# Hearing loss meta-analysis

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#### Methods

#### Eligibility Criteria

We included peer-review publications in English, involving patients with bilateral congenital and mixed hearing loss and controls with structural Magnetic Resonance Imaging. We included cross-sectional studies with control groups, that investigated the structural relation between MRI changes and the hearing loss. The most common MRI measures were **volume**, **FA**, **VBM** and **thickness**. Each measure was assignated to a specific ROI and to a big brain area. (eg. HG and superior temporal lobe belong to **temporal lobe**). A total of 59 studies were included, 6 of them contained incomplete information. A total of 2778 patients and 4214 controls.

Notes for inclusion:

- 1. I excluded Xia et al. Chin J Rad, 2008 because I don't understand chinese and it appears to be the same data as Xia et al. Chin J Med Img Tech, 2008.
- 2. Kim et al. Hear Res 2014 used two groups prelingual deaf and post lingual deaf, I used the average for the main table.
- 3. Xia et al. Chin J Med Img Tech, 2008 had a total of 40 patients, two groups 9-12 years and 19-22 years, no controls.
- 4. Zheng et al. Sci Rep, 2017 this variables change; Con rangeLow Con rangeHigh. Why? I didn't find them on the orignal paper.

Effect size direction was directly include in the Cohen's D value by mutiplying by -1 if the effect was decrease and by 1 if it was none of increased. Forests plots were generated form the meta-regression with subgroups left and right. We measure a general regression for white and gray mater by Etiology (congenital and acquired) with subgroups (left and right).

Effects were summarized across studies using the generic inverse-variance weighting method with DerSimonian and Laird random effects. Studies were weighted by 1/SEš (where SE is the standard error). For the effect size we used Hedges'G, wich takes into account the sample size.

$$Hedges'G = \frac{X_1 - X_2}{\sqrt{\frac{(n_1 - 1)s_1^2 + (n_2 - 1)s_2^2}{n_1 + n_2 - 2}}}$$

#### **Assumptions:**

- 1. We assume that the calculation of the cohen's D is correct.
- 2. We assume that the direction of the effect is correct.

3. Variance was estimated using the cohen's D and sample size of each study. Our estimated variance was used for all meta-regressions, therefore we could have and additional bias in-between studies variance and heterogeneity calculations. We should have calculated the effect size from the mean and standard deviation from each study. Variance was estimated using the following formula:

$$Variance = \frac{n1+n2}{n1\times n2} + \frac{Hedges'G^2}{2\times (n1+n2-2)}$$

#### Estimation of heterogeneity per model

We estimated heterogeneity in results using the  $\tau$  statistic, which represents the standard deviation in the meta-regression models, we used the heterogeneity test x2 and I2.

We performed a multi-level meta-analytic model, over our multiple effect size estimates nested withing variables: Etiology, side and Big brain area. We expected that the underlying true effects are more similar for the same level of the grouping variables than thrue effects arising from different levels.

We can account for the correlation in the true effects by adding a random effect to the model at the level corresponding to the grouping variable.

The dataset contains the result from 54 studies, each comparing different measurements between patients and controls. The difference of between groups was quantified in terms of Hedges'G and Cohen's D.

#### Total included studies

Table 1: Total unique studies 54

	Hearing Loss	Healthy
Total number of patients	2778	4214
Number mean	47.08	71.42
Number sd	128.5	250.6
Age mean	33.07	30.96
m Age~SD	22.66	20.5
%Female mean	50.02	55.71
%Female sd	12.05	12.78

Table 2: Acquired studies 13

	Hearing Loss	Healthy
Total number of patients	1766	3146
Number mean	110.4	196.6
Number sd	239	468.2
Age mean	64.6	55.27
Age~SD	7.863	13.85
%Female mean	45.1	56.52
%Female sd	15.09	12.9

Table 3: Congenital studies 41

	Hearing Loss	Healthy
Total number of patients	1012	1068
Number mean	23.53	24.84
Number sd	17.18	15.04
Age mean	21.06	20.99
$\widetilde{\mathrm{Age}}\ \mathrm{SD}$	12.48	13.06
%Female mean	51.83	55.41
%Female sd	10.37	12.88

# Table of included studies (Figure 1.A)

Table 4: Studies with incomplete information (NA)

	Source	MRI Tesla	all.techniques	all.measures
5	2006, Kara et al. J Neuroradiol	1.5	VBM	length, Thickness, volume
9	2008, Xia et al. Chin J Med Img	1.5	VBM	volume
12	2010, Husain et al. Brain Res	3	DTI, VBM	FA, volume
<b>26</b>	2014, Lyness et al. Neuroimage	1.5	DTI	FA, MD, RD
41	2017, Karns et al. Hear Res	3	DTI	AD, FA, RD, volume
48	2018, Kumar U, Mishra M. Brain Res	3	VBM	Thickness, VBM

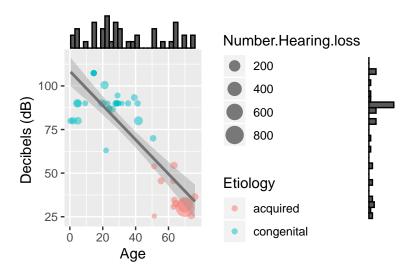
Table 5: Included studies

	Source	MRI Tesla	all.techniques	all.measures
1	2000, Bavelier et al. J Neurosci	1.5	VBM	volume
2	2003, Emmorey et al. PNAS	1.5	VBM	asymmetry, GM+WM, ratio GM/WM, volume
3	2003, Penhune et al. Neuroimage	1.5	VBM	asymmetry, ratio GM/WM, volume
4	2004, Chang et al. Neuroreport	3	DTI	asymmetry, FA
6	2007, Meyer et al. Restor Neurol Neurosci	3	VBM	volume
7	2007, Shibata DK. Am J Neuroradiol	1.5	VBM	volume
8	2008, Allen et al. J Neurosci	1.5	VBM	asymmetry, ratio GM/WM, Vo
10	2009, Kim et al. Neuroreport	3	DTI, VBM	FA, volume
11	2009, Wang et al. Chin J Med Img Tech	3	DTI	FA
13	2010, Leporé et al. Hum Brain Mapp	1.5	VBM	VBM
14	2010, Li, et al. J Clin Rad	1.5	VBM	volume
15	2010, Liu et al. Chin J Med Img Tech	3	CT	FA
16	2011, Smith et al. Cereb Cortex	3	VBM	asymmetry, ratio GM/WM, volume
17	2012, Li et al. Brain Res	3	CT	Thickness
18	2012, Li et al. Hum Brain Mapp	3	DTI	AD, FA, RD
19	2013, Allen et al. Front Neuroanat	1.5	VBM	asymmetry, volume
20	2013, Boyen et al. Hear Res	3	VBM	volume
21	2013, Miao et al. Am J Neuroradiol	3	DTI	FA, RD
<b>22</b>	2013, Pénicaud et al. Neuroimage	1.5	VBM	volume
23	2014, Hribar et al. Hear Res	3	DTI, VBM	AD, FA, Thickness
<b>24</b>	2014, Kim et al. Hear Res	3	VBM	volume
<b>25</b>	2014, Lin et al. Neuroimage	1.5	VBM	volume
27	2014, Olulade et al. J Neurosci	3	VBM	volume
28	2014, Profant et al. Neuroscience	3	DTI, VBM	AD, CT, FA, MD, RD, Surface volume
29	2014, Profant et al. Neuroscience	3	DTI, VBM	AD, CT, FA, MD, RD, Surface volume
30	2015, Huang et al. PLoS One	1.5	DTI	FA, MD
31	2015, Tae Investig Magn Reson Imaging	1.5	VBM	VBM
32	2016, Amaral et al. Eur J Neurosci	3	VBM	asymmetry, Thickness
33	2016, Chinnadurai et al. Magn Reson Imaging	1.5	DTI	AD, Axial Kurtosis, FA, Mean Kurtosis, Radial Kurtosis, RD
34	2016, Ma et al. AJNR Am J Neuroradiol	3	DTI	AD, FA, MD, RD
35	2016, Shi et al. Neuroreport	3	VBM	volume
36	2016, Shiell et al. Neural Plasticity	3	CT	Thickness
37	2016, Smittenaar et al. Open Neuroimag J	1.5	CT	CT

	Source	MRI Tesla	all.techniques	all.measures
38	2016, Wu et al. Brain Res	1.5	VBM	ADC, FA
39	2016, Wu et al. Brain Res	1.5	VBM	ADC, FA
40	2016, Wu et al. Brain Res	1.5	VBM	ADC, FA
$\bf 42$	2017, Kim et al. Neuroreport	3	DTI	FA
43	2017, Shiell & Zatorre. Hear Res	3	DTI	AD, MD, RD, volume
44	2017, Zheng et al. Sci Rep	3	DTI	FA, Mean Kurtosis
45	2018, Benetti et al. Neuroimage	4	DTI	AD, FA, RD
46	2018, Chen et al. Behav Neurosci	3	VBM	volume
47	2018, Feng et al. PNAS	3	VBM	VBM
<b>49</b>	2018, Park et al. Biomed Res Int	3	DTI	FA
50	2018, Pereira-Jorge et al. Neural Plast	1.5	VBM	volume
<b>51</b>	2018, Ren et al. Front Neurosci	3	CT, VBM	Thickness, volume
<b>52</b>	2018, Uchida et al. Front Aging Neurosci	3	VBM	volume
53	2018, Uchida et al. Front Aging Neurosci	3	VBM	volume
$\bf 54$	2018, Zou et al. Otol Neurotol	3	DTI	AK, FA, MK, RK
55	2019, Belkhiria et al. Front. Aging Neurosci	3	VBM	CT, volume
56	2019, Belkhiria et al. Front. Aging Neurosci	3	VBM	CT, volume
57	2019, Luan et al. Front Neurosci	3	DTI, VBM	FA, MD, volume
58	2019, Ponticorvo et al. Hum Brain Mapp	3	VBM	volume
59	2019, Xu et al. J Magn Reson Imaging	3	VBM	volume

# Relation between hearing loss (dB) and age (Figure 2.D)

# Hearing loss vs A



# Studies characteristics (Figure 2.E, 2.F)

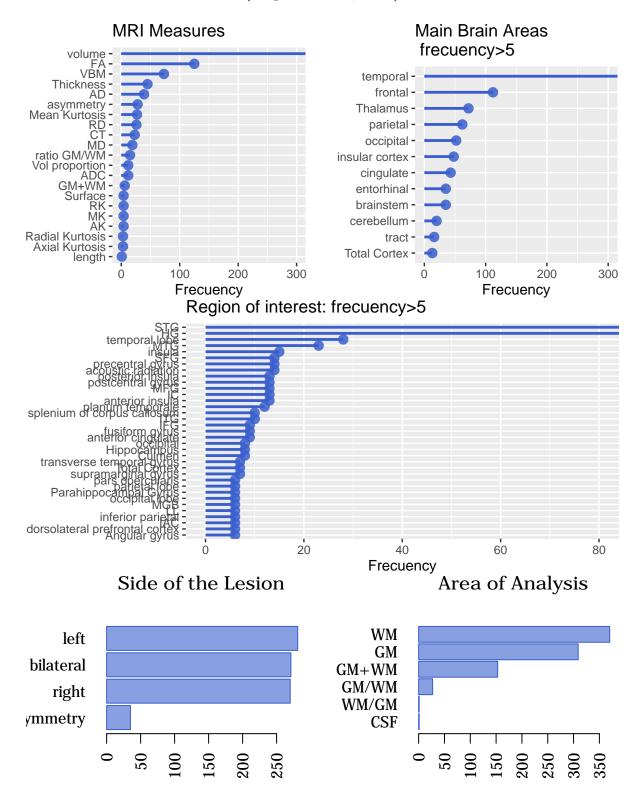
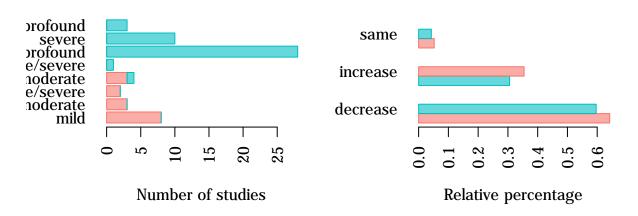


Fig.2.E – Severity

Fig.2.F – Effect direction



### Brain structure (GM, WM) and MRI measures

**Highlights** a. Most of the studies that measured Gray matter focus on cortical changes (volume, thicknes and VBM).

- b. White matter studies are more heterogeneous in their measurements.
- c. Diffusion tensor (DT) derived mesurements are the most frequent in white matter, followed by volume. c.1 It is harder to interpret a meta-analysis of multiple white matter measurements because its effect varies

widely in different directions. The measurements derived from DT have the most differences. We conduct our meta-analysis using the **TWO** most frequent measurements for gray and white matter. We use *volume* for GM and *fractional anysotropy* for WM.

Further meta regressions can be found in the supplementary material.

