matter microstructure plasticity reflects recovery from depression. *Journal of Affective Disorders*, 170, 143–149.
- Burcusa, S. L., & Iacono, W. G. (2007). Risk for recurrence in depression. *Clinical Psychology Review*, 27, 959–985.
- Campbell, S., & MacQueen, G. (2004). The role of the hippocampus in the pathophysiology of major depression. *Journal of Psychiatry and Neuroscience*, 29, 417–426.
- Cross, D., Fani, N., Powers, A., & Bradley, B. (2017). Neurobiological development in the context of childhood trauma. *Clinical Psychology*, 24, 111–124.
- Dalby, R. B., Frandsen, J., Chakravarty, M. M., Ahidian, J., Sørensen, L., Rosenberg, R., ... Ostergaard, L. (2010). Depression severity is correlated to the integrity of white matter fiber tracts in late-onset major depression. *Psychiatry Research*, 184, 38–48.
- de Carlo, V., Calati, R., & Serretti, A. (2016). Socio-demographic and clinical predictors of non-response/non-remission in treatment resistant depressed patients: A systematic review. *Psychiatry Research*, 240, 421–430.
- de Diego-Adeliño, J., Pires, P., Gómez-Ansón, B., Serra-Blasco, M., Vives-Gilabert, Y., Puigdemont, D., ... Portella, M. J. (2014). Microstructural white-matter abnormalities associated with treatment resistance, severity and duration of illness in major depression. *Psychological Medicine*, 44, 1171–1182.
- Doolin, K., Andrews, S., Carballedo, A., McCarthy, H., O’Hanlon, E., Tozzi, L., & Frodl, T. (2019). Longitudinal diffusion weighted imaging of limbic regions in patients with major depressive disorder after 6 years and partial to full remission. *Psychiatry Research: Neuroimaging*, 287, 75–86.
- Evans, V. C., Iverson, G. L., Yatham, L. N., & Lam, R. W. (2014). The relationship between neurocognitive and psychosocial functioning in major depressive disorder: A systematic review. *The Journal of Clinical Psychiatry*, 75, 1359–1370.
- Feldman, H. M., Yeatman, J. D., Lee, E. S., Barde, L. H. F., & Gaman-Bean, S. (2010). Diffusion tensor imaging: A review for pediatric researchers and clinicians. *Journal of Developmental and Behavioral Pediatrics: JDBP*, 31, 346–356.
- Ferrari, A. J., Charlson, F. J., Norman, R. E., Patten, S. B., Freedman, G., Murray, C. J. L., ... Whiteford, H. A. (2013). Burden of depressive disorders by country, sex, age, and year: Findings from the global burden of disease study 2010. *PLoS Medicine*, 10, e1001547.
- Gorwood, P., Corruble, E., Falissard, B., & Goodwin, G. M. (2008). Toxic effects of depression on brain function: Impairment of delayed recall and

- the cumulative length of depressive disorder in a large sample of depressed outpatients. *The American Journal of Psychiatry*, 165, 731–739.
- Guo, W.-B., Liu, F., Chen, J.-D., Xu, X.-J., Wu, R.-R., Ma, C.-Q., ... Zhao, J.-P. (2012a). Altered white matter integrity of forebrain in treatment-resistant depression: A diffusion tensor imaging study with tract-based spatial statistics. *Progress in Neuro-Psychopharmacology & Biological Psychiatry*, 38, 201–206.
- Guo, W.-B., Liu, F., Xue, Z.-M., Gao, K., Wu, R.-R., Ma, C.-Q., ... Zhao, J.-P. (2012b). Altered white matter integrity in young adults with first-episode, treatment-naïve, and treatment-responsive depression. *Neuroscience Letters*, 522, 139–144.
- Harada, K., Ikuta, T., Nakashima, M., Watanuki, T., Hirotsu, M., Matsubara, T., ... Matsuo, K. (2018). Altered connectivity of the anterior cingulate and the posterior superior temporal gyrus in a longitudinal study of late-life depression. *Frontiers in Aging Neuroscience*, 10, 31.
