# 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.

Study design was using cross-sectional including control groups, ivestigating the structural relation between MRI changes and the Hearing loss. ROI were selected among the included studies Analysis method, the most common outcome measures were **volume**, **FA**, **VBM** and **thickness**. A total of 41 studies were included, with a total of 851 patients and 964 patients.

Notes:

- 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 mutipliying by -1 if the effect was decrease and by 1 if it was none of increased. Forests plots were calculated the meta-regression for left and right. We measure a global meta regression for white and gray mater by side (Left and right). We include those ROI with most frequency of apearing.

Effects were summarized across studies using the generic inverse-variance weighting method with DerSimonian and Laird random effects, meaning studies were weighted by 1/SEš (where SE is the standard error)

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

#### **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 in the between studies variance and eterogeneity calculations. We should have done the mean and standart deviation from each study. Variance was estimated using the following formula:

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

#### Estimate of heterogeneity per model

"We estimated heterogeneity in results using the  $\tau$  statistic, which represents the standard deviation" such as overall.hetstat or a heterogeneity test x2 and I2 index

We performed a multi-level meta-analytic model, over our multiple effect size estimates nested withing variables: Big brain area, ROI, etc. We expect that the underlying true effects are more similar for the same level of the grouping variable 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 N studies, each comparing different measurements between patients and controls. The difference of between groups was quantified in terms of Cohen's D.

#### **Notes from Francis**

Remember I created the effect size because are the variables and metric were different. This take mean and SD out of the equation. But you can represent the mean and SD as another variable if you want to calculate what effect sizes would create a cohen's d and cc r for specific variables. That way we would standardize it across our different means and SD. 2.1.16. CC which is the correlation coefficient

2.3. Calculate Hedges' g which takes into consideration sample size

> I used Cohen's D instead of these last two because ai calculated the variace from the cohen's D, but I can do it for the Hedges'G as well

#### Table of included studies

Table 1: 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 Tech	1.5	VBM	volume
12	2010, Husain et al. Brain Res	3	DTI, VBM	FA, volume
26	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

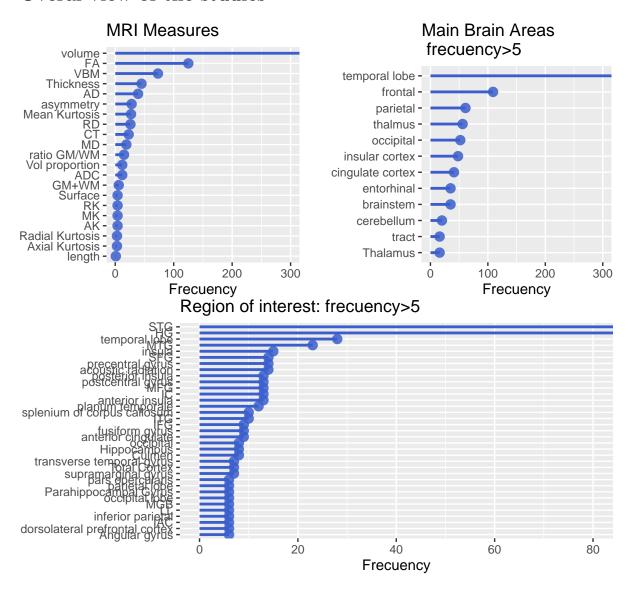
Table 2: Included studies

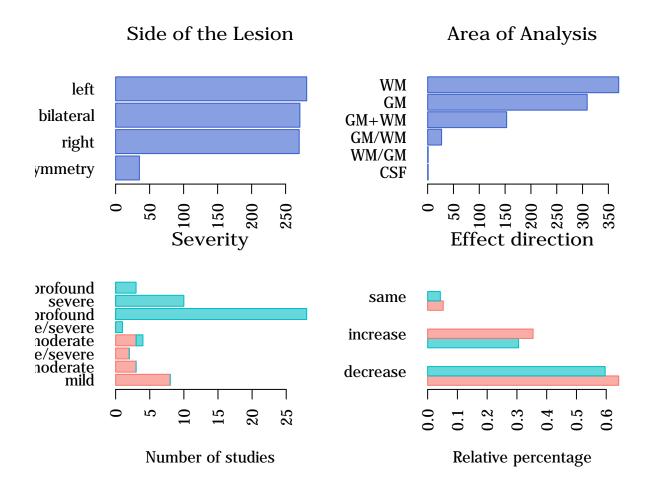
	Source	MRI Tesla	all.techniques	all.measures
1	2000, Bavelier et al. J Neurosci	1.5	VBM	volume
2	2003, Emmorey et al. PNAS	1.5	VBM	asymmetry, GM+WM, ratio GM/WM, volume
3	2003, Penhune et al. Neuroimage	1.5	VBM	asymmetry, ratio GM/WM, volume
4	2004, Chang et al. Neuroreport	3	DTI	asymmetry, FA
6	2007, Meyer et al. Restor Neurol Neurosci	3	VBM	volume

	Source	MRI Tesla	all.techniques	all.measures
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, Vol proportion, volume
10	2009, Kim et al. Neuroreport	9	DTI VDM	1 1
10	,	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, Anen et al. Front Neuroanat 2013, Boyen et al. Hear Res	3	VBM	volume
21	2013, Miao et al. Am J Neuroradiol	3	DTI	FA, RD
22	2013, Pénicaud et al. Neuroimage	1.5	VBM	volume
23	2014, Hribar et al. Hear Res	3	DTI, VBM	AD, FA, Thickness
${\bf 24}$	2014, Kim et al. Hear Res	3	VBM	volume
25	2014, Lin et al. Neuroimage	1.5	VBM	volume
<b>27</b>	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	$\operatorname{CT}$	Thickness
37	2016, Smittenaar et al. Open Neuroimag J	1.5	$\operatorname{CT}$	CT
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
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
40 47	2018, Feng et al. PNAS	3	VBM	VBM
49	2018, Park et al. Biomed Res Int	3	DTI	FA
50	2018, Pereira-Jorge et al. Neural Plast	1.5	VBM	volume
51	2018, Ren et al. Front Neurosci	3	CT, VBM	Thickness, volume
52	2018, Uchida et al. Front Aging Neurosci	3	VBM	volume
53	2018, Uchida et al. Front Aging Neurosci	3	VBM	volume
E 1		9	DTI	AK FA MK DK
$\frac{54}{55}$	2018, Zou et al. Otol Neurotol 2019, Belkhiria et al. Front.	3 3	VBM	AK, FA, MK, RK CT, volume
56	Aging Neurosci 2019, Belkhiria et al. Front.	3	VBM	CT, volume
57	Aging Neurosci 2019, Luan et al. Front Neurosci	3	DTI, VBM	FA, MD, volume

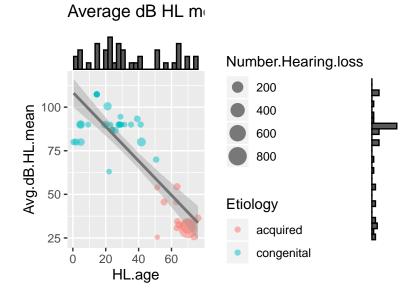
	Source	MRI Tesla	all.techniques	all.measures
58	2019, Ponticorvo et al. Hum Brain Mapp	3	VBM	volume
59	2019, Xu et al. J Magn Reson Imaging	3	VBM	volume

## Overal view of the studies





## Relation of Average dB and Age



#### Frequency and contingensy tables

- 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 their effect varies widely and 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

# Frequency tables: measures of WM & GM

Table 3: Matter vs measure (continued below)

	AD	ADC	AK	asymmetry	Axial Kurtosis	CT	FA	GM+WM
GM WM	0 39	0 12	2	9	0	23	8 117	0

Table 4: Table continues below

	length	MD	Mean Kurtosis	MK	Radial Kurtosis	ratio $GM/WM$	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
GM	2	4	14	43	6	194
$\mathbf{W}\mathbf{M}$	2	0	10	16	6	79

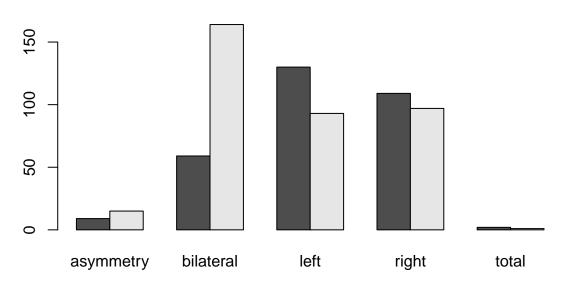
Table 6: Matter vs Side

	asymmetry	bilateral	left	right	total
GM	9	59	130	109	2
WM	15	164	93	97	1

## Contingensy tables: Main areas and sub-areas

Sub-analys were carried out in those areas that had measures from at least five studies. Those areas were selected from the contingensy tables.