#### **Gray Matter**

- thickness
- VBM

#### White Matter integrity

- mean diffusivity MD
- radial diffusivity RD
- axial diffusivity AD
- mean kurtosis

#### White Matter volume

- thickness (I am unsure how they did this)
- VBM
- volume

#### Biletareal - GM volume

- WM volume
- WM fractional anisotropy

#### Frequency table: Brain structure (GM, WM) and MRI measures

Table 6: Matter vs measure (continued below)

	AD	ADC	AK	asymmetry	Axial Kurtosis	CT	FA	GM+WM
GM WM	0 39	0 12	$\frac{2}{2}$	9	0 3	23 0	8 117	0

Table 7: Table continues below

	length	MD	Mean Kurtosis	MK	Radial Kurtosis	ratio $\mathrm{GM}/\mathrm{WM}$	RD
GM	0	2	0	2	0	0	0
$\mathbf{W}\mathbf{M}$	1	17	27	2	3	0	26

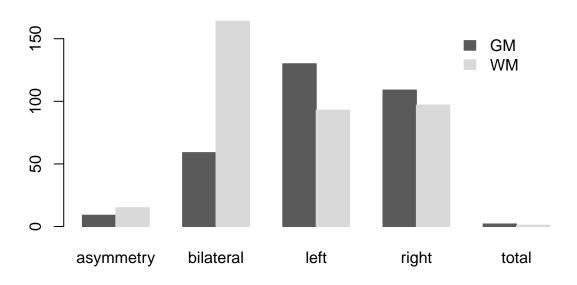
	RK	Surface	Thickness	VBM	Vol proportion	volume
$rac{\mathbf{G}\mathbf{M}}{\mathbf{W}\mathbf{M}}$	$\frac{2}{2}$	4 0	14 10	43 16	6 6	194 79

Table 9: Matter vs Side

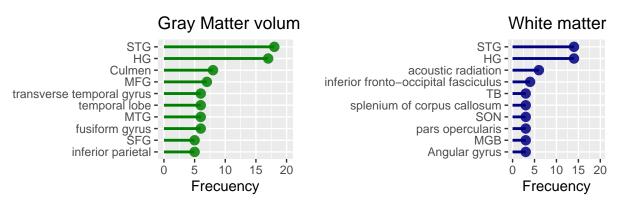
	asymmetry	bilateral	left	right	total
GM	9	59	130	109	2
$\mathbf{W}\mathbf{M}$	15	164	93	97	1

## Brain structure (GM, WM) and side

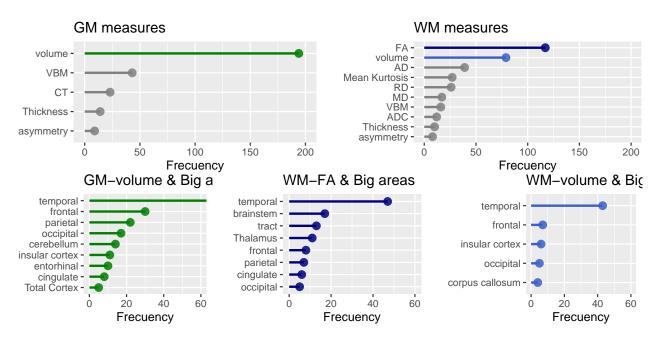
#### **Matter vs Side**



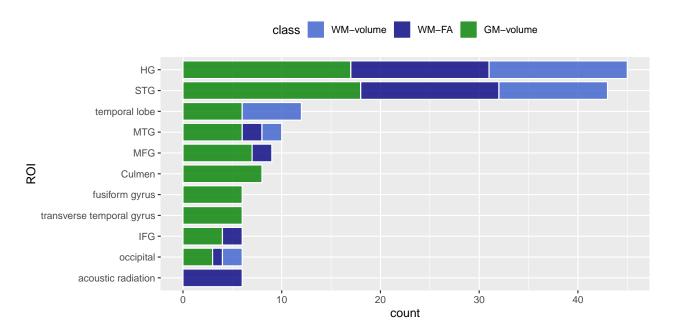
Brain structure (GM, WM) by MRI measure (volume) and ROI



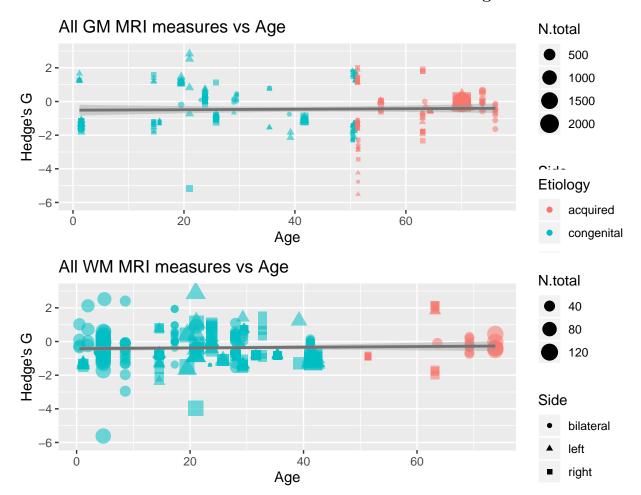
# Studies characteristics (Figure 2.A, 2.B): Brain structure (GM, WM) by MRI measure (volume and FA)



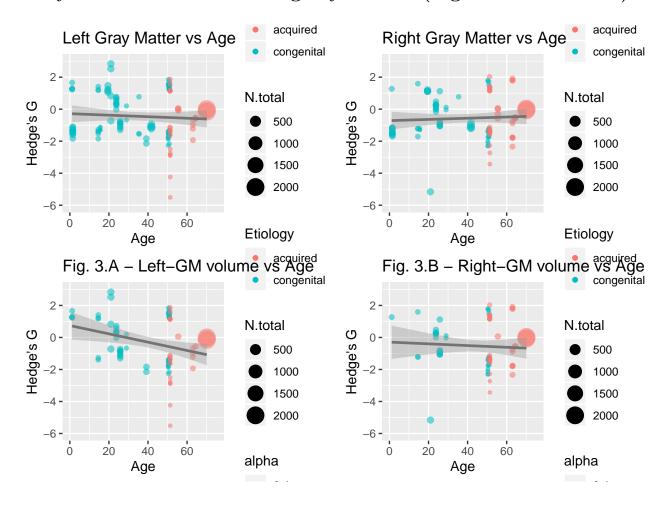
# MRI measures by ROI (Figure 2.C)



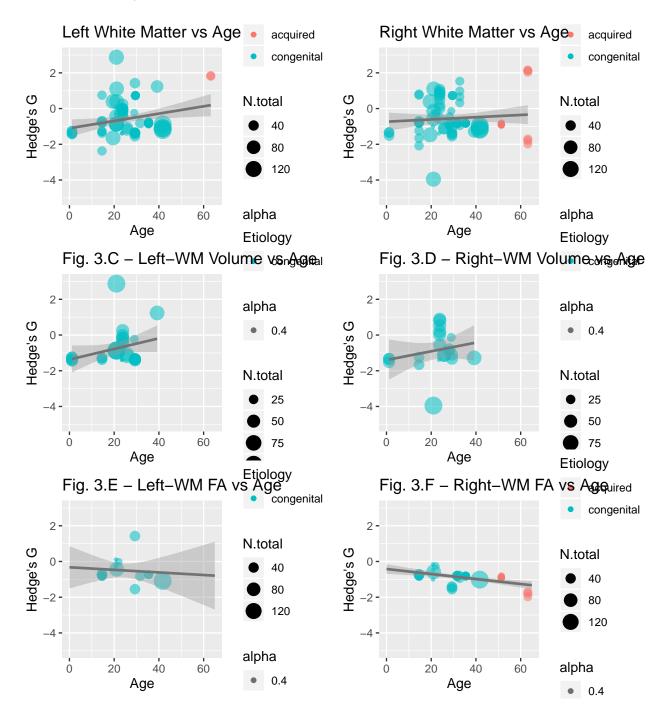
# Relations of all MRI measurements of GM and WM with age



## Gray matter relation with Age by volume (Figures 3.A and 3.B)



# White matter relation with Age by volume and FA (Figures 3.C, 3.D and 3.F)

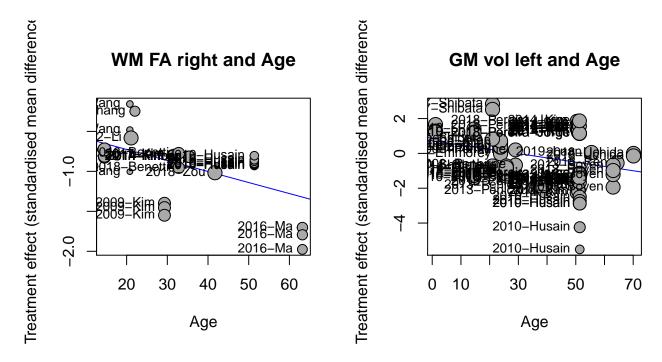


## Gray and White matter relation with Age by asymmetry



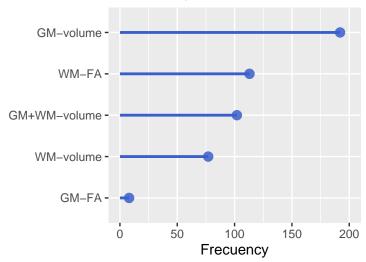
Table of estimates and meta-regression: WM and GM relation with age by MRI measures (volume and FA)

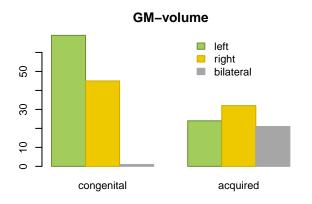
Model	r	p-value	t.stat	df
GM.vol.L	-0.27	0.0103	-2.62	85
WM.vol.L	0.26	0.1687	1.41	28
WM.fa.L	-0.09	0.7393	-0.34	13
GM.vol.R	-0.07	0.5343	-0.62	69
WM.vol.R	0.23	0.316	1.03	19
WM.fa.R	-0.55	2e-04	-4.04	38

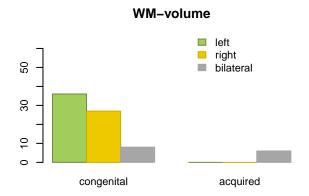


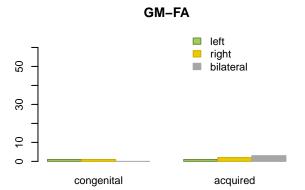
# Meta-regression: Variables by Etiology, Brain matter and MRI measure

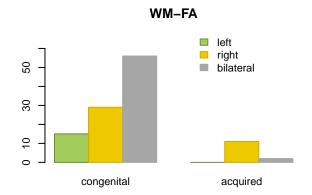
# Variables per brain matter MRI me

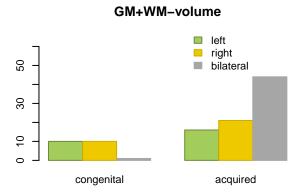












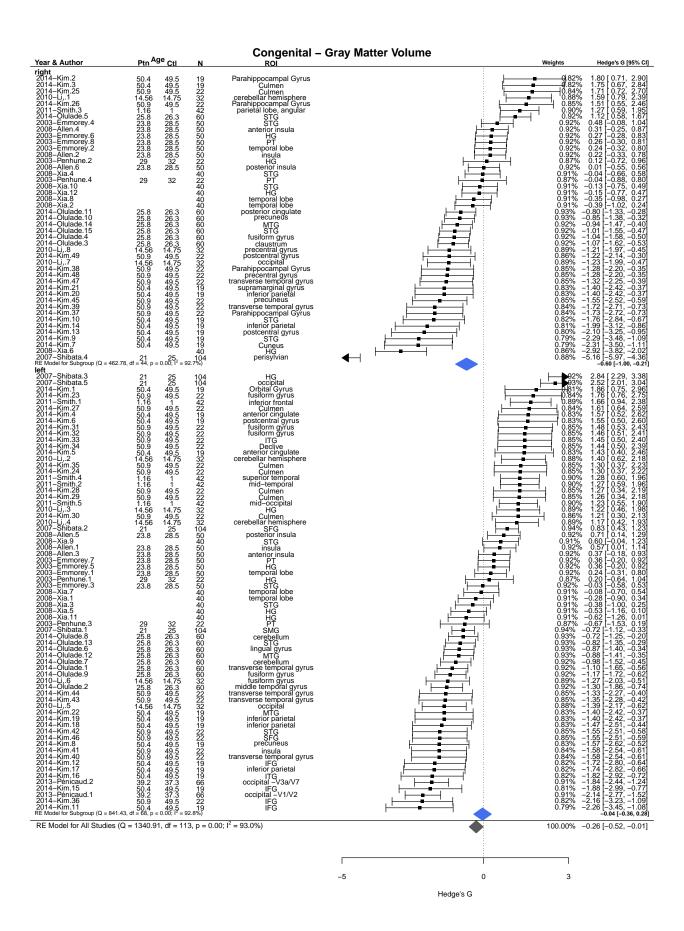
# Gray Matter Volume: Random effects model no intercept covariated by Big area

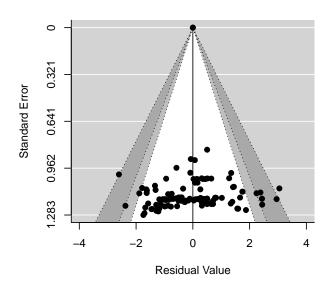
Table 11: Table continues below

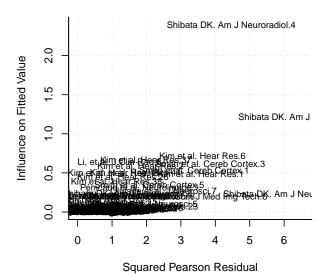
	${\it HedgeG}$	se	zval	ci.lo	ci.up
left cerebellum	0.9013	0.3735	2.413	0.1693	1.633
left cingulate	1.5	0.9037	1.66	-0.2712	3.271
left frontal	-0.588	0.4468	-1.316	-1.464	0.2877
left insular cortex	0.0628	0.6065	0.1035	-1.126	1.252
left occipital	-0.5252	0.4567	-1.15	-1.42	0.3699

	HedgeG	se	zval	ci.lo	ci.up
left parietal	-0.8875	0.5149	-1.724	-1.897	0.1217
left temporal	-0.116	0.2235	-0.5189	-0.554	0.3221
left Thalamus	1.282	1.213	1.056	-1.097	3.66
right cerebellum	1.682	0.7284	2.309	0.254	3.109
right cingulate	-0.8018	1.193	-0.6721	-3.14	1.536
right entorhinal	0.05865	0.6339	0.09251	-1.184	1.301
right frontal	-2.559	0.7143	-3.583	-3.959	-1.159
right insular cortex	-0.1339	0.598	-0.2239	-1.306	1.038
right occipital	-1.73	0.8957	-1.932	-3.486	0.02545
right parietal	-1.113	0.4445	-2.503	-1.984	-0.2413
right temporal	-0.5427	0.2729	-1.989	-1.078	-0.007815

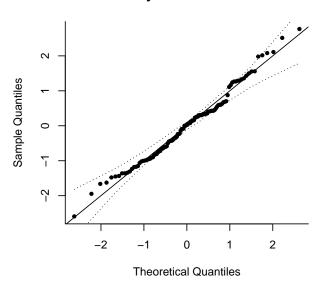
	pval
left cerebellum	0.01581
left cingulate	0.09694
left frontal	0.1882
left insular cortex	0.9175
left occipital	0.2502
left parietal	0.08478
left temporal	0.6039
left Thalamus	0.2909
right cerebellum	0.02096
right cingulate	0.5015
right entorhinal	0.9263
right frontal	0.0003399
right insular cortex	0.8228
right occipital	0.05341
right parietal	0.01233
${f right\ temporal}$	0.04675







#### **Gray Matter Volume**



# ACQUIRED - Meta-regressions of Gray Matter by Volume

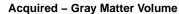
Random effects model no intercept covariated by Big area