- Hovens, J. G. F. M., Giltay, E. J., Wiersma, J. E., Spinthoven, P., Penninx, B. W. J. H., & Zitman, F. G. (2012). Impact of childhood life events and trauma on the course of depressive and anxiety disorders. *Acta Psychiatrica Scandinavica*, 126, 198–207.
- Hua, K., Zhang, J., Wakana, S., Jiang, H., Li, X., Reich, D. S., ... Mori, S. (2008). Tract probability maps in stereotaxic spaces: Analyses of white matter anatomy and tract-specific quantification. *NeuroImage*, 39, 336–347.
- Huang, H., Gundapuneedi, T., & Rao, U. (2012). White matter disruptions in adolescents exposed to childhood maltreatment and vulnerability to psychopathology. *Neuropsychopharmacology*, 37, 2693–2701.
- Jauregui-Huerta, F., Ruvalcaba-Delgadillo, Y., Gonzalez-Castañeda, R., Garcia-Estrada, J., Gonzalez-Perez, O., & Luquin, S. (2010). Responses of glial cells to stress and glucocorticoids. *Current Immunology Reviews*, 6, 195–204.
- Jenkinson, M., Beckmann, C. F., Behrens, T. E. J., Woolrich, M. W., & Smith, S. M. (2012). FSL. *NeuroImage*, 62, 782–790.
- Jia, Z., Huang, X., Wu, Q., Zhang, T., Lui, S., Zhang, J., ... Gong, Q. (2010). High-field magnetic resonance imaging of suicidality in patients with major depressive disorder. *The American Journal of Psychiatry*, 167, 1381–1390.
- Jones, D. K., Knösche, T. R., & Turner, R. (2013). White matter integrity, fiber count, and other fallacies: The do's and don'ts of diffusion MRI. *NeuroImage*, 73, 239–254.
- Kanai, T., Takeuchi, H., Furukawa, T. A., Yoshimura, R., Imaizumi, T., Kitamura, T., & Takahashi, K. (2003). Time to recurrence after recovery from major depressive episodes and its predictors. *Psychological Medicine*, 33, 839–845.
- Keller, M. B., & Boland, R. J. (1998). Implications of failing to achieve successful long-term maintenance treatment of recurrent unipolar major depression. *Biological Psychiatry*, 44, 348–360.
- Kennedy, N., Abbott, R., & Paykel, E. S. (2004). Longitudinal syndromal and sub-syndromal symptoms after severe depression: 10-year follow-up study. *The British Journal of Psychiatry*, 184, 330–336.
- Kerchner, G. A., Racine, C. A., Hale, S., Wilheim, R., Laluz, V., Miller, B. L., & Kramer, J. H. (2012). Cognitive processing speed in older adults: Relationship with white matter integrity. *PLoS One*, 7, e50425.
- Kircher, T., Wöhr, M., Nenadic, I., Schwarting, R., Schrott, G., Alferink, J., ... Dannlowski, U. (2019). Neurobiology of the major psychoses: A translational perspective on brain structure and function—the FOR2107 consortium. *European Archives of Psychiatry and Clinical Neuroscience*, 269, 949–962.
- Kraus, C., Kadriu, B., Lanzenberger, Jr. R., Carlos Jr. A. Z., & Kasper, S. (2019). Prognosis and improved outcomes in major depression: A review. *Translational Psychiatry* 9, 127.
- Le Bihan, D. (2003). Looking into the functional architecture of the brain with diffusion MRI. *Nature Reviews Neuroscience*, 4, 469–480.
- Li, L., Ma, N., Li, Z., Tan, L., Liu, J., Gong, G., ... Xu, L. (2007). Prefrontal white matter abnormalities in young adult with major depressive disorder: A diffusion tensor imaging study. *Brain Research*, 1168, 124–128.