#### **Matter vs Side**



#### Most common ROI for WM-FA and GM-volume

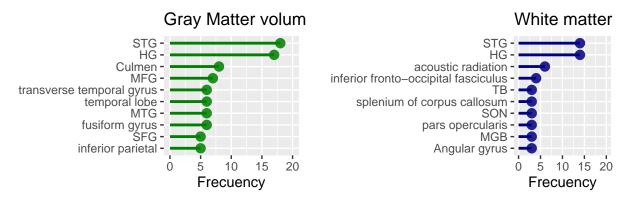
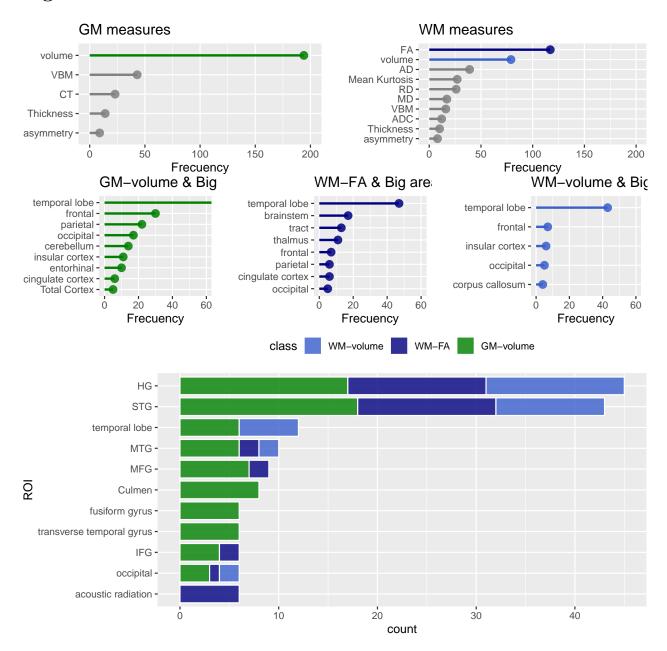
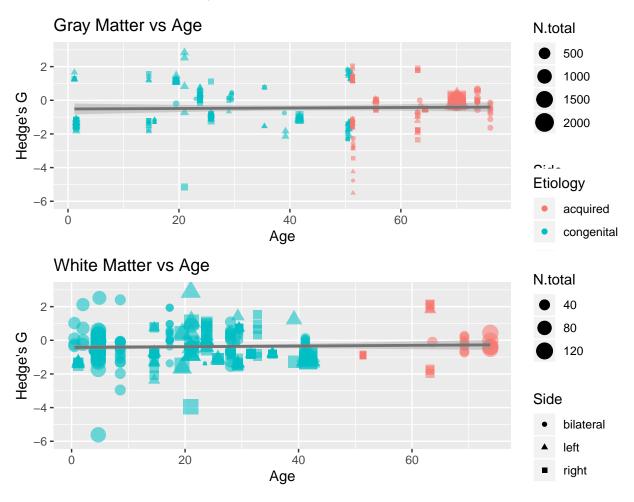


Figure 2 - Studies characteristics

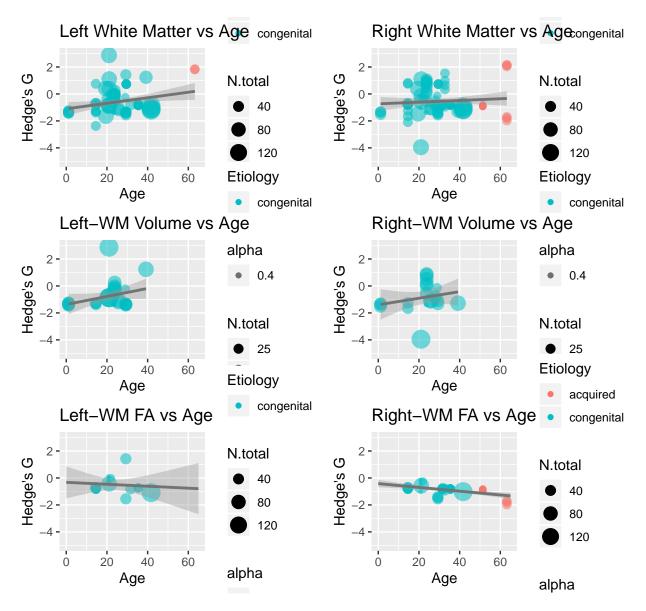


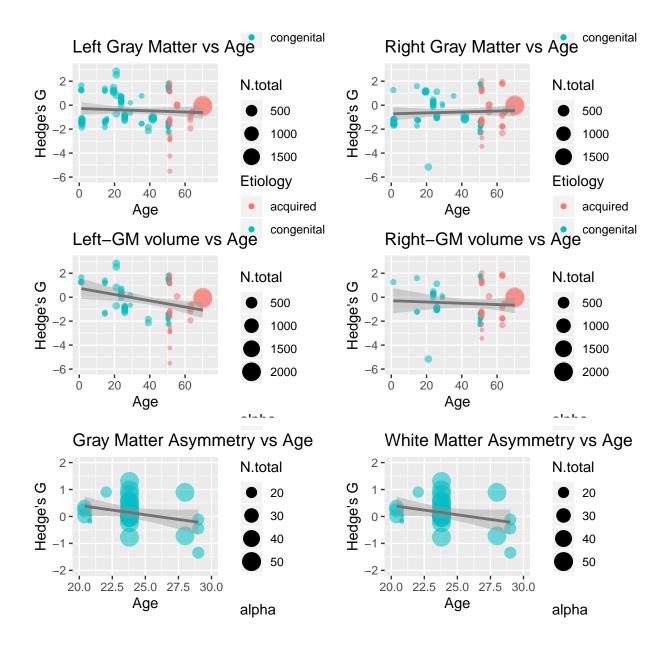
# Gray and White matter relation with Age (All MRI measurements)

All measurements of Gray and White matter



## Relation of gray and white matter to age by volume and FA





# CONGENITAL - Meta-regressions of Gray Matter by Volume

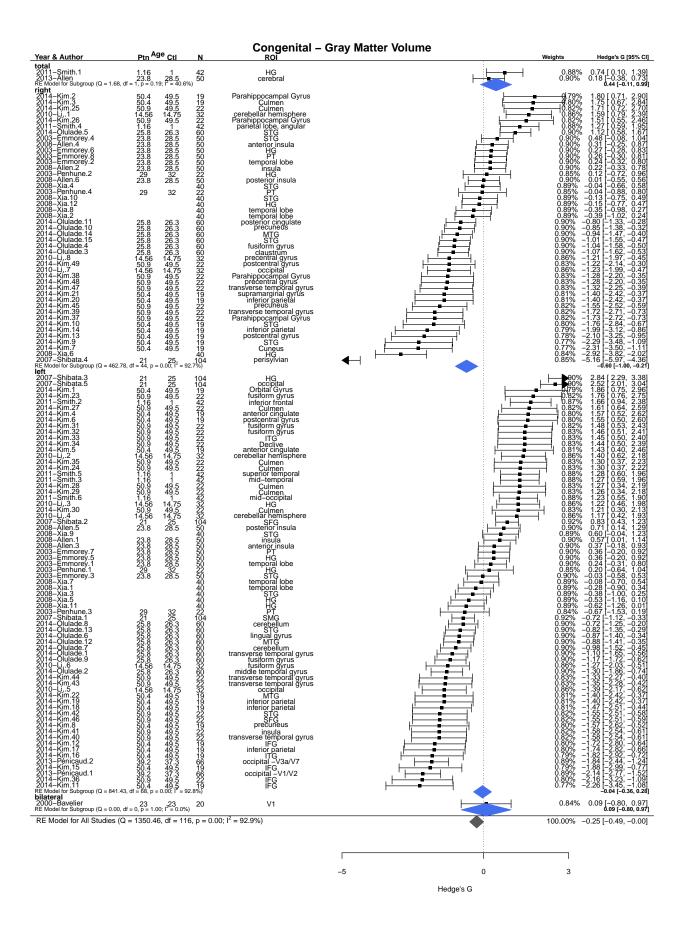
Gray Matter Volume: Random effects model no intercept covariated by Big area

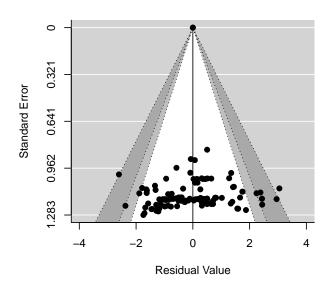
Table 7: Table continues below

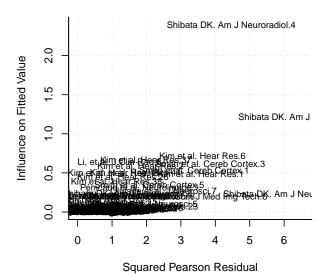
	${\it HedgeG}$	se	zval	ci.lo	ci.up
left cerebellum	0.9013	0.3735	2.413	0.1693	1.633
left cingulate	1.5	0.9037	1.66	-0.2712	3.271
left frontal	-0.588	0.4468	-1.316	-1.464	0.2877
left insular cortex	0.0628	0.6065	0.1035	-1.126	1.252
left occipital	-0.5252	0.4567	-1.15	-1.42	0.3699
left parietal	-0.8875	0.5149	-1.724	-1.897	0.1217
left temporal lobe	-0.116	0.2235	-0.5189	-0.554	0.3221

	HedgeG	se	zval	ci.lo	ci.up
left thalmus	1.282	1.213	1.056	-1.097	3.66
right cerebellum	1.682	0.7284	2.309	0.254	3.109
right cingulate cortex	-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 lobe	-0.5427	0.2729	-1.989	-1.078	-0.007815