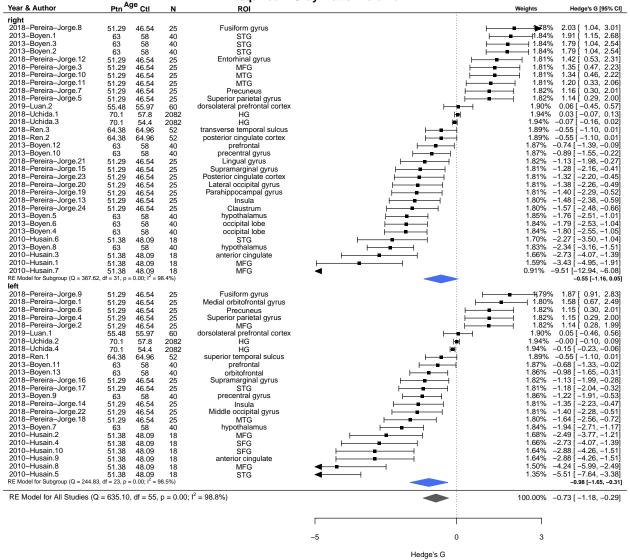
Table 13: Table continues below

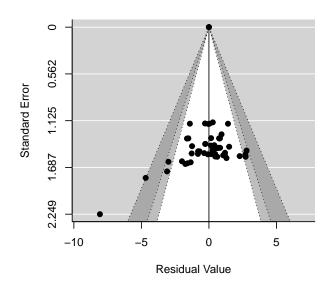
	HedgeG	se	zval	ci.lo	ci.up
left cingulate	-2.883	1.728	-1.669	-6.269	0.5024
left frontal	-1.14	0.5268	-2.164	-2.173	-0.1076
left hypothalamus	-1.937	1.626	-1.191	-5.124	1.25
left insular cortex	-1.353	1.641	-0.8248	-4.57	1.863
left occipital	-1.398	1.642	-0.8514	-4.616	1.82
left parietal	0.3896	0.9454	0.4121	-1.463	2.243
left temporal	-0.8302	0.6236	-1.331	-2.052	0.3921
right cingulate	-1.483	0.954	-1.554	-3.352	0.3872

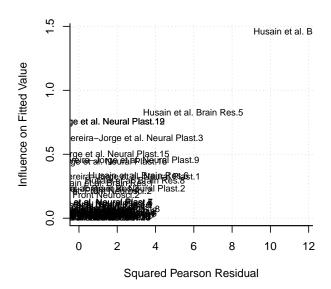
	HedgeG	se	zval	ci.lo	ci.up
right entorhinal	0.007072	1.161	0.006091	-2.269	2.283
right frontal	-1.438	0.7012	-2.05	-2.812	-0.06331
right hypothalamus	-2.047	1.151	-1.778	-4.304	0.2095
right insular cortex	-1.525	1.163	-1.311	-3.803	0.7542
right occipital	-1.524	0.8158	-1.868	-3.123	0.07522
right parietal	0.3405	0.9459	0.36	-1.513	2.194
right temporal	0.727	0.5141	1.414	-0.2806	1.735

	pval
left cingulate	0.09509
left frontal	0.03045
left hypothalamus	0.2335
left insular cortex	0.4095
left occipital	0.3945
left parietal	0.6803
left temporal	0.1831
right cingulate	0.1202
right entorhinal	0.9951
right frontal	0.04034
right hypothalamus	0.0754
right insular cortex	0.1898
right occipital	0.0618
right parietal	0.7189
right temporal	0.1573

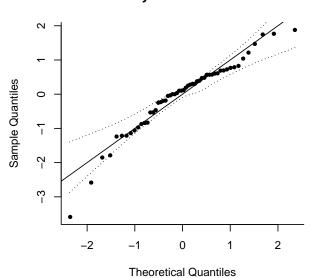








## **Gray Matter Volume**



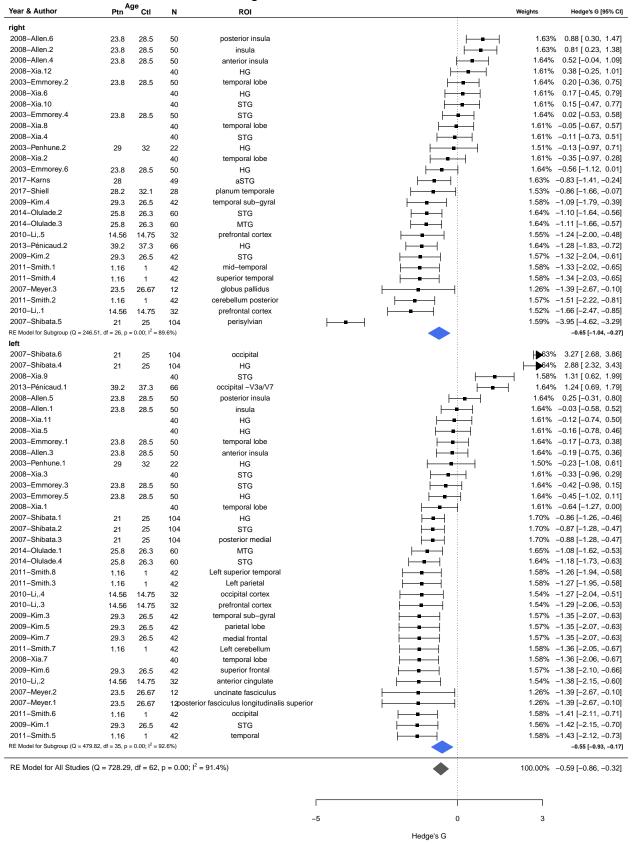
## CONGENITAL - White Matter by VOLUME

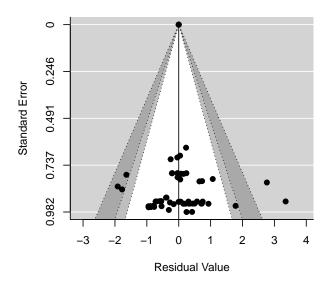
Random effects model no intercept covariated by Big area

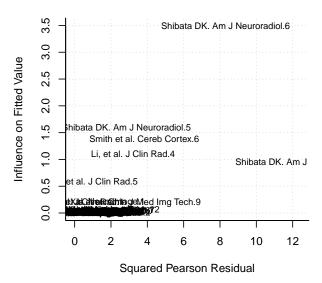
	HedgeG	se	zval	ci.lo	ci.up	N
left cerebellum	-1.107	0.6745	-1.641	-2.429	0.2149	2
left cingulate	-1.379	0.9926	-1.389	-3.324	0.5668	1
left frontal	-1.34	0.5684	-2.358	-2.454	-0.2262	3
left insular cortex	0.007913	0.5504	0.01438	-1.071	1.087	3
left occipital	0.5024	0.4846	1.037	-0.4475	1.452	4
left parietal	-1.308	0.6914	-1.892	-2.663	0.04705	2
left temporal	-0.478	0.2211	-2.163	-0.9113	-0.04478	19
left tract	-1.386	0.7931	-1.747	-2.94	0.1688	2
right cerebellum	-1.513	0.9789	-1.546	-3.432	0.4051	1
right forebrain	-1.386	1.122	-1.235	-3.584	0.8126	1

	HedgeG	se	zval	ci.lo	ci.up	N
right frontal	-2.31	0.5697	-4.055	-3.426	-1.193	3
right insular cortex	0.737	0.5521	1.335	-0.3451	1.819	3
right temporal	-0.5529	0.2218	-2.493	-0.9875	-0.1183	19

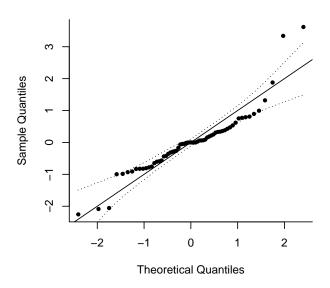
#### **Congenital White Matter Volume**







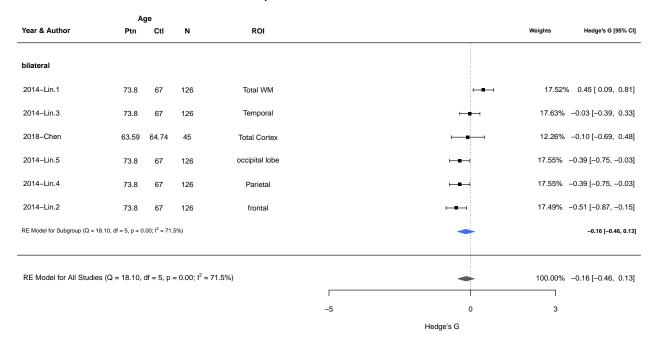
#### **Congenital White Matter Volume**

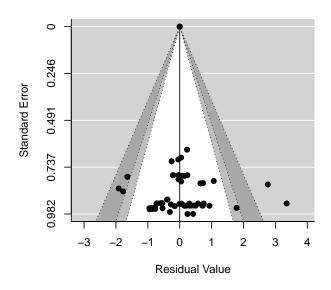


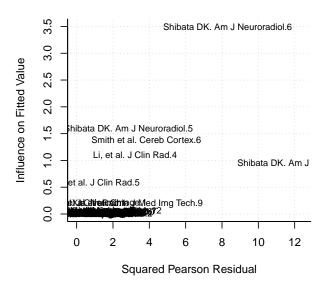
ACQUIRED - White Matter by VOLUME

Not enough values for the Random effects model no intercept covariated by Big area and Side (left or right)

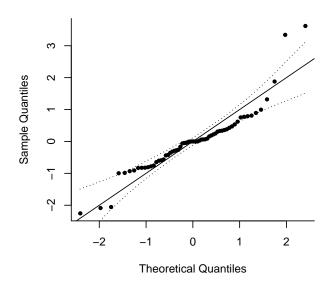
#### acquired White Matter Volume







#### acquired White Matter Volume



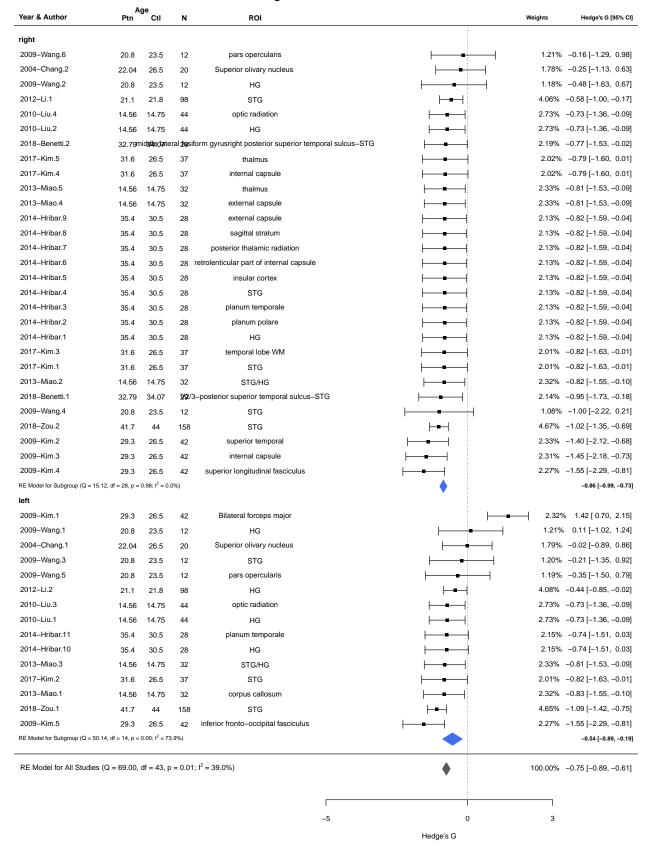
## CONGENITAL - White Matter by FA fractional anisotropy

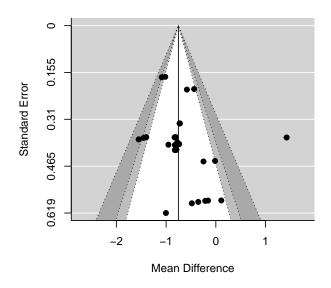
Random effects model no intercept covariated by Big area

	${\it HedgeG}$	se	zval	ci.lo	ci.up	N
left brainstem	-0.01557	0.4891	-0.03183	-0.9742	0.9431	1
left cingulate	0.297	0.2965	1.002	-0.2841	0.8781	2
left occipital	-0.7254	0.3791	-1.914	-1.468	0.01755	1
left temporal	-0.698	0.1265	-5.518	-0.946	-0.4501	10
left tract	-1.549	0.425	-3.645	-2.382	-0.7163	1
right brainstem	-0.2476	0.4908	-0.5044	-1.21	0.7145	1
right insular cortex	-0.8178	0.4415	-1.852	-1.683	0.04759	1
right occipital	-0.7254	0.3791	-1.914	-1.468	0.01755	1
right temporal	-0.8298	0.1036	-8.013	-1.033	-0.6269	16
right Thalamus	-0.9238	0.1789	-5.164	-1.274	-0.5732	6

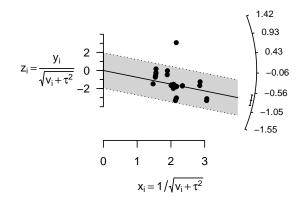
	${\it HedgeG}$	se	zval	ci.lo	ci.up	N
right tract	-1.004	0.2156	-4.656	-1.427	-0.5813	4