- Madhavan, K. M., McQueeny, T., Howe, S. R., Shear, P., & Szaflarski, J. (2014). Superior longitudinal fasciculus and language functioning in healthy aging. *Brain Research*, 1562, 11–22.
- McEwen, B. S. (2003). Interacting mediators of allostatic and allostatic load: Towards an understanding of resilience in aging. *Metabolism: Clinical and Experimental*, 52, 10–16.
- Meinert, S., Repple, J., Nenadic, I., Krug, A., Jansen, A., Grotegerd, D., ... Dannlowski, U. (2019). Reduced fractional anisotropy in depressed patients due to childhood maltreatment rather than diagnosis. *Neuropsychopharmacology*, 44, 2065–2072.
- Meinert, S., Repple, J., Nenadic, I., Krug, A., Jansen, A., Grotegerd, D., ... Dannlowski, U. (accepted). Changes in white matter fiber structure in depressed patients due to childhood maltreatment rather than diagnosis. *Neuropsychopharmacology*.
- Mori, S., Wakana, S., van Zijl, P. C. M., & Nagae-Poetscher, L. M. (2005). *MRI atlas of human white matter*. Amsterdam, The Netherlands: Elsevier.
- Murphy, M. L., & Frodl, T. (2011). Meta-analysis of diffusion tensor imaging studies shows altered fractional anisotropy occurring in distinct brain areas in association with depression. *Biology of Mood & Anxiety Disorders*, 1, 3.
- Nichols, T. E., & Holmes, A. P. (2002). Nonparametric permutation tests for functional neuroimaging: A primer with examples. *Human Brain Mapping*, 15, 1–25.
- Nobuhara, K., Okugawa, G., Sugimoto, T., Minami, T., Tamagaki, C., Takase, K., ... Kinoshita, T. (2006). Frontal white matter anisotropy and symptom severity of late-life depression: A magnetic resonance diffusion tensor imaging study. *Journal of Neurology, Neurosurgery, and Psychiatry*, 77, 120–122.
- Oberlander, T. F., Weinberg, J., Papsdorf, M., Grunau, R., Misri, S., & Devlin, A. M. (2008). Prenatal exposure to maternal depression, neonatal methylation of human glucocorticoid receptor gene (NR3C1) and infant cortisol stress responses. *Epigenetics*, 3, 97–106.
- Oguz, I., Farzinfar, M., Matsui, J., Budin, F., Liu, Z., Gerig, G., ... Styner, M. (2014). DTIPrep: Quality control of diffusion-weighted images. *Frontiers in Neuroinformatics*, 8, 4.
- Ormel, J., Oldehinkel, A. J., Nolen, W. A., & Vollebergh, W. (2004). Psychosocial disability before, during, and after a major depressive episode: A 3-wave population-based study of state, scar, and trait effects. *Archives of General Psychiatry*, 61, 387–392.
- Otte, C., Gold, S. M., Penninx, B. W., Pariante, C. M., Etkin, A., Fava, M., ... Schatzberg, A. F. (2016). Major depressive disorder. *Nature Reviews Disease Primers*, 2, 16065.
- Paykel, E. S. (2008). Partial remission, residual symptoms, and relapse in depression. *Dialogues in Clinical Neuroscience*, 10, 431–437.
- PDR (2017). *Physicians' desk reference* (71st ed., 2017). Montvale, NJ: PDR Network.
- Penninx, B. W. J. H., Nolen, W. A., Lamers, F., Zitman, F. G., Smit, J. H., Spinthoven, P., ... Beekman, A. T. F. (2011). Two-year course of depressive and anxiety disorders: Results from the Netherlands study of depression and anxiety (NESDA). *Journal of Affective Disorders*, 133, 76–85.
- Post, R. M., Leverich, G. S., Xing, G., & Weiss, R. B. (2001). Developmental vulnerabilities to the onset and course of bipolar disorder. *Development and Psychopathology*, 13, 581–598.