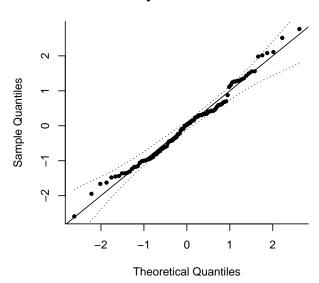
	pval
left cerebellum	0.01581
left cingulate	0.09694
left frontal	0.1882
left insular cortex	0.9175
left occipital	0.2502
left parietal	0.08478
left temporal lobe	0.6039
left thalmus	0.2909
right cerebellum	0.02096
right cingulate cortex	0.5015
right entorhinal	0.9263
right frontal	0.0003399
right insular cortex	0.8228
right occipital	0.05341
right parietal	0.01233
right temporal lobe	0.04675







#### **Gray Matter Volume**



## ACQUIRED - Meta-regressions of Gray Matter by Volume

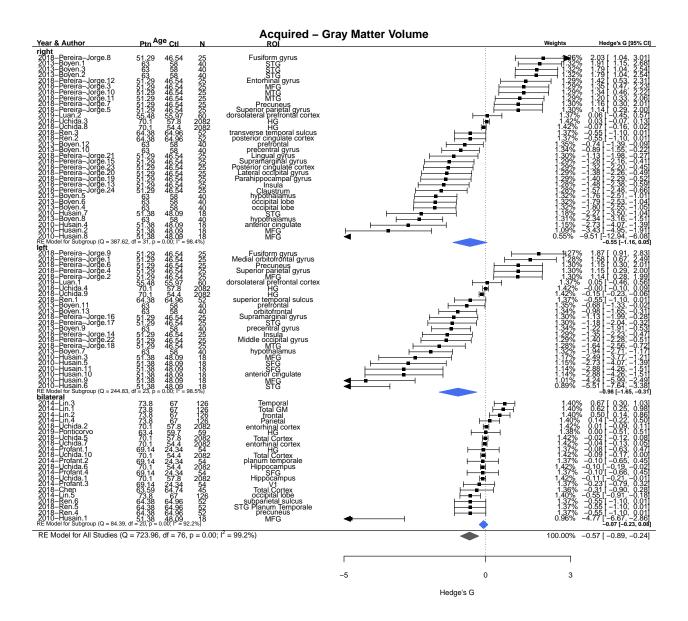
Random effects model no intercept covariated by Big area

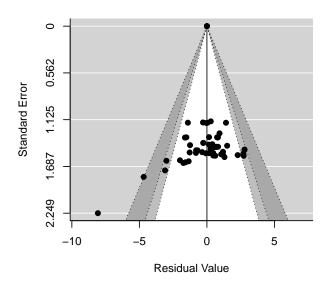
Table 9: Table continues below

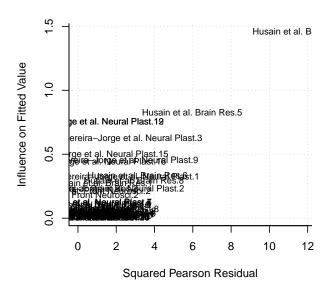
	HedgeG	se	zval	ci.lo	ci.up
left cingulate cortex	-2.883	1.728	-1.669	-6.269	0.5024
left frontal	-1.14	0.5268	-2.164	-2.173	-0.1076
left hypothalamus	-1.937	1.626	-1.191	-5.124	1.25
left insular cortex	-1.353	1.641	-0.8248	-4.57	1.863
left occipital	-1.398	1.642	-0.8514	-4.616	1.82
left parietal	0.3896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

	${\it HedgeG}$	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 lobe	0.727	0.5141	1.414	-0.2806	1.735

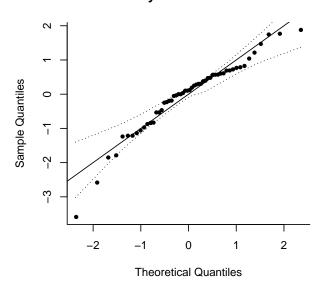
	pval
left cingulate cortex	0.09509
left frontal	0.03045
left hypothalamus	0.2335
left insular cortex	0.4095
left occipital	0.3945
left parietal	0.6803
left temporal lobe	0.1831
right cingulate cortex	0.1202
right entorhinal	0.9951
right frontal	0.04034
right hypothalamus	0.0754
right insular cortex	0.1898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







## **Gray Matter Volume**



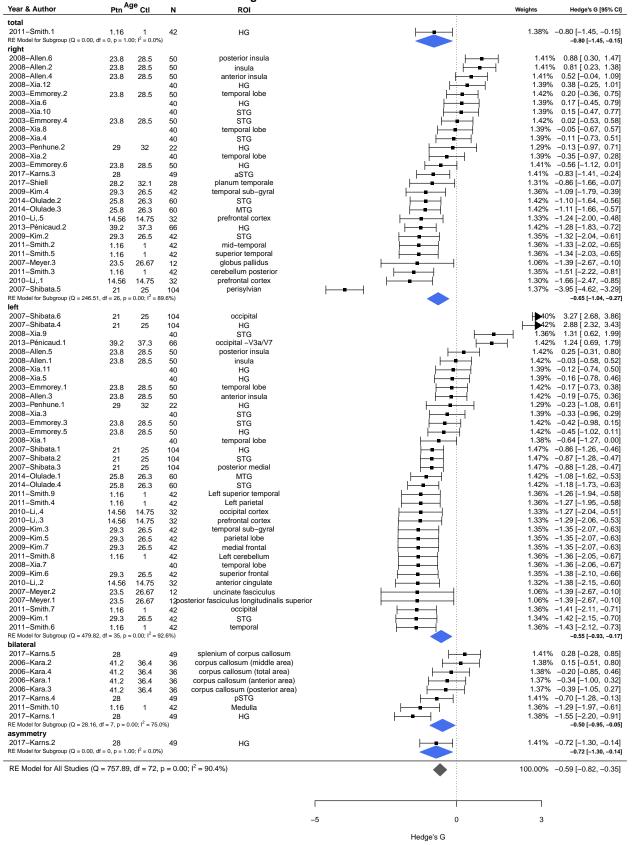
## CONGENITAL - White Matter by VOLUME

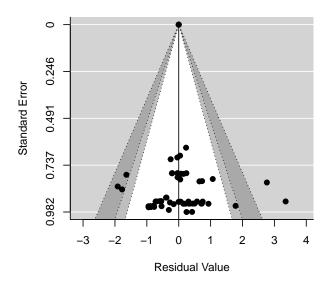
Random effects model no intercept covariated by Big area

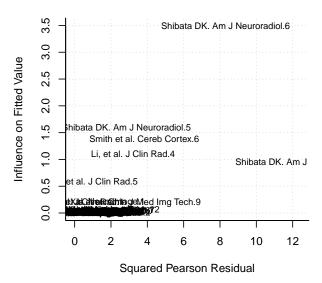
	HedgeG	se	zval	ci.lo	ci.up	N
left cerebellum	-1.107	0.6745	-1.641	-2.429	0.2149	2
left cingulate cortex	-1.379	0.9926	-1.389	-3.324	0.5668	1
left frontal	-1.34	0.5684	-2.358	-2.454	-0.2262	3
left insular cortex	0.007913	0.5504	0.01438	-1.071	1.087	3
left occipital	0.5024	0.4846	1.037	-0.4475	1.452	4
left parietal	-1.308	0.6914	-1.892	-2.663	0.04705	2
left temporal lobe	-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 lobe	-0.5529	0.2218	-2.493	-0.9875	-0.1183	19