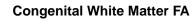
#### **Congenital White Matter FA**

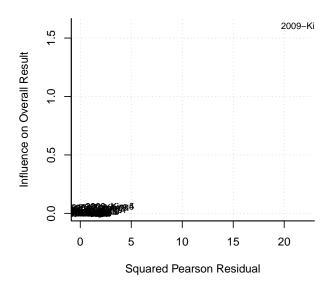


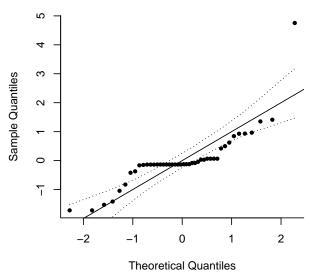


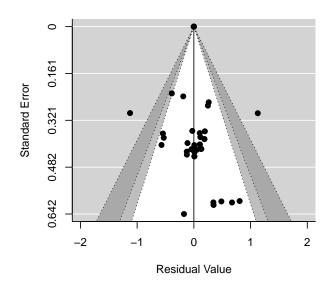
# Congenital White Matter FA

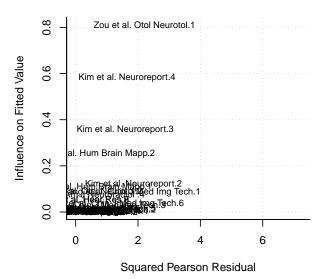




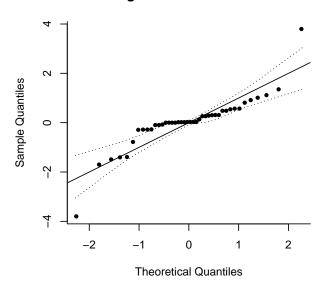








#### **Congenital White Matter FA**

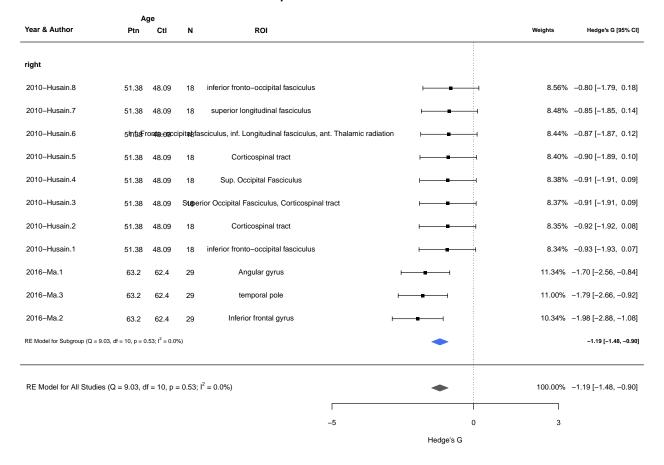


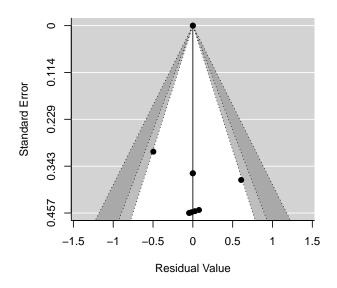
# ACQUIRED - White Matter by FA fractional anisotropy

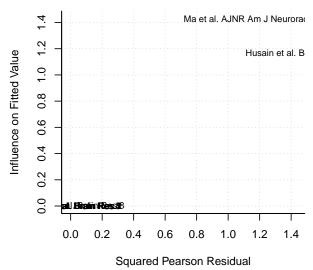
Random effects model no intercept covariated by Big area

	HedgeG	se	zval	ci.lo	ci.up	N
right frontal	-1.48	0.3403	-4.35	-2.147	-0.8134	2
right occipital	-0.9105	0.3603	-2.527	-1.617	-0.2042	2
right parietal	-1.703	0.4379	-3.888	-2.561	-0.8443	1
right temporal	-1.793	0.4446	-4.034	-2.665	-0.922	1
right tract	-0.8812	0.2272	-3.878	-1.326	-0.4359	5