- Redlich, R., Dohm, K., Grotegerd, D., Opel, N., Zwitserlood, P., Heindel, W., ... Dannlowski, U. (2015). Reward processing in unipolar and bipolar depression: A functional MRI study. *Neuropsychopharmacology*, 40, 2623–2631.
- Repple, J., Mauritz, M., Meinert, S., de Lange, S. C., Grotegerd, D., Opel, N., ... van den Heuvel, M. P. (2020). Severity of current depression and remission status are associated with structural connectome alterations in major depressive disorder. *Molecular Psychiatry*, 25, 1550–1558.
- Repple, J., Zaremba, D., Meinert, S., Grotegerd, D., Redlich, R., Förster, K., ... Dannlowski, U. (2019). Time heals all wounds? A 2-year longitudinal diffusion tensor imaging study in major depressive disorder. *Journal of Psychiatry & Neuroscience: JPN*, 44, 407–413.
- Rock, P. L., Roiser, J. P., Riedel, W. J., & Blackwell, A. D. (2014). Cognitive impairment in depression: A systematic review and meta-analysis. *Psychological Medicine*, 44, 2029–2040.
- Salami, A., Eriksson, J., Nilsson, L.-G., & Nyberg, L. (2012). Age-related white matter microstructural differences partly mediate age-related decline in processing speed but not cognition. *Biochimica et Biophysica Acta*, 1822, 408–415.
- Shen, X., Adams, M. J., Ritakari, T. E., Cox, S. R., McIntosh, A. M., & Whalley, H. C. (2019). White matter microstructure and its relation to longitudinal measures of depressive symptoms in mid- and late life. *Biological Psychiatry*, 86, 759–768.

- Smith, S. M. (2002). Fast robust automated brain extraction. *Human Brain Mapping*, 17, 143–155.
- Smith, S. M., Jenkinson, M., Johansen-Berg, H., Rueckert, D., Nichols, T. E., Mackay, C. E., ... Behrens, T. E. J. (2006). Tract-based spatial statistics: Voxelwise analysis of multi-subject diffusion data. *NeuroImage*, 31, 1487–1505.
- Smith, S. M., Jenkinson, M., Woolrich, M. W., Beckmann, C. F., Behrens, T. E. J., Johansen-Berg, H., ... Matthews, P. M. (2004). Advances in functional and structural MR image analysis and implementation as FSL. *NeuroImage*, 23(Suppl 1), S208–S219.
- Smith, S. M., & Nichols, T. E. (2009). Threshold-free cluster enhancement: Addressing problems of smoothing, threshold dependence and localisation in cluster inference. *NeuroImage*, 44, 83–98.
- Solomon, D. A., Keller, M. B., Leon, A. C., Mueller, T. I., Lavori, P. W., Shea, M. T., ... Endicott, J. (2000). Multiple recurrences of major depressive disorder. *The American Journal of Psychiatry*, 157, 229–233.
- Spijker, J., de Graaf, R., Bijl, R. V., Beekman, A. T. F., Ormel, J., & Nolen, W. A. (2004). Determinants of persistence of major depressive episodes in the general population. Results from the Netherlands Mental Health Survey and Incidence Study (NEMESIS). *Journal of Affective Disorders*, 81, 231–240.
- Targum, S. D. (1984). Persistent neuroendocrine dysregulation in major depressive disorder: A marker for early relapse. *Biological Psychiatry*, 19, 305–318.
- Tarullo, A. R., & Gunnar, M. R. (2006). Child maltreatment and the developing HPA axis. *Hormones and Behavior*, 50, 632–639.
- Terenina, E. E., Cavigelli, S., Mormede, P., Zhao, W., Parks, C., Lu, L., ... Mulligan, M. K. (2019). Genetic factors mediate the impact of chronic stress and subsequent response to novel acute stress. *Frontiers in Neuroscience*, 13, 438.