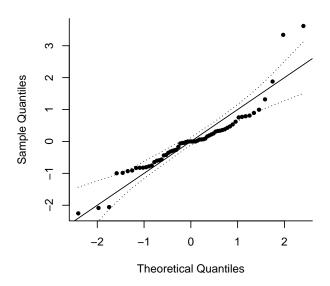
#### **Congenital White Matter Volume**







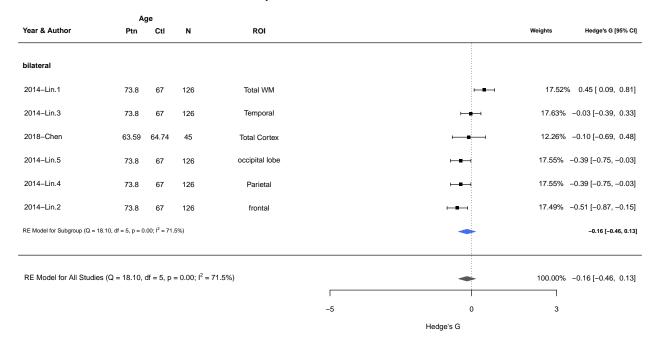
#### **Congenital White Matter Volume**

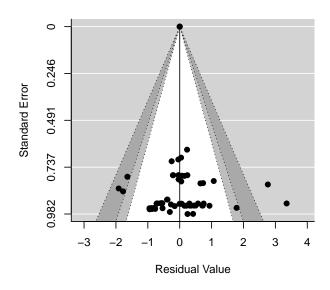


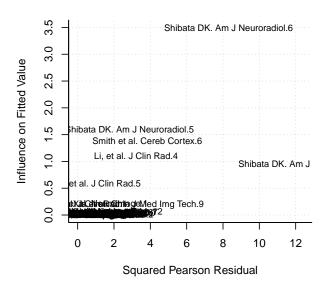
ACQUIRED - White Matter by VOLUME

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

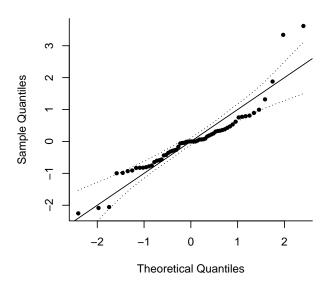
#### acquired White Matter Volume







#### acquired White Matter Volume

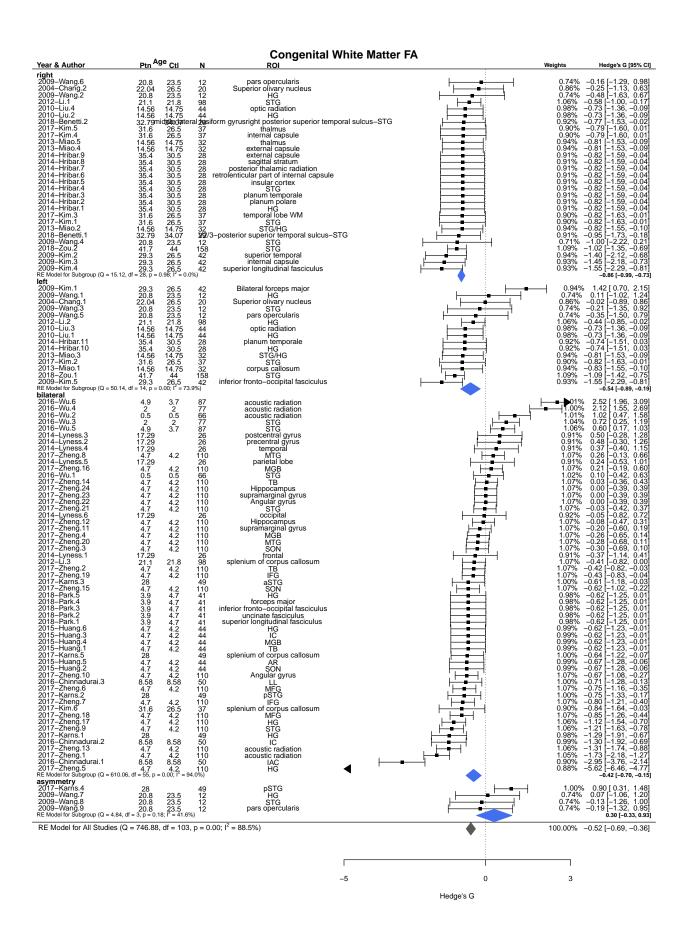


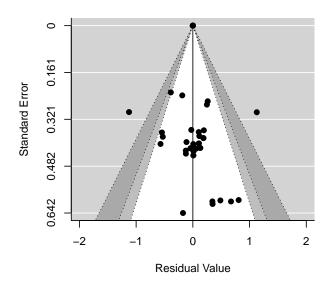
## CONGENITAL - White Matter by FA fractional anisotropy

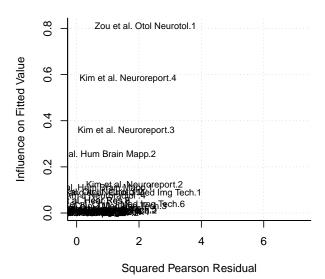
Random effects model no intercept covariated by Big area

	HedgeG	se	zval	ci.lo	ci.up	N
left brainstem	-0.01557	0.4891	-0.03183	-0.9742	0.9431	1
left cingulate cortex	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 lobe	-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 lobe	-0.8298	0.1036	-8.013	-1.033	-0.6269	16
right thalmus	-0.9238	0.1789	-5.164	-1.274	-0.5732	6

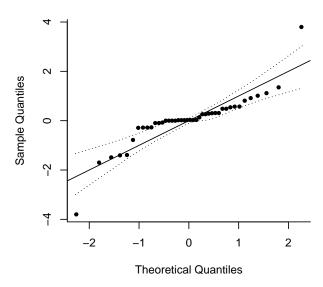
	HedgeG	se	zval	ci.lo	ci.up	N
right tract	-1.004	0.2156	-4.656	-1.427	-0.5813	4







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

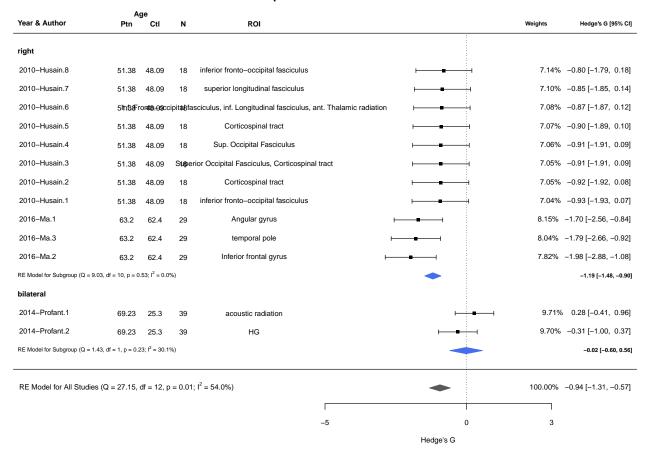


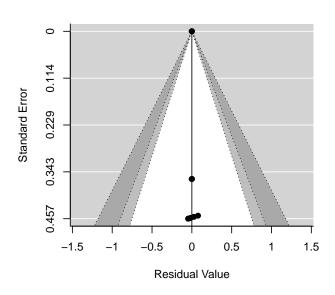
## ACQUIRED - White Matter by FA fractional anisotropy

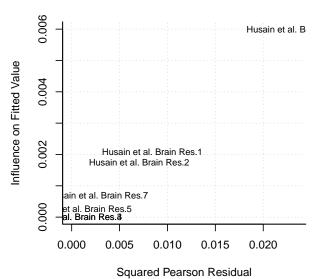
Random effects model no intercept covariated by Big area

	HedgeG	se	zval	ci.lo	ci.up	N
right frontal	-0.8736	0.5076	-1.721	-1.868	0.1212	1
right frontal lobe	-1.976	0.4587	-4.308	-2.875	-1.077	1
right occipital	-0.9105	0.3603	-2.527	-1.617	-0.2042	2
right parietal lobe	-1.703	0.4379	-3.888	-2.561	-0.8443	1
right temporal lobe	-1.793	0.4446	-4.034	-2.665	-0.922	1
right tract	-0.8812	0.2272	-3.878	-1.326	-0.4359	5

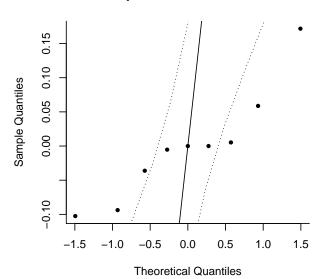
#### acquired White Matter FA





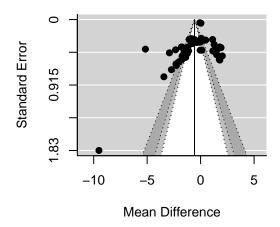


## acquired White Matter FA

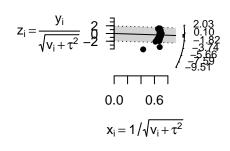


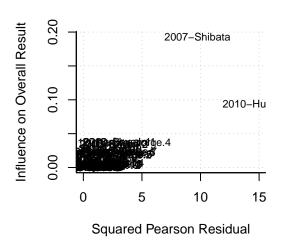
# Supplementary material: heterogeneity per model