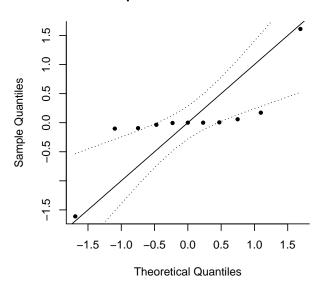
#### acquired White Matter FA





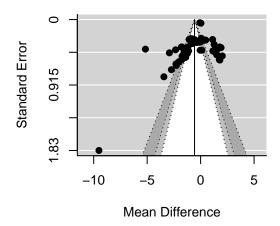


## acquired White Matter FA

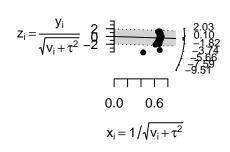


# Supplementary material: heterogeneity per model

## Heterogeney: GM volume Right

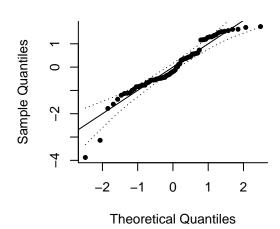


# **GM** volume Right

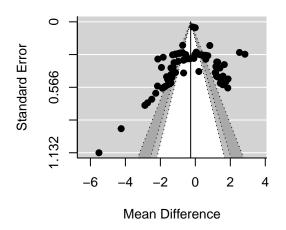


# 

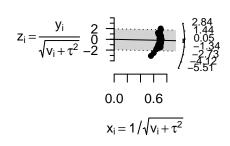
# **GM** volume Right



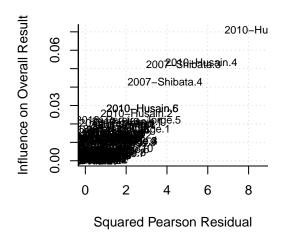
## Heterogeney: GM volume Left

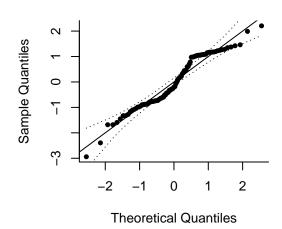


#### **GM volume Left**

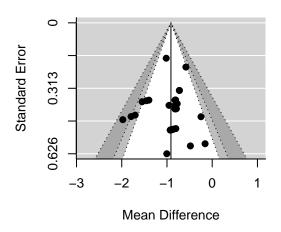


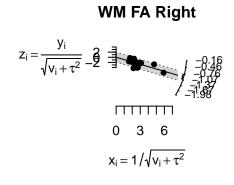
# **GM volume Left**

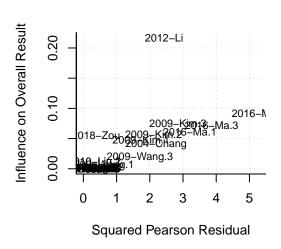


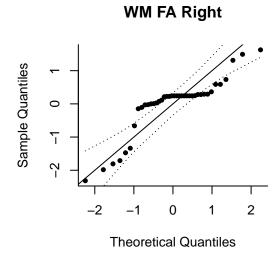


# Heterogeney: WM FA Right

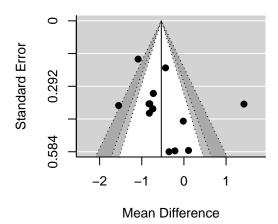




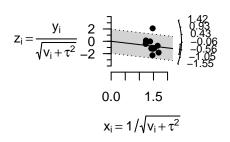




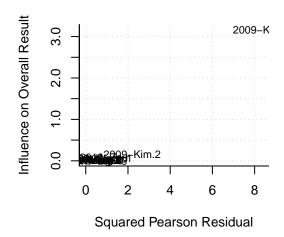
# Heterogeney: WM FA Left

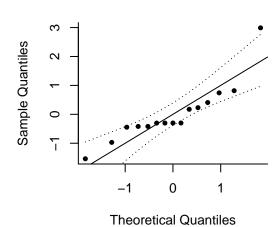


# **WM FA Left**

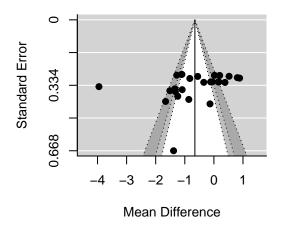


# WM FA Left

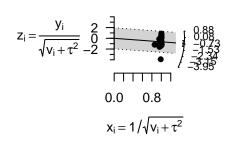




# Heterogeney: WM volume Right

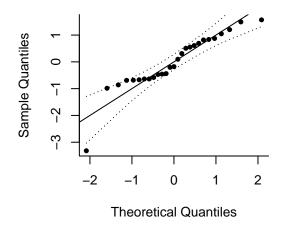


# **WM volume Right**

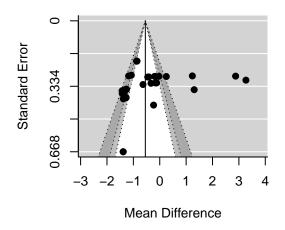


# Influence on Overall Result Re

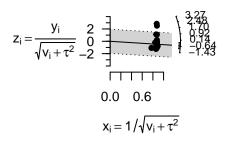
# **WM** volume Right



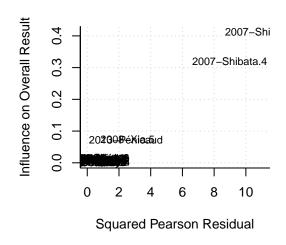
# Heterogeney: WM volume Left

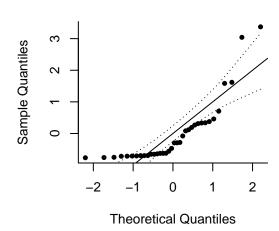


## **WM volume Left**

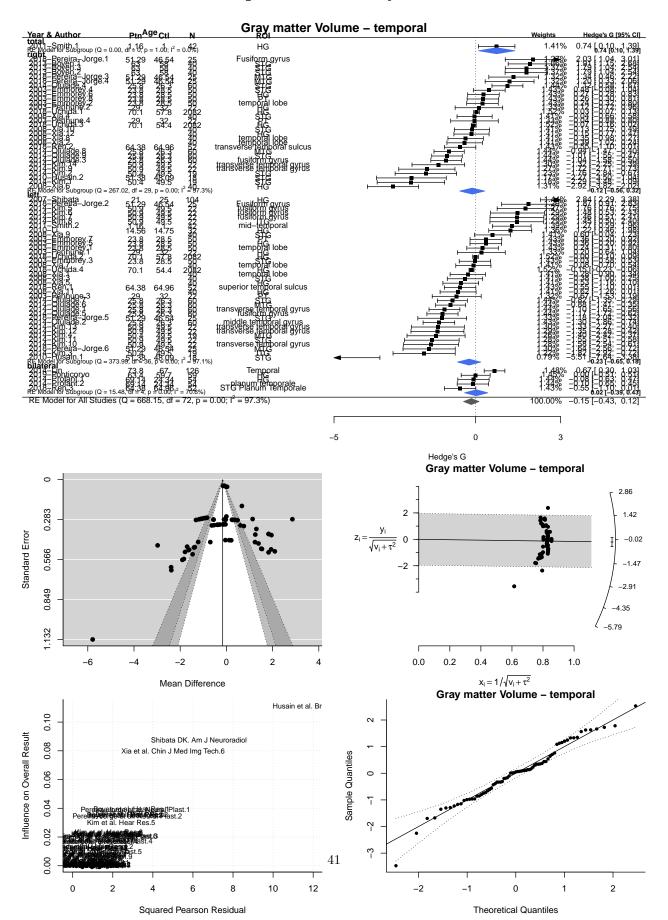


# **WM** volume Left

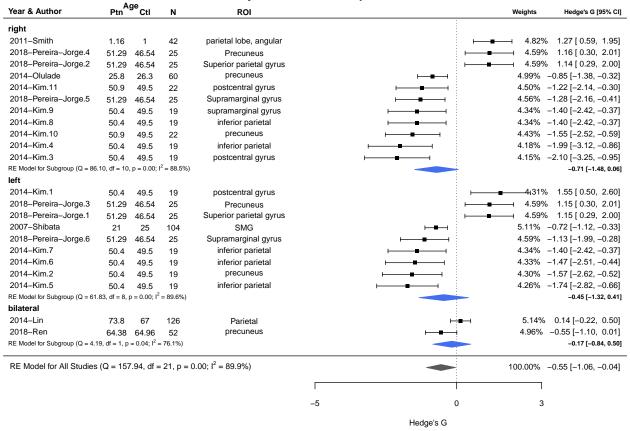


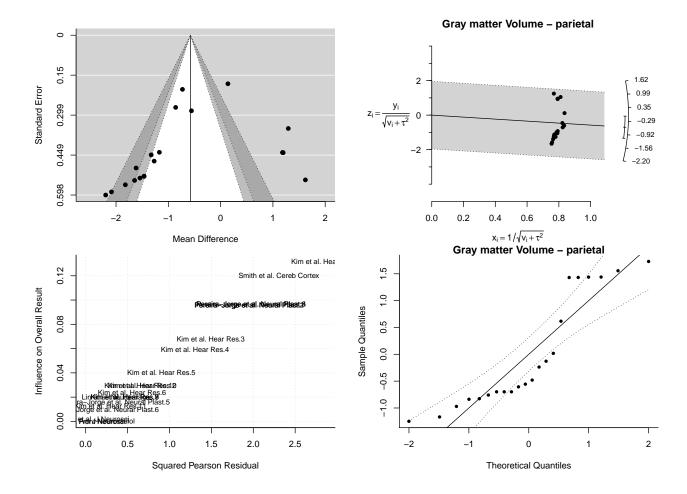


# Meta-regressions of Gray Matter Volume & Brain Areas: Random effects model no intercept covariated by Side

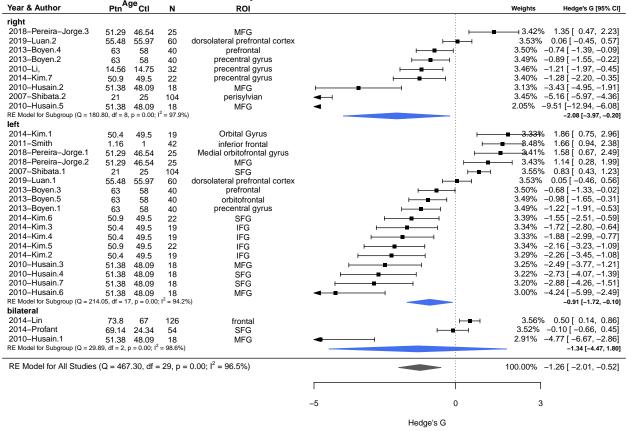


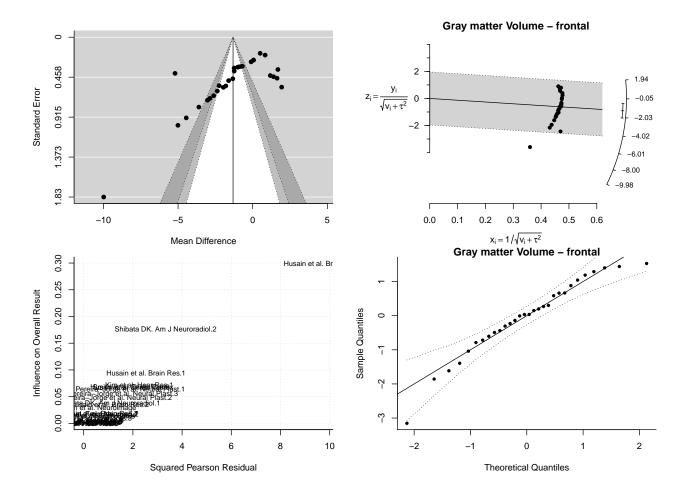
#### Gray matter Volume - parietal





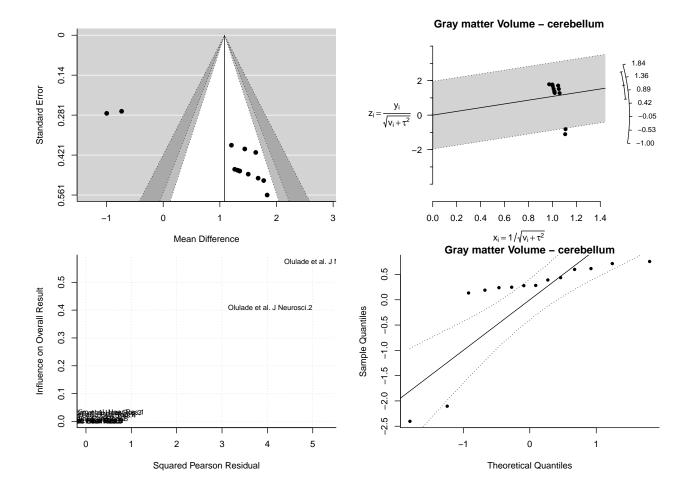
#### **Gray matter Volume - frontal**





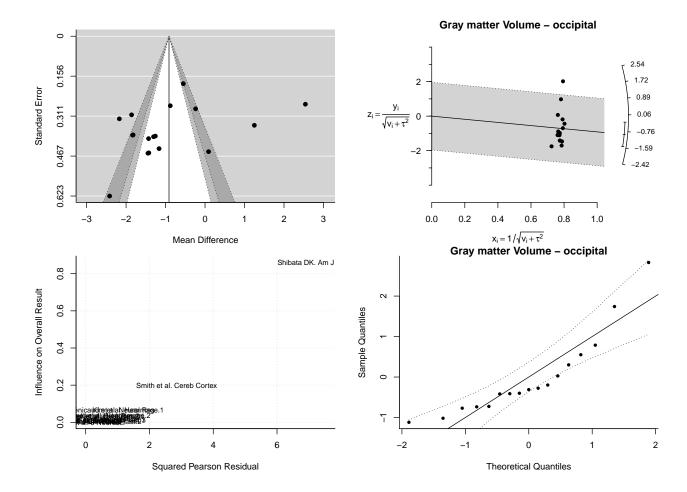
# **Gray matter Volume – cerebellum**

Vara O. Austhan	_ A	ge						
Year & Author	Ptn	Ctl	N	ROI			Weights	Hedge's G [95% CI]
right								
2014-Kim.1	50.4	49.5	19	Culmen			<del>- 6.31</del> %	1.75 [ 0.67, 2.84]
2014-Kim.3	50.9	49.5	22	Culmen			<del>- 6.6</del> 8%	1.71 [ 0.72, 2.70]
2010-Li,.1	14.56	14.75	32	cerebellar hemisphere			<b></b> -7.36%	1.59 [ 0.79, 2.39]
RE Model for Subgroup (Q =	= 0.06, df = 2, p	= 0.97; I <sup>2</sup> =	= 0.0%)					1.67 [1.13, 2.21]
left								
2014-Kim.4	50.9	49.5	22	Culmen			<del>- 6</del> ,74%	1.61 [ 0.64, 2.59]
2014-Kim.8	50.9	49.5	22	Declive			<b></b> 6.83%	1.44 [ 0.50, 2.39]
2010-Li,.2	14.56	14.75	32	cerebellar hemisphere			<b>⊢</b> 7.44%	1.40 [ 0.62, 2.18]
2014-Kim.9	50.9	49.5	22	Culmen			<b>⊢</b> 6.90%	1.30 [ 0.37, 2.23]
2014-Kim.2	50.9	49.5	22	Culmen			⊢ 6.90%	1.30 [ 0.37, 2.22]
2014-Kim.5	50.9	49.5	22	Culmen			<b></b> 6.92%	1.27 [ 0.34, 2.19]
2014-Kim.6	50.9	49.5	22	Culmen			<b></b> 6.92%	1.26 [ 0.34, 2.18]
2014-Kim.7	50.9	49.5	22	Culmen			<b></b> 6.94%	1.21 [ 0.30, 2.13]
2010-Li,.3	14.56	14.75	32	cerebellar hemisphere			<del> 1</del> 7.52%	1.17 [ 0.42, 1.93]
2014-Olulade.2	25.8	26.3	60	cerebellum		<del></del>	8.27% -	0.72 [-1.25, -0.20]
2014-Olulade.1	25.8	26.3	60	cerebellum		<b>⊢</b> ■	8.23% -	0.98 [-1.52, -0.45]
RE Model for Subgroup (Q =	= 80.79, df = 10,	, p = 0.00;	I <sup>2</sup> = 82.9%)					0.88 [0.30, 1.46]
RE Model for All Stud	lies (Q = 96.	30, df =	13, p = 0	0.00; I <sup>2</sup> = 80.5%)			100.00%	1.04 [ 0.54, 1.53]
							:	
					-5	(	3	
						Hedge's G		



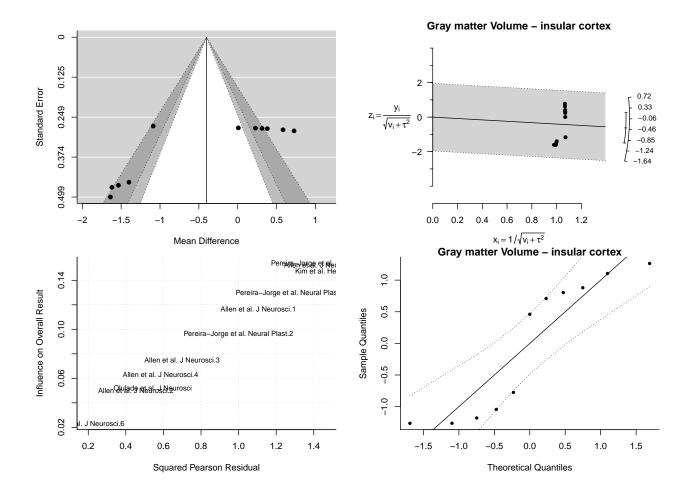
# Gray matter Volume – occipital

Year & Author	A Ptn	ge Ctl	N	ROI			Weight	s Hedge's G [95% CI]
right								
2018-Pereira-Jorge.2	51.29	46.54	25	Lingual gyrus		<b>⊢</b>	⊣ 5.72	% -1.13 [-1.98, -0.27]
2010-Li,.3	14.56	14.75	32	occipital		<del></del>	5.86	% -1.23 [-1.99, -0.47]
2018-Pereira-Jorge.1	51.29	46.54	25	Lateral occipital gyrus		<del></del>	5.67	% -1.38 [-2.26, -0.49]
2013-Boyen.2	63	58	40	occipital lobe		<del></del>	5.87	% -1.79 [-2.53, -1.04]
2013-Boyen.1	63	58	40	occipital lobe		<del></del>	5.87	% -1.80 [-2.55, -1.05]
2014-Kim	50.4	49.5	19	Cuneus		<del></del>	5.15	% -2.31 [-3.50, -1.11]
RE Model for Subgroup (Q = 4.1	13, df = 5, p	= 0.53; I <sup>2</sup> =	= 0.0%)			•		-1.55 [-1.89, -1.21]
left								
2007-Shibata	21	25	104	occipital			⊢€	2.52 [ 2.01, 3.04]
2011-Smith	1.16	1	42	mid-occipital			<b>⊢</b> ■ 5.	97% 1.23 [ 0.55, 1.90]
2014-Olulade	25.8	26.3	60	lingual gyrus		⊢-	6.15	% -0.87 [-1.40, -0.34]
2010-Li,.2	14.56	14.75	32	fusiform gyrus		<del></del>	5.85	% -1.27 [-2.03, -0.51]
2010-Li,.1	14.56	14.75	32	occipital		<del></del>	5.83	% -1.39 [-2.17, -0.62]
2018-Pereira-Jorge.3	51.29	46.54	25	Middle occipital gyrus		<b>⊢</b>	5.67	% -1.40 [-2.28, -0.51]
2013-Pénicaud.2	39.2	37.3	66	occipital -V3a/V7		<del></del>	6.07	% -1.84 [-2.44, -1.24]
2013-Pénicaud.1	39.2	37.3	66	occipital -V1/V2		<del></del>	6.04	% -2.14 [-2.77, -1.52]
RE Model for Subgroup (Q = 21	5.68, df = 7	, p = 0.00;	$I^2 = 96.0\%$ )					-0.64 [-1.78, 0.51]
bilateral								
2000-Bavelier	23	23	20	V1		<b>⊢</b>	5.6	8% 0.09 [-0.80, 0.97]
2014-Profant	69.14	24.34	54	V1		<b>—</b>	■ 6.1	3% -0.23 [-0.79, 0.32]
2014-Lin	73.8	67	126	occipital lobe		⊢=	→ 6.31	% -0.55 [-0.91, -0.18]
RE Model for Subgroup (Q = 2.1	14, df = 2, p	= 0.34; I <sup>2</sup> =	= 12.3%)					-0.38 [-0.70, -0.06]
RE Model for All Studies	(Q = 256	6.56, df =	= 16, p =	$0.00; I^2 = 92.8\%)$		-	- 100.00	% -0.89 [-1.49, -0.29]
							İ	$\neg$
					-5		0	3
						Hedge's G	j	

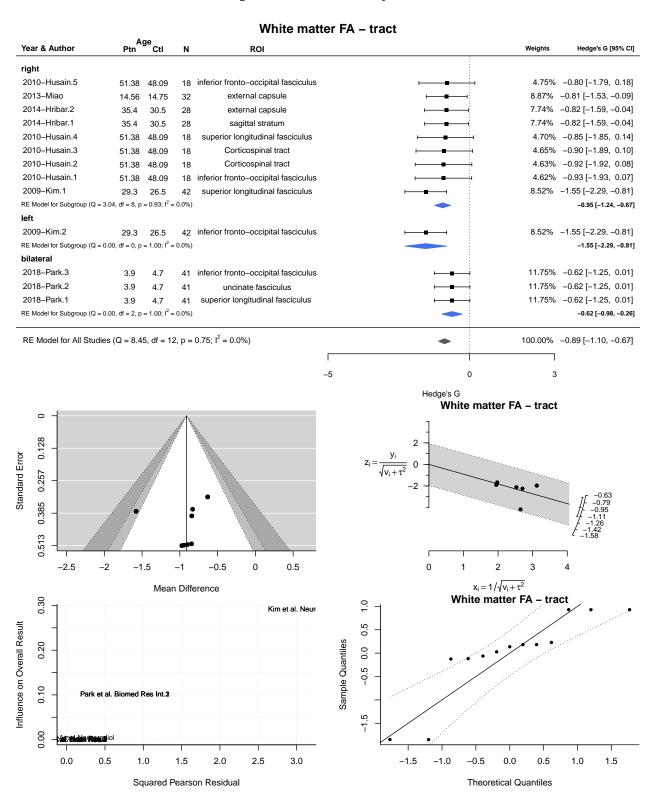


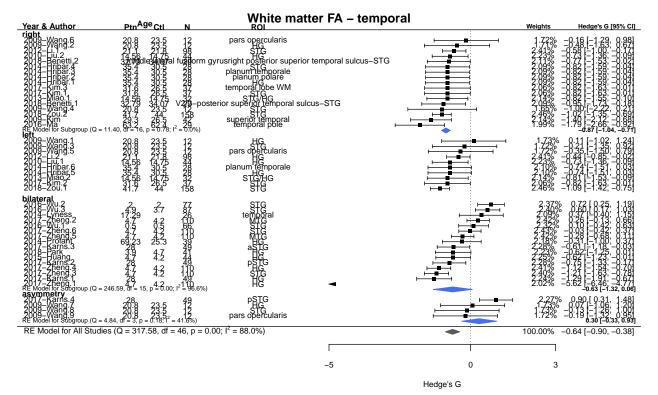
# Gray matter Volume – insular cortex

		ge					
Year & Author	Ptn	Ctl	N	ROI	:	Weights	Hedge's G [95% CI]
right							
2008-Allen.4	23.8	28.5	50	anterior insula	<b>⊢</b>	9.59%	0.31 [-0.25, 0.87]
2008-Allen.2	23.8	28.5	50	insula	<b>⊢</b> •	9.59%	0.22 [-0.33, 0.78]
2008-Allen.6	23.8	28.5	50	posterior insula	<del></del> -	9.60%	0.01 [-0.55, 0.56]
2014-Olulade	25.8	26.3	60	claustrum	<b>⊢</b> ■	9.64%	-1.07 [-1.62, -0.53]
2018-Pereira-Jorge.1	51.29	46.54	25	Insula	<b>⊢</b>	8.29%	-1.48 [-2.38, -0.59]
2018-Pereira-Jorge.3	51.29	46.54	25	Claustrum	<b>⊢</b>	8.24%	-1.57 [-2.48, -0.66]
RE Model for Subgroup (Q = 30.	.86, df = 5,	$p = 0.00; I^2$	= 85.5%)				-0.54 [-1.23, 0.14]
left							
2008-Allen.5	23.8	28.5	50	posterior insula	<b>⊢-</b>	9.54%	0.71 [ 0.14, 1.29]
2008-Allen.1	23.8	28.5	50	insula	<b>⊢</b>	9.56%	0.57 [ 0.01, 1.14]
2008-Allen.3	23.8	28.5	50	anterior insula	ı <del></del>	9.58%	0.37 [-0.18, 0.93]
2018-Pereira-Jorge.2	51.29	46.54	25	Insula	<b>⊢</b>	8.36%	-1.35 [-2.23, -0.47]
2014-Kim	50.9	49.5	22	insula	<b>⊢</b>	8.01%	-1.58 [-2.54, -0.61]
RE Model for Subgroup (Q = 29.	.88, df = 4,	p = 0.00; I <sup>2</sup>	= 90.0%)				-0.20 [-1.15, 0.75]
				2			
RE Model for All Studies	(Q = 67.	44, df =	10, p = 0	.00; I <sup>2</sup> = 87.4%)		100.00%	-0.38 [-0.93, 0.17]
					-		
					<b>-</b> 5 0	3	
					Hedge's G		