- Tozzi, L., Carballedo, A., Wetterling, F., McCarthy, H., O'Keane, V., Gill, M., ... Frodl, T. (2016). Single-Nucleotide polymorphism of the FKBP5 gene and childhood maltreatment as predictors of structural changes in brain areas involved in emotional processing in depression. *Neuropsychopharmacology*, 41, 487–497.
- Turken, A., Whitfield-Gabrieli, S., Bammer, R., Baldo, J. V., Dronkers, N. F., & Gabrieli, J. D. E. (2008). Cognitive processing speed and the structure of white matter pathways: Convergent evidence from normal variation and lesion studies. *NeuroImage*, 42, 1032–1044.
- Varghese, F. P., & Brown, E. S. (2001). The hypothalamic-pituitary-adrenal axis in Major depressive disorder: A brief primer for primary care physicians. *Primary Care Companion to The Journal of Clinical Psychiatry*, 3, 151–155.
- Vogelbacher, C., Möbius, T. W. D., Sommer, J., Schuster, V., Dannlowski, U., Kircher, T., ... Bopp, M. H. A. (2018). The Marburg-Münster affective disorders cohort study (MACS): A quality assurance protocol for MR neuroimaging data. *NeuroImage*, 172, 450–460.
- Vos, T., Haby, M. M., Barendregt, J. J., Kruijshaar, M., Corry, J., & Andrews, G. (2004). The burden of major depression avoidable by longer-term treatment strategies. *Archives of General Psychiatry*, 61, 1097–1103.
- Wakana, S., Caprihan, A., Panzenboeck, M. M., Fallon, J. H., Perry, M., Gollub, R. L., ... Mori, S. (2007). Reproducibility of quantitative tractography methods applied to cerebral white matter. *NeuroImage*, 36, 630–644.
- Williams, J. M. G., Barnhofer, T., Crane, C., Herman, D., Raes, F., Watkins, E., & Dalgleish, T. (2007). Autobiographical memory specificity and emotional disorder. *Psychological Bulletin*, 133, 122–148.
- Winston, G. P. (2012). The physical and biological basis of quantitative parameters derived from diffusion MRI. *Quantitative Imaging in Medicine and Surgery*, 2, 254–265.
- Wise, T., Radua, J., Nortje, G., Cleare, A. J., Young, A. H., & Arnone, D. (2016). Voxel-based meta-analytical evidence of structural disconnection in major depression and bipolar disorder. *Biological Psychiatry*, 79, 293–302.
- Wittchen, H.-U., Wunderlich, U., Gruschwitz, S., & Zaudig, M. (1997). SKID I. *Strukturiertes klinisches Interview für DSM-IV. Achse I: Psychische Störungen. Interviewheft und beurteilungsheft. Eine deutschsprachige, erweiterte Bearbeitung der amerikanischen Originalversion des SKID I*. Göttingen: Hogrefe.
- Woolrich, M. W., Jbabdi, S., Patenaude, B., Chappell, M., Makni, S., Behrens, T. E. J., ... Smith, S. M. (2009). Bayesian Analysis of neuroimaging data in FSL. *NeuroImage*, 45, S173–S186.
- Wu, F., Tang, Y., Xu, K., Kong, L., Sun, W., Wang, F., ... Liu, Y. (2010). White matter abnormalities in single-episode, medication-naïve, short term duration of illness subjects with major depressive disorder. *Psychiatry Research*, 191, 80–83.
- Zaremba, D., Dohm, K., Redlich, R., Grottegerd, D., Strojny, R., Meinert, S., ... Dannlowski, U. (2018a). Association of brain cortical changes with relapse in patients with major depressive disorder. *JAMA Psychiatry*, 75, 484–492.
- Zaremba, D., Enneking, V., Meinert, S., Förster, K., Bürger, C., Dohm, K., ... Dannlowski, U. (2018b). Effects of cumulative illness severity on hippocampal gray matter volume in major depression: A voxel-based morphometry study. *Psychological Medicine*, 48, 2391–2398.