## Heterogeney: GM volume Right

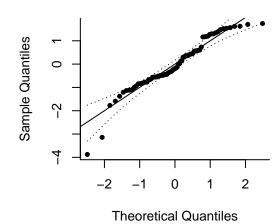


## **GM** volume Right

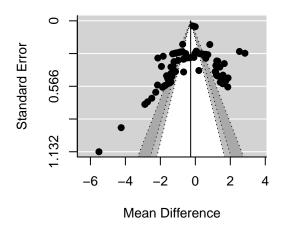




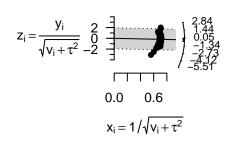
## **GM** volume Right



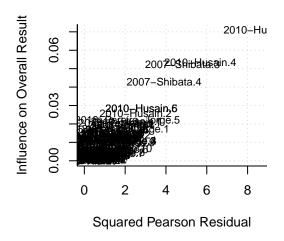
#### Heterogeney: GM volume Left

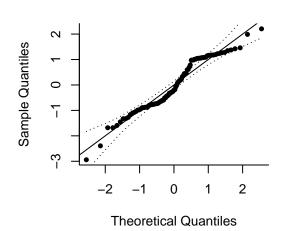


#### **GM volume Left**

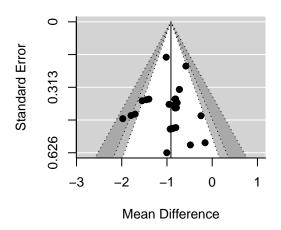


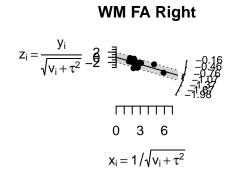
## **GM volume Left**

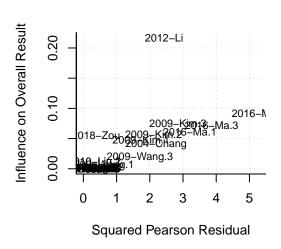


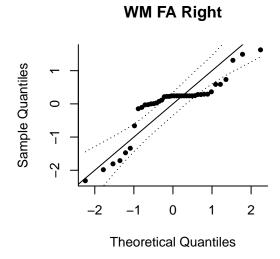


## Heterogeney: WM FA Right

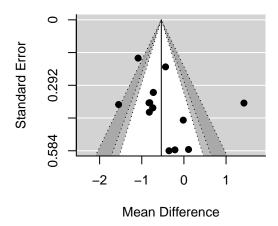




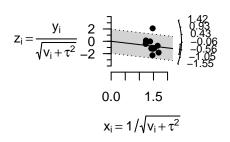




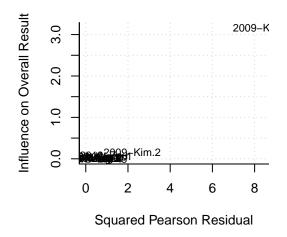
## Heterogeney: WM FA Left

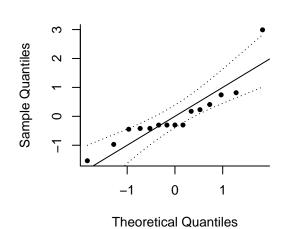


#### **WM FA Left**

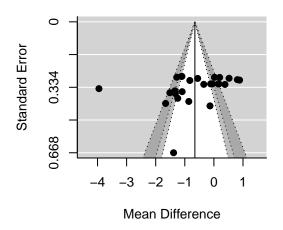


## WM FA Left

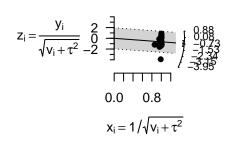




## Heterogeney: WM volume Right



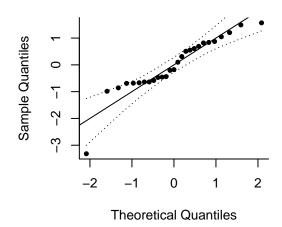
## **WM volume Right**



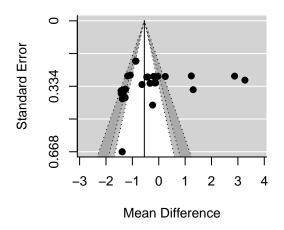
# Influence on Overall Result 0.0 0.2 0.4 0.6 0.8 0 2 4 6 8 10

Squared Pearson Residual

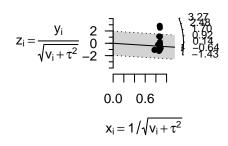
# **WM volume Right**



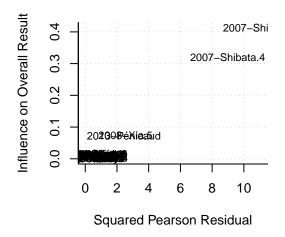
## Heterogeney: WM volume Left



#### **WM volume Left**



## **WM volume Left**



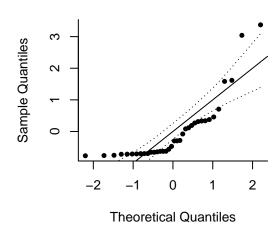
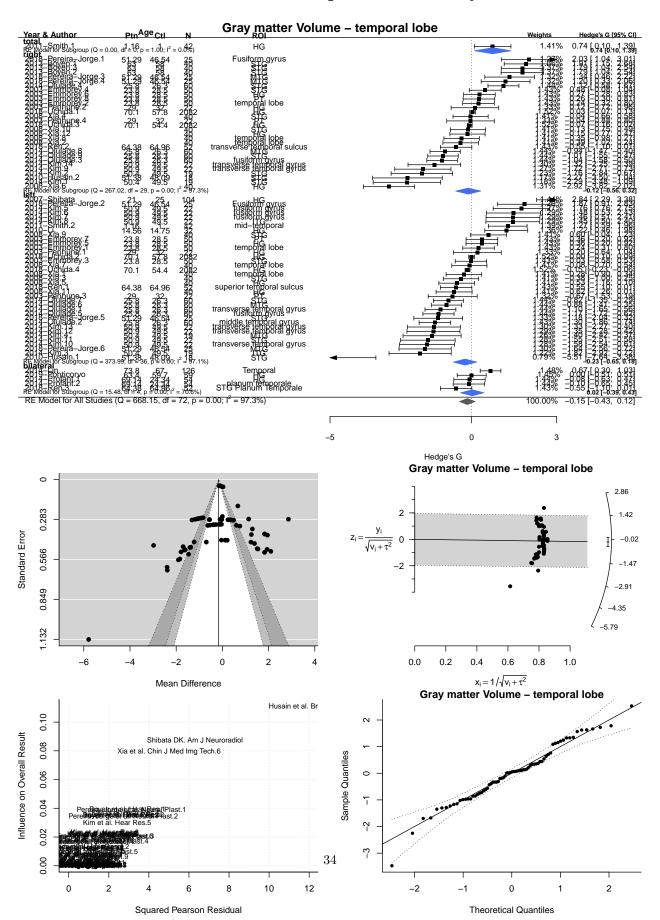
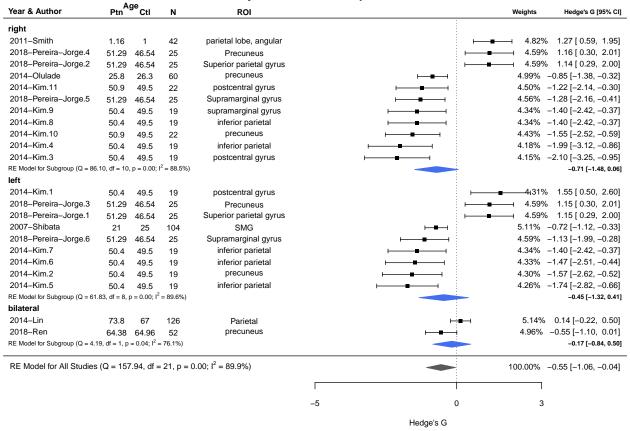
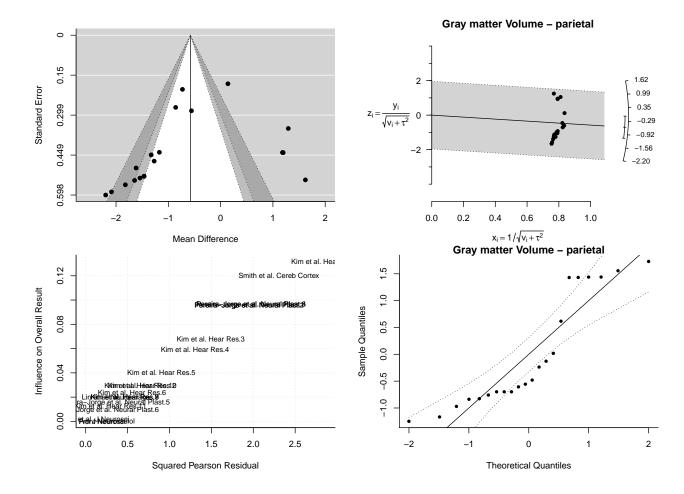


Figure 4 - Meta-regressions of Gray Matter Volume & Brain Areas: Random effects model no intercept covariated by Side