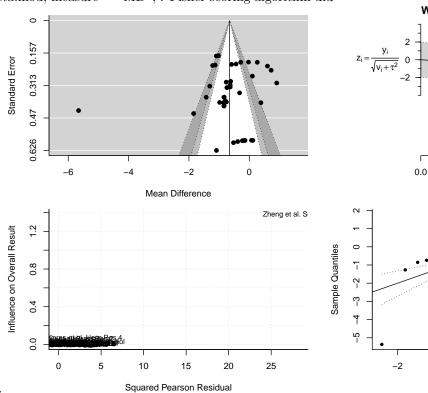


# Meta-regressions of White Matter FA & Brain Areas: Random effects model no intercept covariated by Side



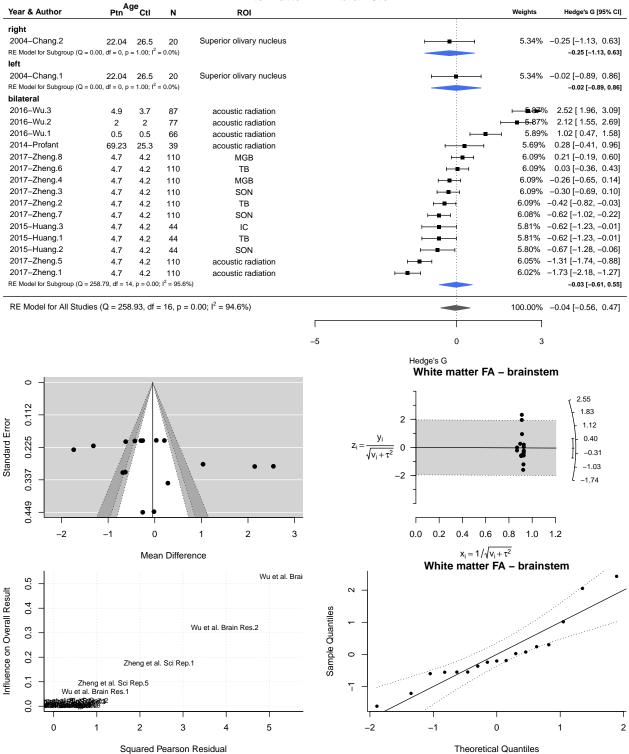


Error in rma(yi = hedgesG, vi = varG, data = meta.mod, measure = "MD", : Fisher scoring algorithm did

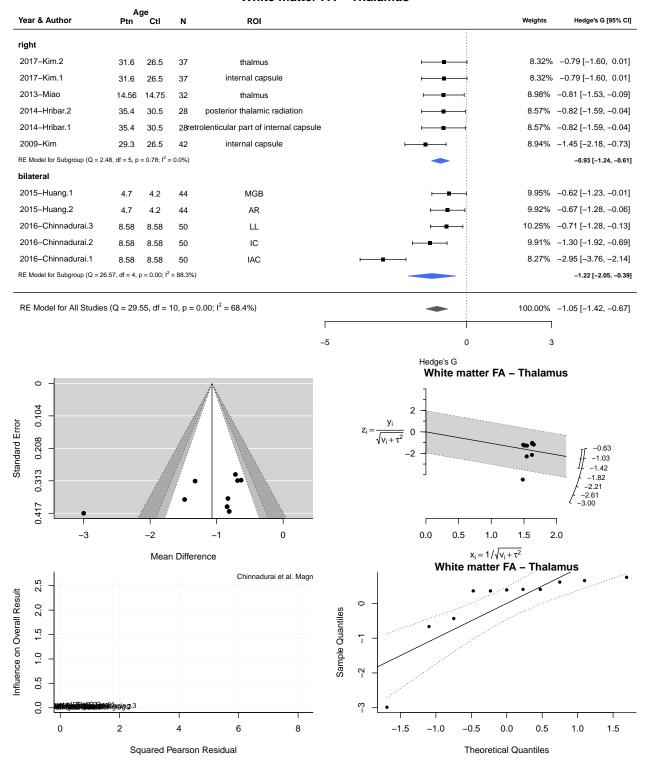


not converge. See 'help(rma)' for possible remedies.

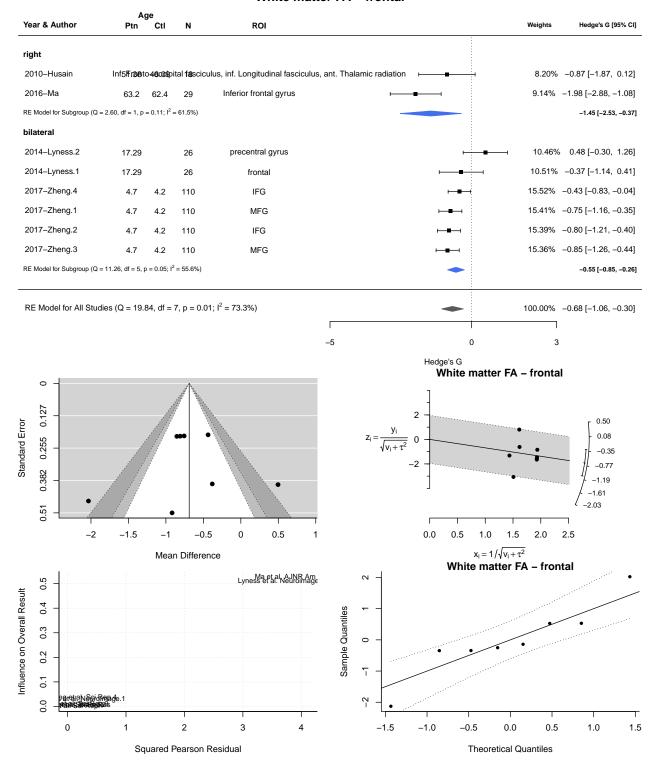
#### White matter FA - brainstem



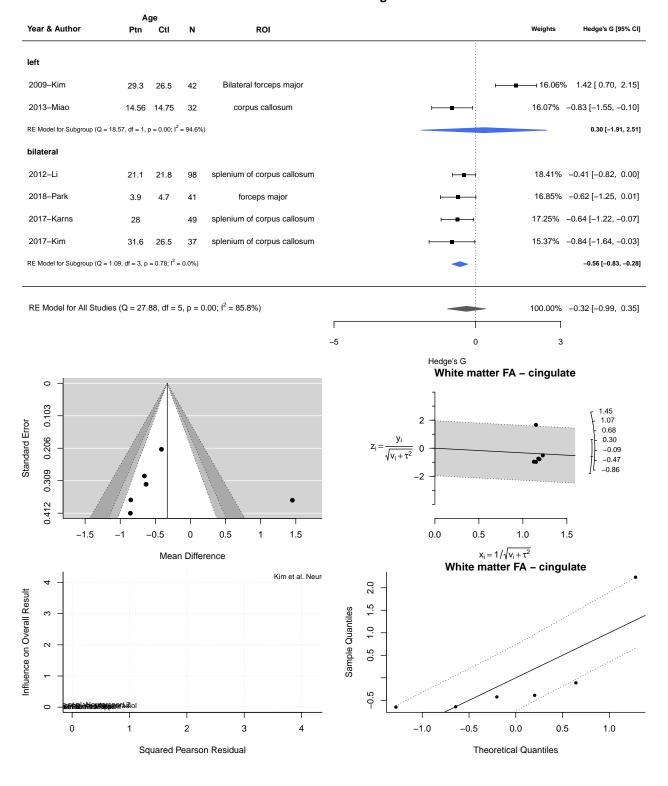
#### White matter FA - Thalamus



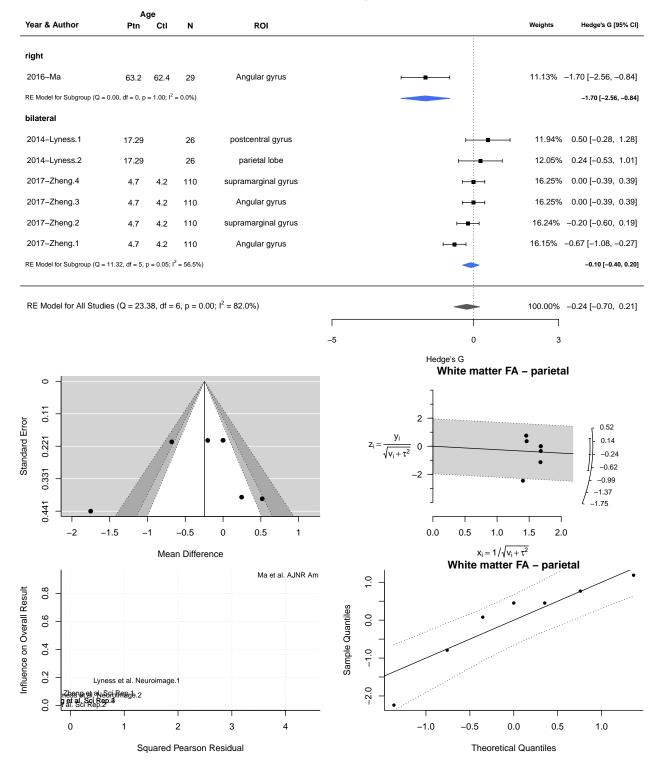
#### White matter FA - frontal



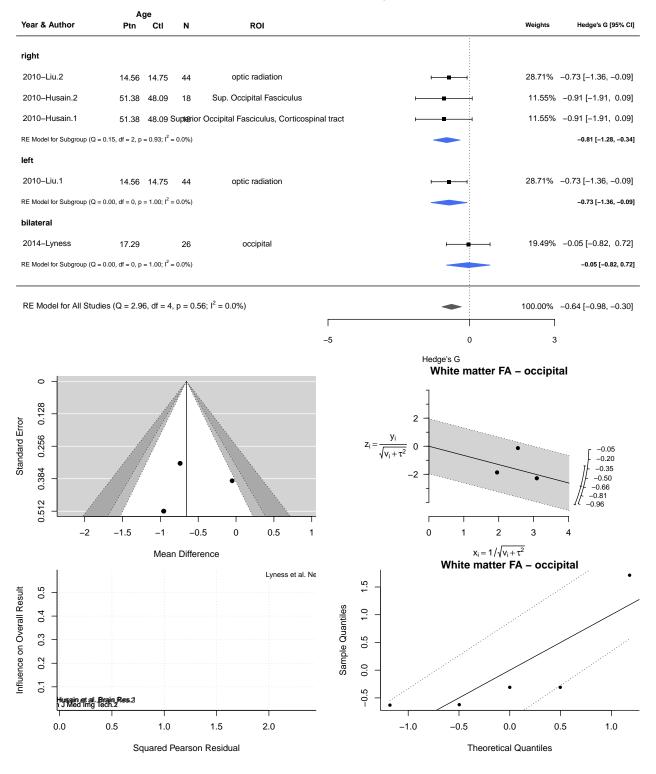
#### White matter FA - cingulate



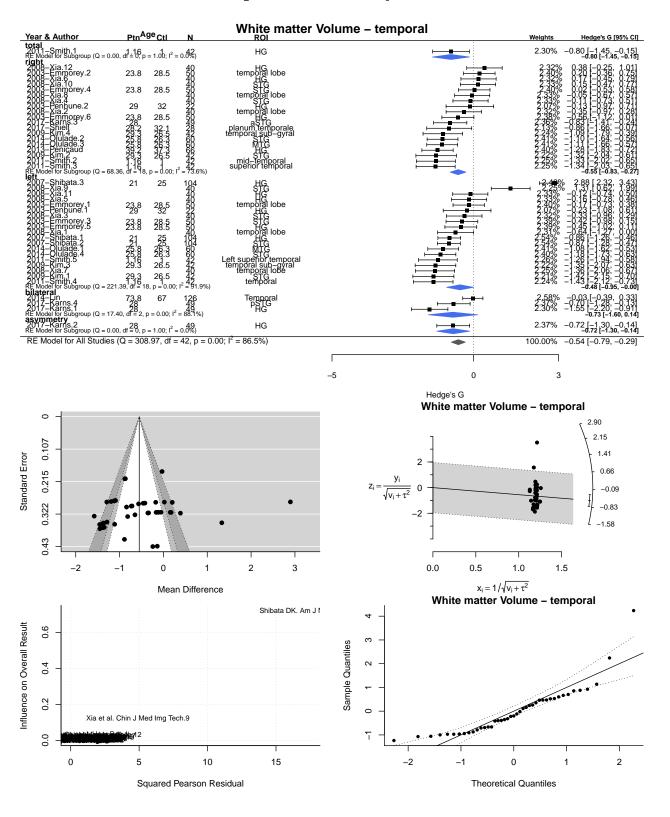
#### White matter FA - parietal



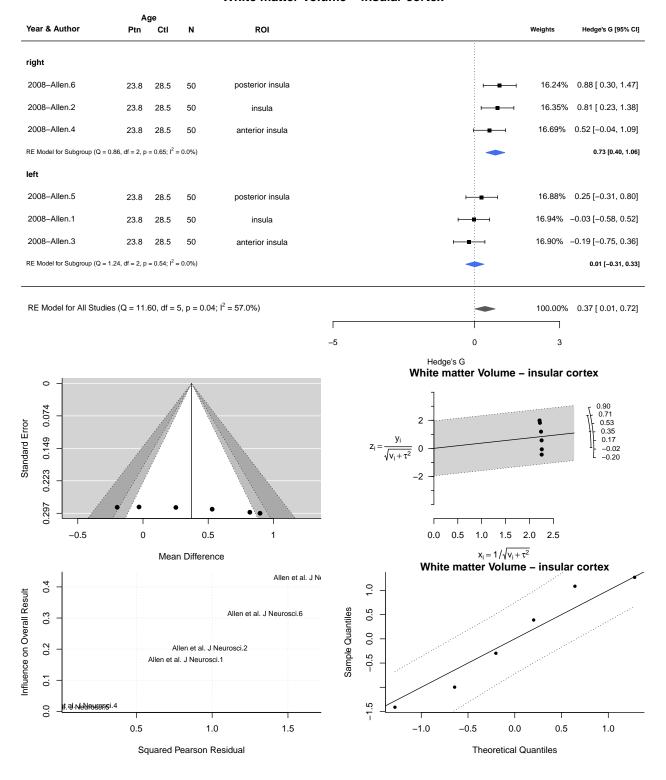
#### White matter FA - occipital



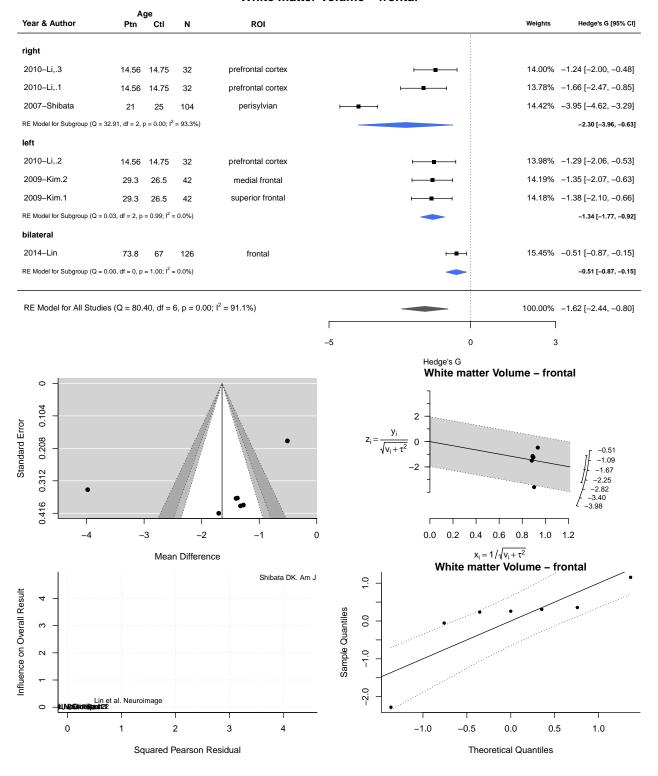
# Meta-regressions of White Matter Volume & Brain Areas: Random effects model no intercept covariated by Side