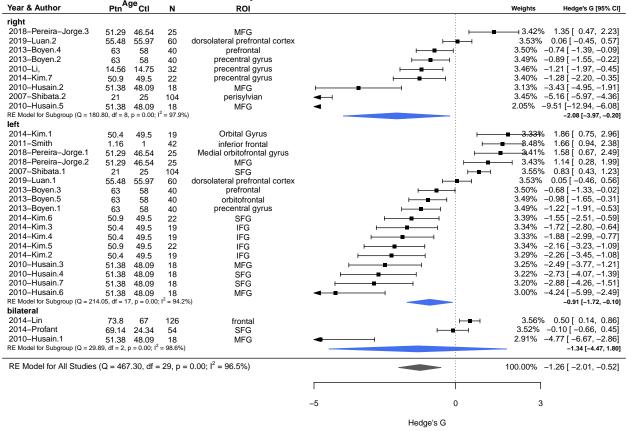


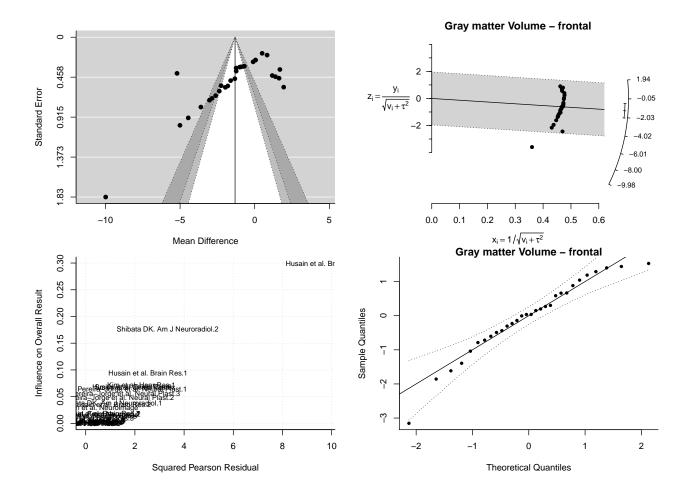
#### Gray matter Volume - parietal





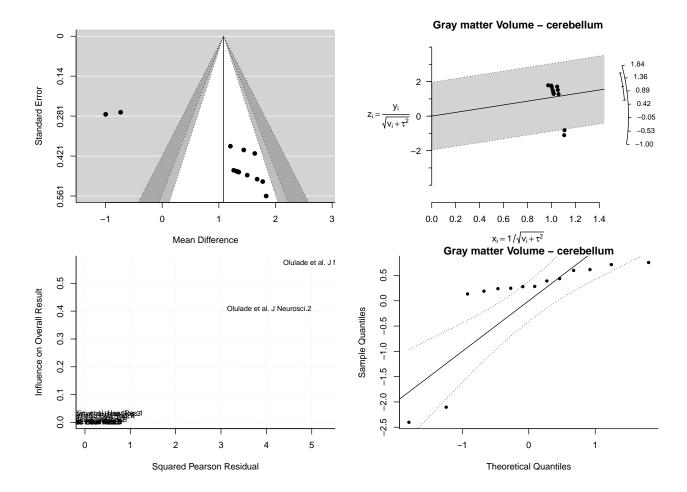
### **Gray matter Volume - frontal**





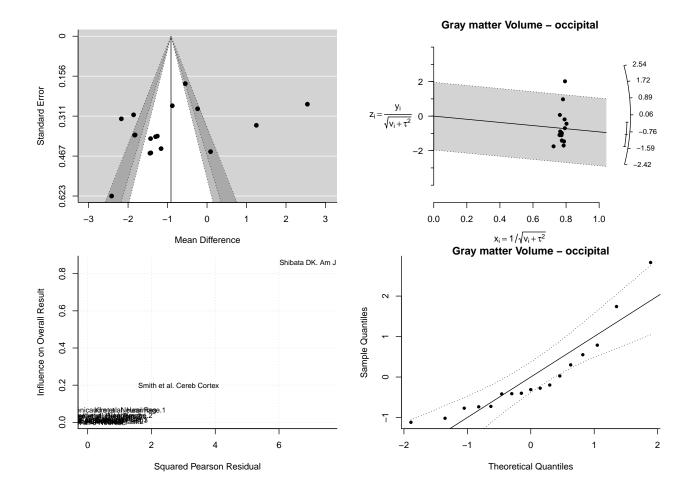
# Gray matter Volume – cerebellum

Year & Author	A: Ptn	ge 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			<del></del>	1.59 [ 0.79, 2.39]
RE Model for Subgroup (Q = 0			= 0.0%)	·				1.67 [1.13, 2.21]
left								
2014-Kim.4	50.9	49.5	22	Culmen			<b>——6</b> ,74%	1.61 [ 0.64, 2.59]
2014-Kim.8	50.9	49.5	22	Declive			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			<b>⊢</b> 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			<b>⊢</b> 7.52%	1.17 [ 0.42, 1.93]
2014-Olulade.2	25.8	26.3	60	cerebellum		<b>⊢</b> ■	8.27% -	0.72 [-1.25, -0.20]
2014-Olulade.1	25.8	26.3	60	cerebellum		<del>⊢=</del>	8.23% -	0.98 [-1.52, -0.45]
RE Model for Subgroup (Q = 8	30.79, df = 10,	p = 0.00;	<sup>2</sup> = 82.9%)	)				0.88 [0.30, 1.46]
RE Model for All Studie	es (Q = 96.	30, df =	13, p = 0	0.00; I <sup>2</sup> = 80.5%)			100.00%	1.04 [ 0.54, 1.53]
					<b>-</b> 5	C	3	
						Hedge's G		

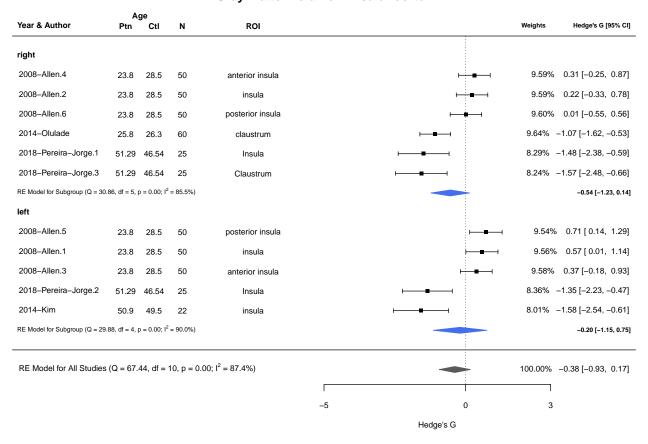


# Gray matter Volume – occipital

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



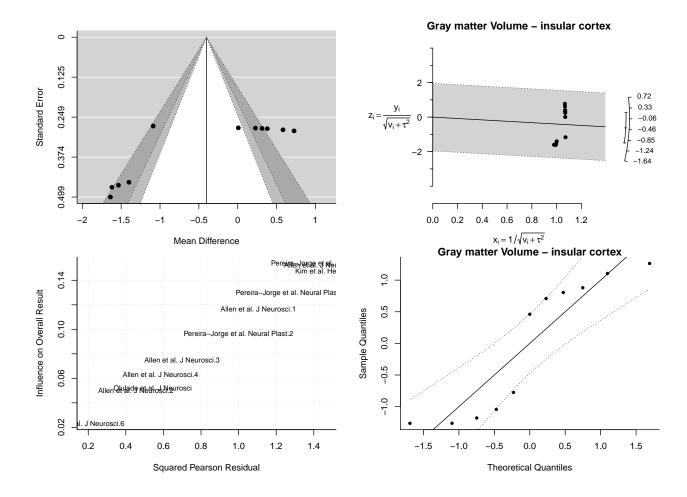
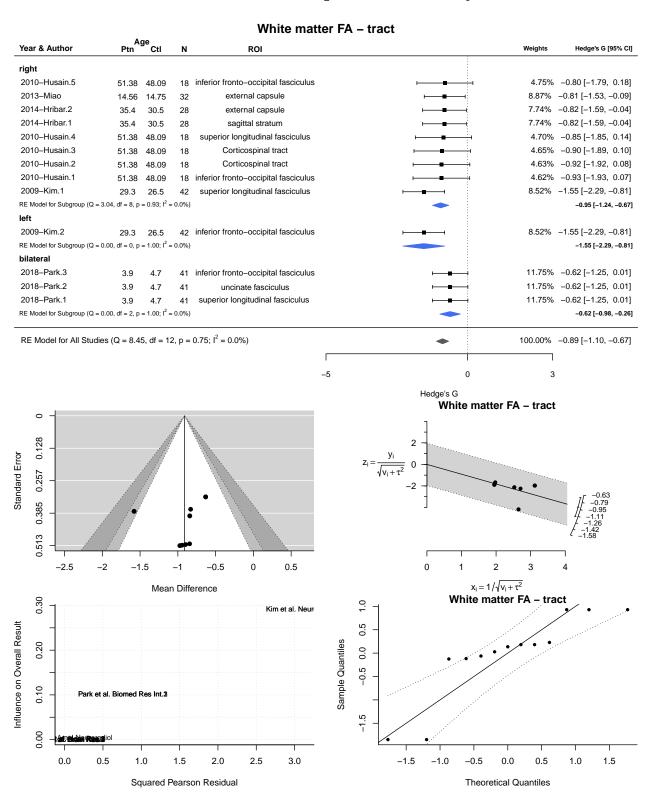
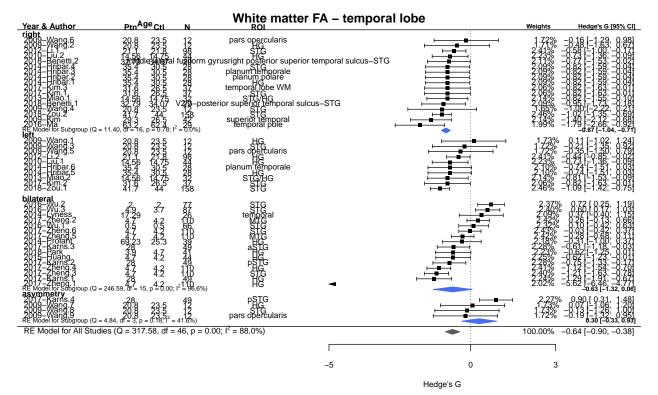
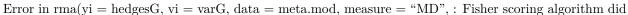
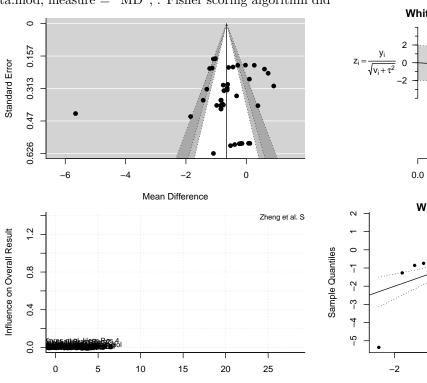


Figure 5 - Meta-regressions of White Matter FA & Brain Areas: Random effects model no intercept covariated by Side





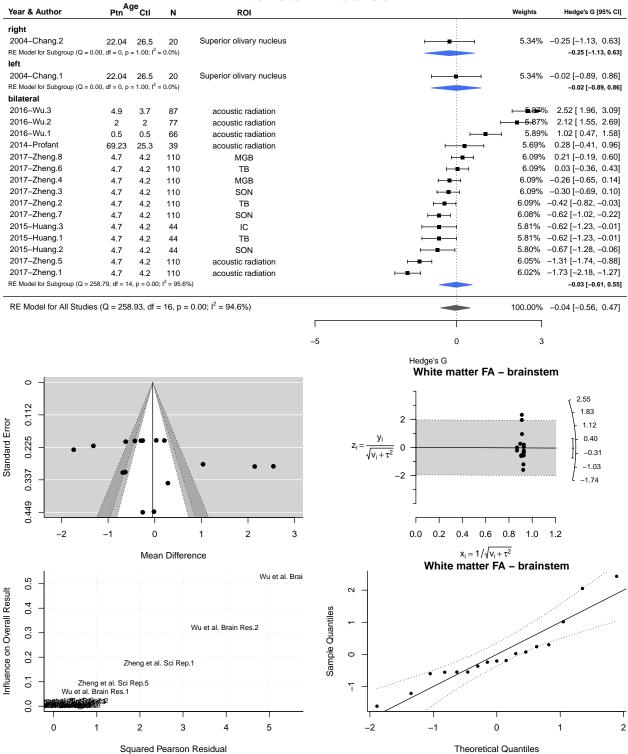




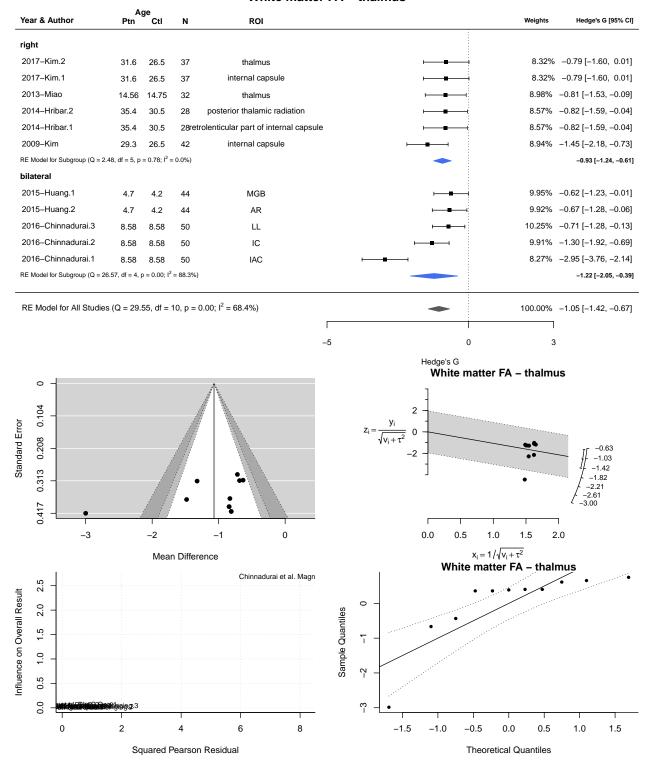
Squared Pearson Residual

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

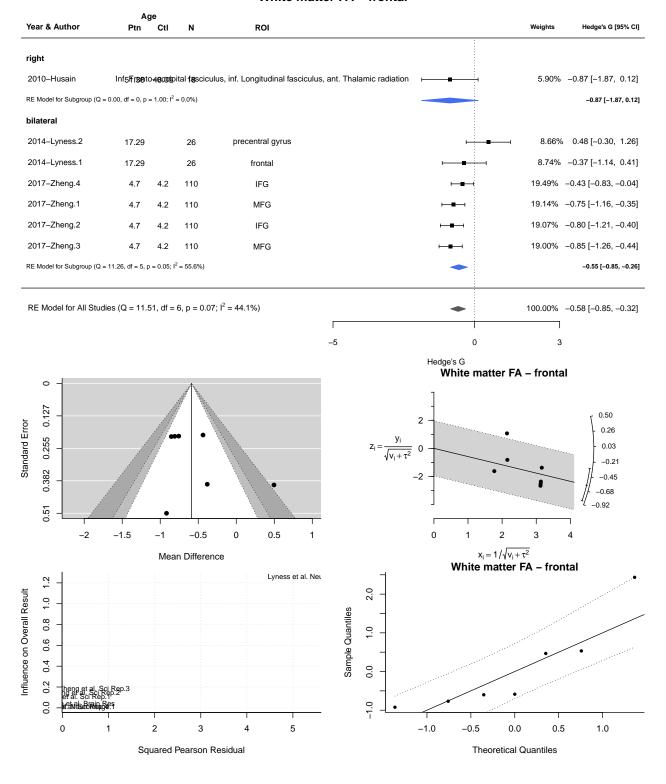
### White matter FA - brainstem



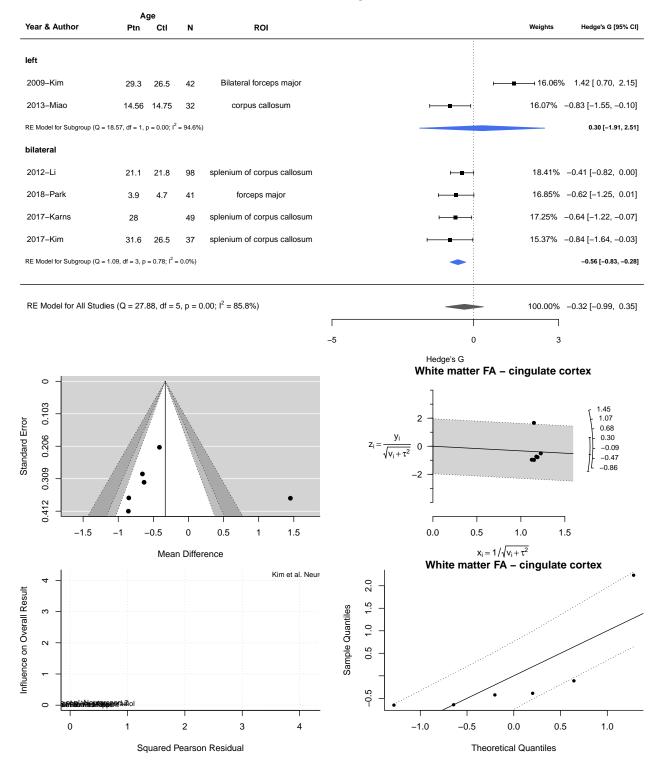
### White matter FA - thalmus



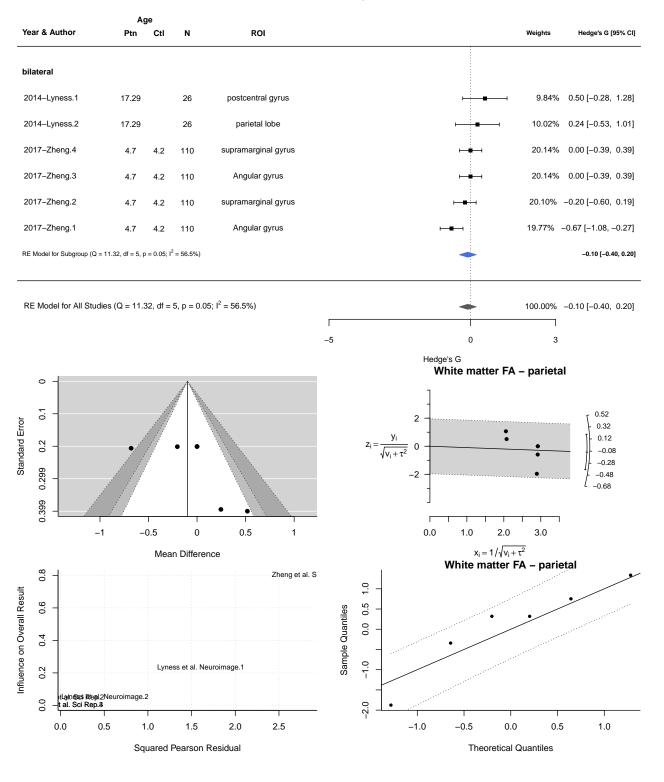
### White matter FA - frontal



## White matter FA - cingulate cortex



## White matter FA - parietal



## White matter FA - occipital

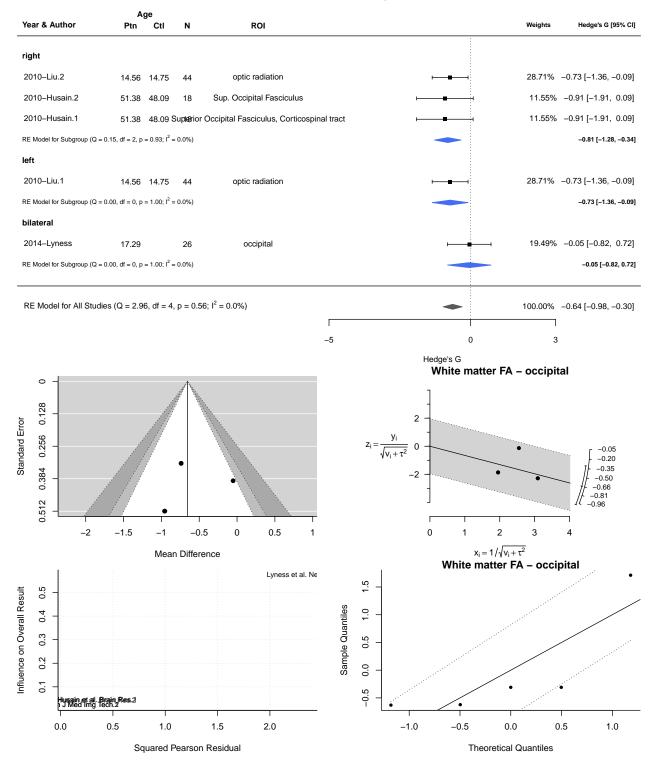