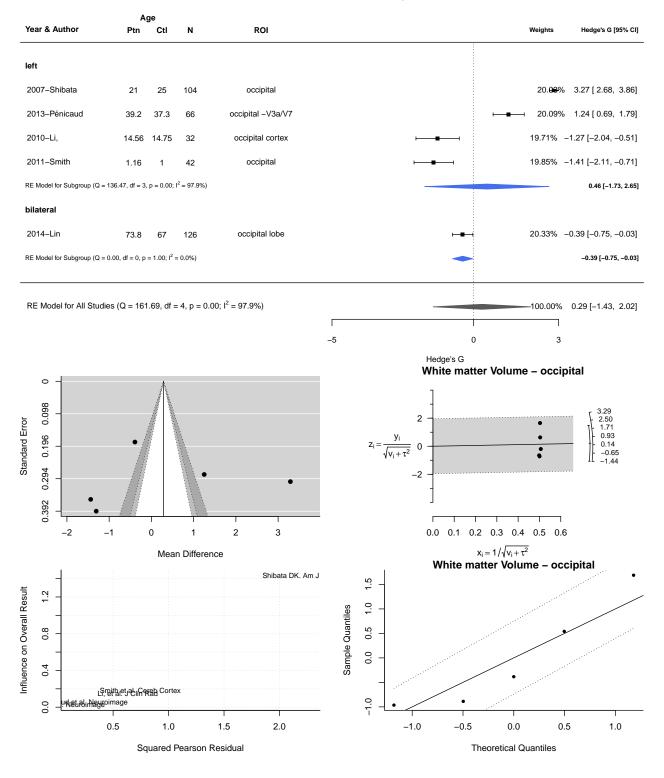
#### White matter Volume - insular cortex



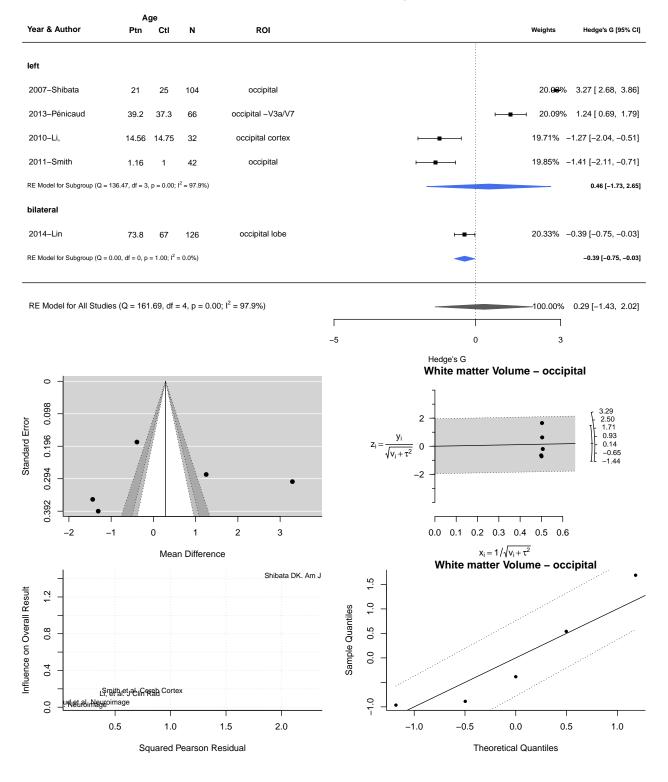
#### White matter Volume - frontal



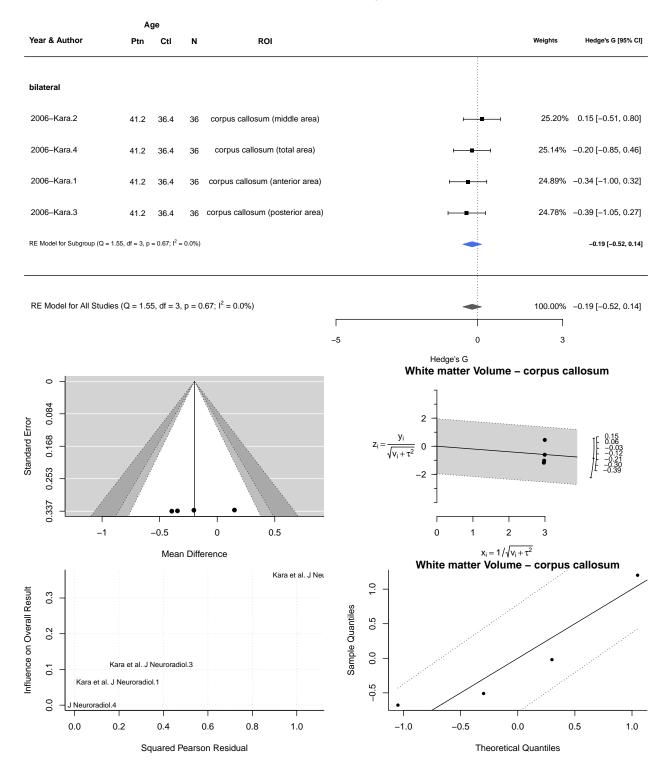
#### White matter Volume - occipital



#### White matter Volume - occipital



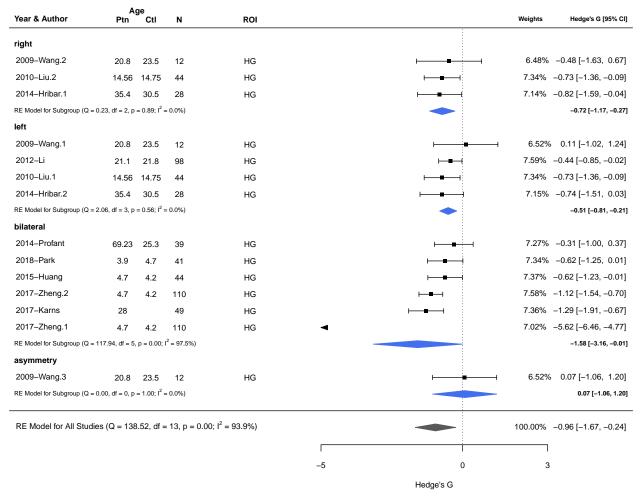
#### White matter Volume - corpus callosum



# Supplementary material: Forest-plots of other Measures

#### Hesch gyrus FA white matter

#### White matter FA and HG



# STG Volume White matter

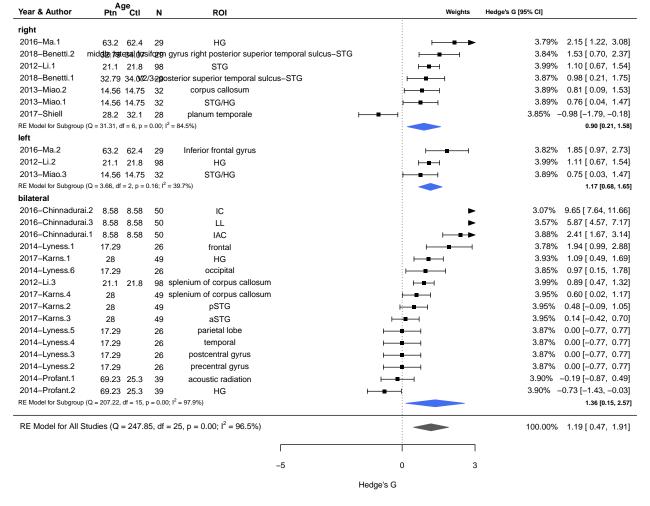
#### White matter FA and STG

Ptn	ge Ctl	N	ROI			Weights	Hedge's G [95% CI]
21.1	21.8	98	STG		<b>⊢</b> ■→	8.31%	-0.58 [-1.00, -0.17]
35.4	30.5	28	STG		<b>⊢</b>	6.55%	-0.82 [-1.59, -0.04]
31.6	26.5	37	STG		<b>⊢</b>	6.38%	-0.82 [-1.63, -0.01]
20.8	23.5	12	STG		<del>  ■</del>	4.55%	-1.00 [-2.22, 0.21]
41.7	44	158	STG		⊢■⊣	8.65%	-1.02 [-1.35, -0.69]
5, df = 4, p	= 0.62; I <sup>2</sup>	= 13.4%)			•		-0.84 [-1.10, -0.58]
20.8	23.5	12	STG		<del>  ■</del>	4.87%	-0.21 [-1.35, 0.92]
31.6	26.5	37	STG		<b>⊢</b>	6.38%	-0.82 [-1.63, -0.01]
41.7	44	158	STG		<del>⊢■</del> ⊣	8.64%	-1.09 [-1.42, -0.75]
9, df = 2, p	= 0.32; I <sup>2</sup>	= 11.6%)			•		-0.95 [-1.31, -0.59]
2	2	77	STG		<b>⊢=</b> →	8.06%	0.72 [ 0.25, 1.19]
4.9	3.7	87	STG		<b>⊢</b> ■→	8.24%	0.60 [ 0.17, 1.03]
0.5	0.5	66	STG		<b>⊢</b> ■	7.80%	0.10 [-0.42, 0.63]
4.7	4.2	110	STG		<u>;</u> ⊢ <del>≡</del>	8.41%	-0.03 [-0.42, 0.37]
4.7	4.2	110	STG		<b>⊢=</b> ⊣	8.27%	-1.21 [-1.63, -0.78]
53, df = 4, p	o = 0.00; I <sup>2</sup>	<sup>2</sup> = 91.3%)					0.03 [-0.64, 0.71]
20.8	23.5	12	STG		<del> </del>	4.88%	-0.13 [-1.26, 1.00]
0, df = 0, p	= 1.00; I <sup>2</sup>	= 0.0%)					-0.13 [-1.26, 1.00]
(Q = 96.	47, df =	13, p = 0.00;	I <sup>2</sup> = 84.5%)		•	100.00%	-0.44 [-0.80, -0.08]
					<u> </u>		
				-5	0	3	
					Hedge's G		
	21.1 35.4 31.6 20.8 41.7 20.8 31.6 41.7 29, df = 2, p 0.5 4.7 4.7 53, df = 4, p	21.1 21.8 35.4 30.5 31.6 26.5 20.8 23.5 41.7 44 35, df = 4, p = 0.62; 1 <sup>2</sup> 20.8 23.5 31.6 26.5 41.7 44 49, df = 2, p = 0.32; 1 <sup>2</sup> 2 2 4.9 3.7 0.5 0.5 4.7 4.2 4.7 4.2 5.5, df = 4, p = 0.00; 1 <sup>2</sup> 20.8 23.5	21.1 21.8 98 35.4 30.5 28 31.6 26.5 37 20.8 23.5 12 41.7 44 158 35, df = 4, p = $0.62$ ; $I^2 = 13.4\%$ ) 20.8 23.5 12 31.6 26.5 37 41.7 44 158 39, df = 2, p = $0.32$ ; $I^2 = 11.6\%$ ) 2 2 77 4.9 3.7 87 0.5 0.5 66 4.7 4.2 110 4.7 4.2 110 53, df = 4, p = $0.00$ ; $I^2 = 91.3\%$ ) 20.8 23.5 12 30, df = 0, p = $1.00$ ; $I^2 = 0.0\%$ )	21.1 21.8 98 STG 35.4 30.5 28 STG 31.6 26.5 37 STG 20.8 23.5 12 STG 41.7 44 158 STG 31.6 26.5 37 STG 20.8 23.5 12 STG 41.7 44 158 STG 31.6 26.5 37 STG 41.7 44 158 STG 41.7 44 158 STG 41.7 44 158 STG 41.7 45 158 STG 45 158 STG 47 45 110 STG 47 45 110 STG 47 45 110 STG 55, df = 4, p = 0.00; $t^2 = 91.3\%$	21.1 21.8 98 STG 35.4 30.5 28 STG 31.6 26.5 37 STG 20.8 23.5 12 STG 41.7 44 158 STG 31.6 26.5 37 STG 20.8 23.5 12 STG 41.7 44 158 STG 31.6 26.5 37 STG 41.7 44 158 STG 41.7 44 158 STG 41.7 44 158 STG 41.7 44 158 STG 41.7 42 158 STG 0.5 0.5 66 STG 4.7 4.2 110 STG 4.7 4.2 110 STG 53, df = 4, p = 0.00; $l^2$ = 91.3%)  20.8 23.5 12 STG 00, df = 0, p = 1.00; $l^2$ = 0.0%)  (Q = 96.47, df = 13, p = 0.00; $l^2$ = 84.5%)	21.1 21.8 98 STG 35.4 30.5 28 STG 31.6 26.5 37 STG 20.8 23.5 12 STG 41.7 44 158 STG  20.8 23.5 12 STG 41.7 44 158 STG  20.8 23.5 12 STG 31.6 26.5 37 STG 41.7 44 158 STG  20.8 23.5 12 STG 41.7 44 158 STG  20.8 23.5 12 STG 41.7 42 158 STG  20.5 0.5 66 STG 4.7 4.2 110 STG	21.1 21.8 98 STG 35.4 30.5 28 STG 31.6 26.5 37 STG 20.8 23.5 12 STG 41.7 44 158 STG 31.6 26.5 37 STG 41.7 44 158 STG 31.6 26.5 37 STG 31.6 26.5 37 STG 31.6 26.5 37 STG 41.7 44 158 STG 31.6 26.5 37 STG 41.7 44 158 STG 41.7 44 158 STG 41.7 42 158 STG 41.7 42 158 STG 41.7 42 158 STG 42.9 = 0.32; l² = 11.6%)  2 2 77 STG 4.9 3.7 87 STG 4.9 3.7 87 STG 4.7 4.2 110 STG 4.8 4.8% 4.88%

# Measures of White matter Integrity

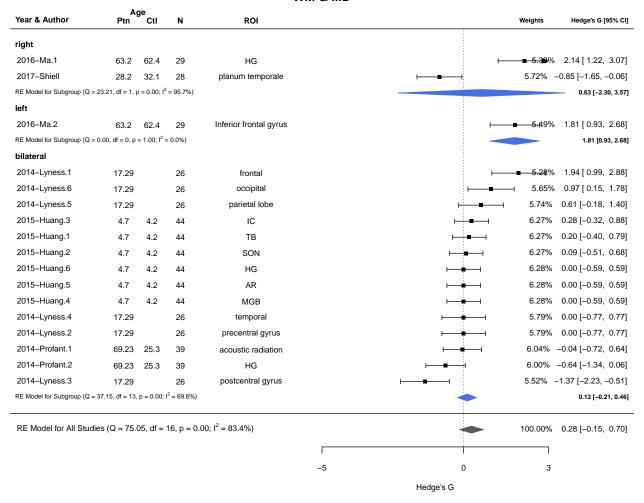
White matter: RD

WM & RD



#### White matter: MD

WM & MD



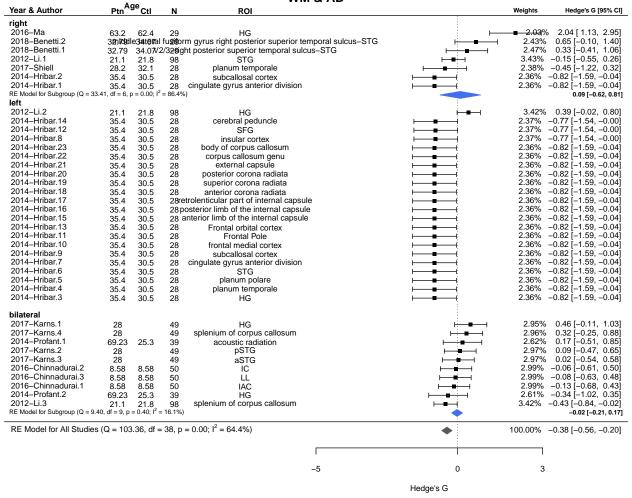
# White matter: Mean Kurtosis

WM & Mean Kurtosis

Year & Author	Ptn	ge Ctl	N	ROI		Weights	Hedge's G [95% CI]
bilateral							
2017-Zheng.9	4.7	4.2	110	STG	<del>⊢■</del> →	3.77%	0.60 [ 0.20, 1.00]
2017-Zheng.19	4.7	4.2	110	IFG	<del>`</del>	3.79%	0.38 [-0.01, 0.78]
2017-Zheng.17	4.7	4.2	110	HG	<del></del>	3.79%	0.38 [-0.02, 0.78]
2017-Zheng.18	4.7	4.2	110	MFG	<del>:</del>	3.80%	0.22 [-0.17, 0.62]
2017-Zheng.13	4.7	4.2	110	acoustic radiation	<del>  ■  </del>	3.81%	0.13 [-0.26, 0.52]
2017-Zheng.15	4.7	4.2	110	SON	<b>⊢</b>	3.81%	0.05 [-0.35, 0.44]
2017-Zheng.24	4.7	4.2	110	Hippocampus	<del>⊢                                    </del>	3.81%	0.00 [-0.39, 0.39]
2017-Zheng.23	4.7	4.2	110	supramarginal gyrus	⊢ <del>∔</del> ·	3.81%	0.00 [-0.39, 0.39]
2017-Zheng.22	4.7	4.2	110	Angular gyrus	⊢ <del>∔</del> ⊣	3.81%	0.00 [-0.39, 0.39]
2017-Zheng.21	4.7	4.2	110	STG	<del>                                      </del>	3.81%	0.00 [-0.39, 0.39]
2017-Zheng.16	4.7	4.2	110	MGB	<b>⊢=</b> -1	3.81%	-0.11 [-0.50, 0.28]
2017-Zheng.20	4.7	4.2	110	MTG	<del>⊢=</del> ∺	3.80%	-0.20 [-0.59, 0.20]
2017-Zheng.12	4.7	4.2	110	Hippocampus	<del>- ■  </del>	3.80%	-0.32 [-0.71, 0.08]
2017–Zheng.14	4.7	4.2	110	TB	<b>⊢=</b> -i	3.79%	-0.46 [-0.86, -0.07]
2017-Zheng.4	4.7	4.2	110	MGB	<del>- ■</del> -i	3.79%	-0.47 [-0.86, -0.07]
2016-Chinnadurai.1	8.58	8.58	50	IAC	<b>⊢</b> ■	3.10%	-0.48 [-1.04, 0.08]
2017-Zheng.7	4.7	4.2	110	IFG	<b>⊢=</b> →	3.78%	-0.50 [-0.90, -0.10]
2016-Chinnadurai.2	8.58	8.58	50	IC	<del> i</del>	3.09%	-0.53 [-1.09, 0.03]
2017-Zheng.8	4.7	4.2	110	MTG	⊢ <del>≡</del> ⊣	3.78%	-0.56 [-0.96, -0.16]
2017-Zheng.5	4.7	4.2	110	HG	⊢ <del>=</del> →	3.78%	-0.58 [-0.98, -0.17]
2017-Zheng.1	4.7	4.2	110	acoustic radiation	<b>⊢=</b> →	3.78%	-0.58 [-0.98, -0.18]
2017-Zheng.3	4.7	4.2	110	SON	⊢ <del>≡</del> ⊣	3.76%	-0.70 [-1.10, -0.29]
2017-Zheng.11	4.7	4.2	110	supramarginal gyrus	<del>⊢∎ </del>	3.76%	-0.71 [-1.11, -0.30]
2017-Zheng.6	4.7	4.2	110	MFG	<b>⊢=</b> →	3.76%	-0.73 [-1.13, -0.32]
2017-Zheng.2	4.7	4.2	110	ТВ	<b>⊢=</b> →	3.74%	-0.84 [-1.25, -0.44]
2016-Chinnadurai.3	8.58	8.58	50	LL	<b>⊢=</b>	3.01%	-0.92 [-1.51, -0.34]
2017-Zheng.10	4.7	4.2	110	Angular gyrus	⊢ <del>≡</del> ⊣	3.70%	-1.07 [-1.49, -0.65]
RE Model for Subgroup (Q = 11	0.59, df = 26	6, p = 0.00	$I^2 = 76.6\%$	)	•		-0.29 [-0.45, -0.12
RE Model for All Studies	s (Q = 110	).59, df =	= 26, p = 0	$0.00; I^2 = 76.6\%)$	•	100.00%	-0.29 [-0.45, -0.12]
					; 		
					-5 0	3	
					Hedge's G		

#### White matter: AD

WM & AD

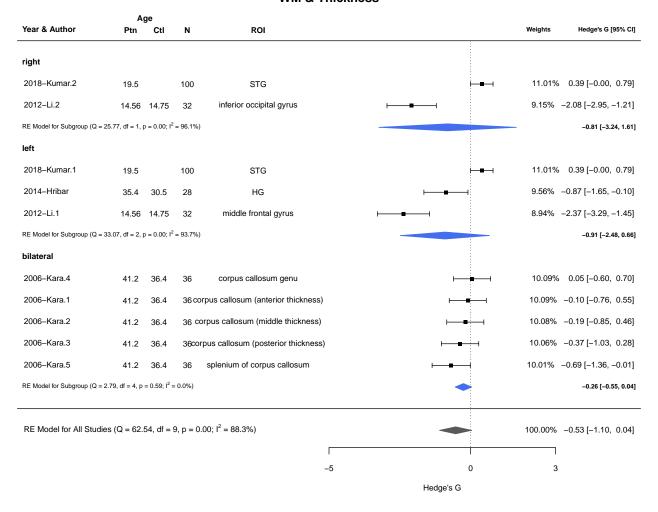


Error in rma(yi = hedgesG, vi = varG, data = meta.mod, measure = "MD", : Fisher scoring algorithm did not converge. See 'help(rma)' for possible remedies.

## Other Measures of White Matter

#### White matter: Thickness

**WM & Thickness** 



#### White matter: VBM

WM & VBM

Year & Author	Ag Ptn	je Ctl	N	ROI				Weights	Hedge's G [95% CI]
right									
2010-Leporé.7	29.5	8	30	MTG		į.		6.12%	0.73 [-0.01, 1.47]
2010-Leporé.4	29.5	8	30	STG		i.		6.12%	0.73 [-0.01, 1.47]
2010-Leporé.1	29.5	8	30	STG		į.		6.12%	0.73 [-0.01, 1.47]
2018-Kumar.2	19.5		100	STG		⊢■→		7.03% -	-1.47 [-1.92, -1.03]
RE Model for Subgroup (Q = 4	19.03, df = 3, p	= 0.00; I	<sup>2</sup> = 91.5%)	)					0.15 [-0.97, 1.27]
left									
2010-Leporé.6	29.5	8	30	MTG		<u> </u>	-	6.12%	0.73 [-0.01, 1.47]
2010-Leporé.5	29.5	8	30	Intraparietal sulcus		į.	-	6.12%	0.73 [-0.01, 1.47]
2010-Leporé.3	29.5	8	30	STG		<u> </u>	-	6.12%	0.73 [-0.01, 1.47]
2010-Leporé.2	29.5	8	30	STG		į.		6.12%	0.73 [-0.01, 1.47]
2018-Kumar.1	19.5		100	STG		<del>⊢ ■                                   </del>		7.00% -	-1.64 [–2.10, –1.19]
RE Model for Subgroup (Q = 6	62.80, df = 4, p	= 0.00; I	<sup>2</sup> = 90.6%)	)					0.23 [-0.75, 1.20]
bilateral									
2010-Leporé.13	29.5	8	30	splenium of corpus callosum		<u> </u>	-	6.12%	0.72 [-0.02, 1.46]
2010-Leporé.12	29.5	8	30	temporal lobe		ļ <del>.</del>		6.13%	0.68 [-0.06, 1.42]
2010-Leporé.8	29.5	8	30	frontal lobe		ı <del>!</del>		6.14%	0.66 [-0.07, 1.40]
2010-Leporé.11	29.5	8	30	parietal lobe		1		6.16%	0.56 [-0.18, 1.29]
2010-Leporé.14	29.5	8	30	corpus callosum genu		ı <u>i</u>		6.18%	0.40 [-0.32, 1.13]
2010-Leporé.10	29.5	8	30	occipital lobe		<b>⊢</b>		6.19%	0.26 [-0.46, 0.98]
2010-Leporé.9	29.5	8	30	limbic lobe		<b>├</b>	<b>→</b>	6.20%	0.01 [-0.71, 0.73]
RE Model for Subgroup (Q = 3	3.03, df = 6, p =	= 0.81; I <sup>2</sup>	= 0.0%)			•	•		0.46 [0.19, 0.74]
RE Model for All Studie	es (Q = 132.	.56, df	= 15, p =	= 0.00; I <sup>2</sup> = 82.9%)			<b>&gt;</b>	100.00%	0.30 [-0.11, 0.71]
						i			
					-5	0		3	
						Hedge's G			

#### Meta Plots

#### The L'Abbé plot

In a L'Abbé plot (based on L'Abbé, Detsky, & O'Rourke, 1987), the arm-level outcomes for two experimental groups (e.g., treatment and control group) are plotted against each other. is treatment versus effect, since you have the cohen's d this should be relatively simple.

> WE DON'T HAVE TWO EXPERIMENTAL GROUPS

## Baujat plot to identify studies contributing to heterogeneity

The plot shows the contribution of each study to the overall Q-test statistic for heterogeneity on the horizontal axis versus the influence of each study (defined as the standardized squared difference between the overall estimate based on a fixed-effects model with and without the ith study included in the model) on the vertical axis 2.17. Funnel plot to illustrate publication bias

#### Galbraith plot

Radial plot (radial) of variables and cohen's d - Galbraith, Rex (1988). "Graphical display of estimates having differing standard errors". Technometrics. Technometrics, Vol. 30, No. 3, 30 (3): 271–281.

2.18.2. We want to see this type of error plot over time for our patient cohorts by age. we want this for each measure WM and GM versus age on the x-axis so we can see GM and WM over time! Do a monte carlo simulation to connect different age population and create the error.

For a fixed-effects model, the plot shows the inverse of the standard errors on the horizontal axis against the individual observed effect sizes or outcomes standardized by their corresponding standard errors on the vertical axis. On the right hand side of the plot, an arc is drawn corresponding to the individual observed effect sizes or outcomes. A line projected from (0,0) through a particular point within the plot onto this arc indicates the value of the individual observed effect size or outcome for that point.

#### Resources

We are following Preferred Reporting Items for Systematic Reviews and Meta-Analyses guidelines: Moher, D., Liberati, A., Tetzlaff, J., Altman, D. G., and Prisma Group. (2009). Preferred reporting items for systematic reviews and meta-analyses: the PRISMA statement. PLoS Med. 6:e1000097. doi: 10.1371/journal.pmed.1000097 AND https://www.bmj.com/content/339/bmj.b2535

- https://stackoverflow.com/questions/14426637/how-to-do-bubble-plot
- https://www.researchgate.net/publication/296680807\_Menstrual\_hygiene\_management\_among\_adolescent girls in India A Systematic review and meta-analysis/figures?lo=1

#### Good explanation of some of the plots:

• https://ora.ox.ac.uk/objects/uuid:ff78831d-6f82-4187-97cc-349058e9abde/download\_file?file\_format=pdf&safe\_filename=Rahimi%2Bet%2Bal%252C%2BData%2Bvisualisation%2Bfor%2Bmeta-analysis.pdf&type\_of\_work=Journal+article