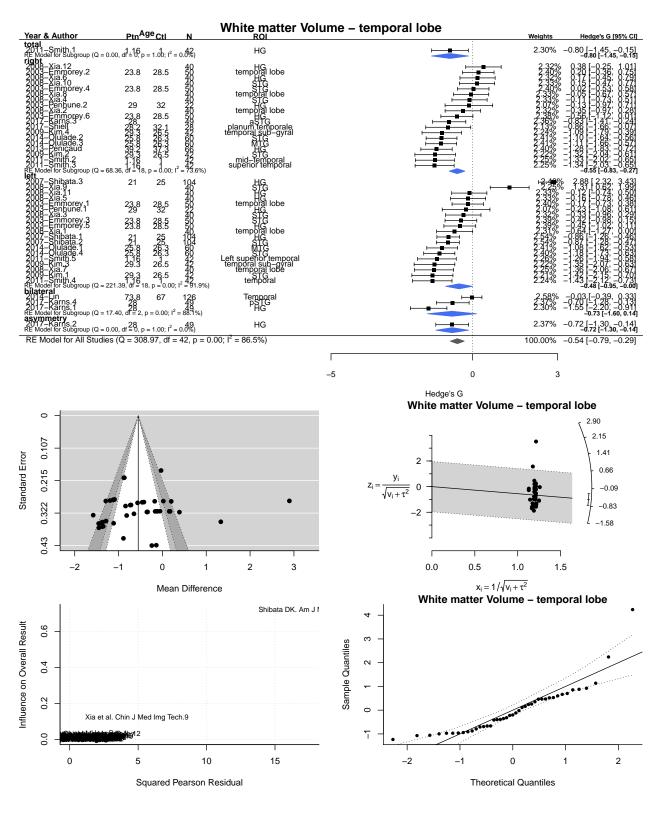
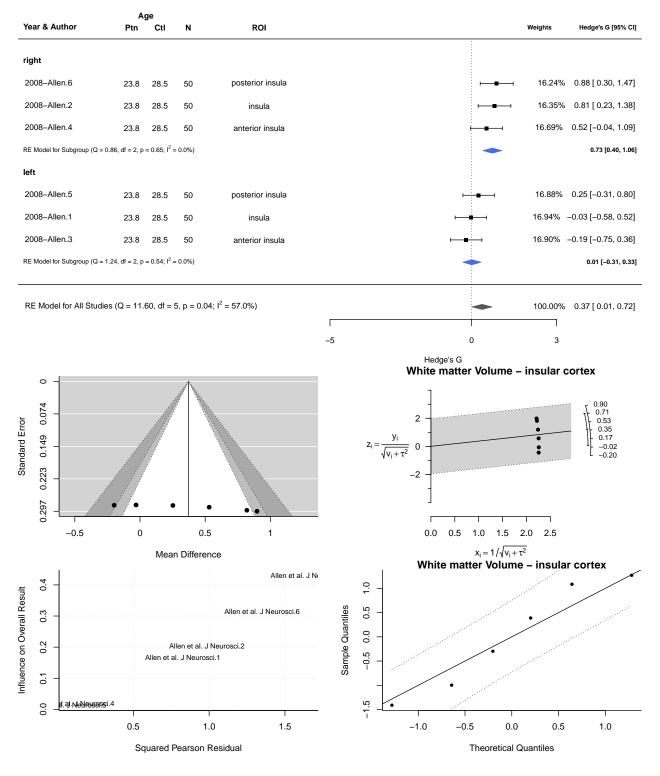


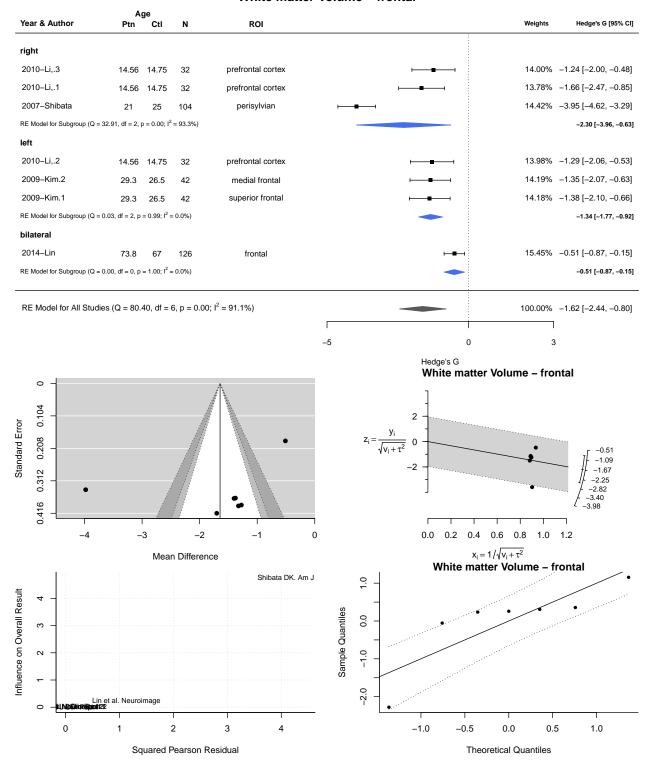
Figure 6 - Meta-regressions of White Matter Volume & Brain Areas: Random effects model no intercept covariated by Side



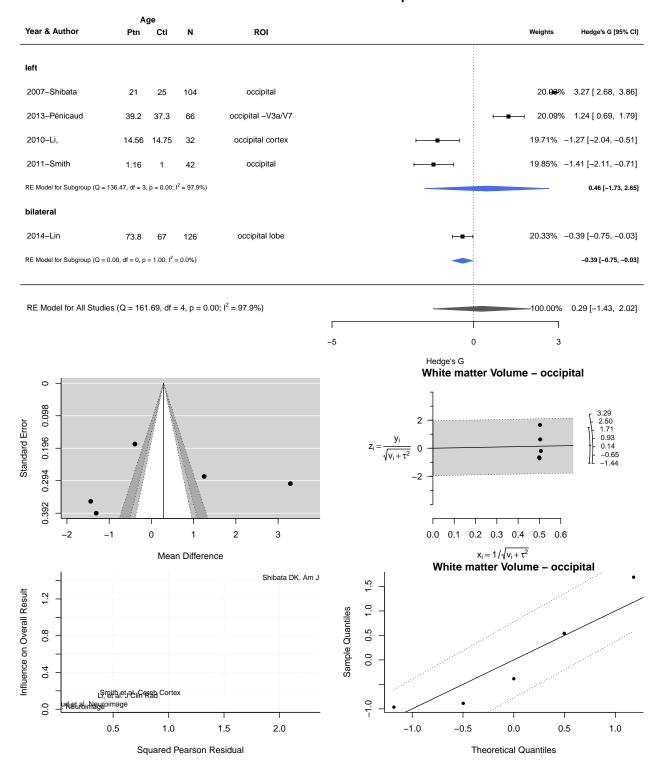
### White matter Volume - insular cortex



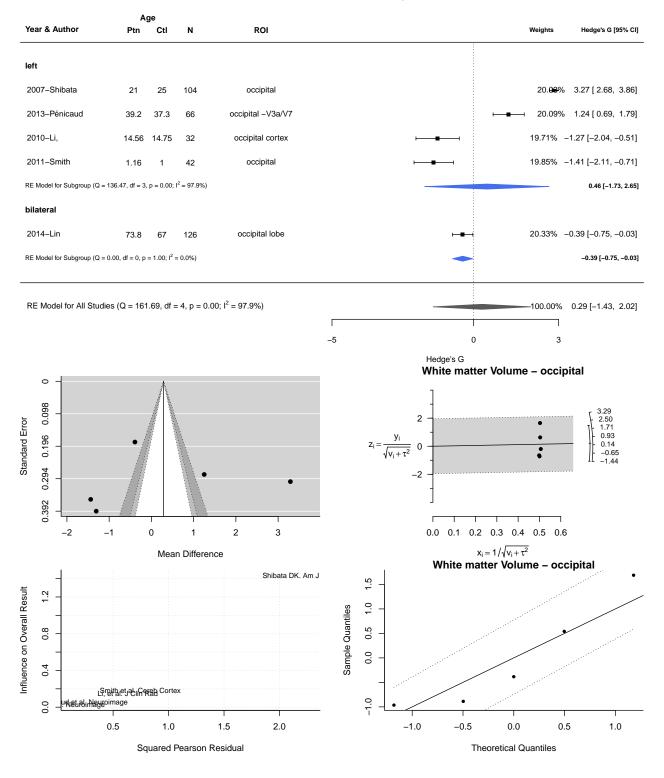
### White matter Volume - frontal



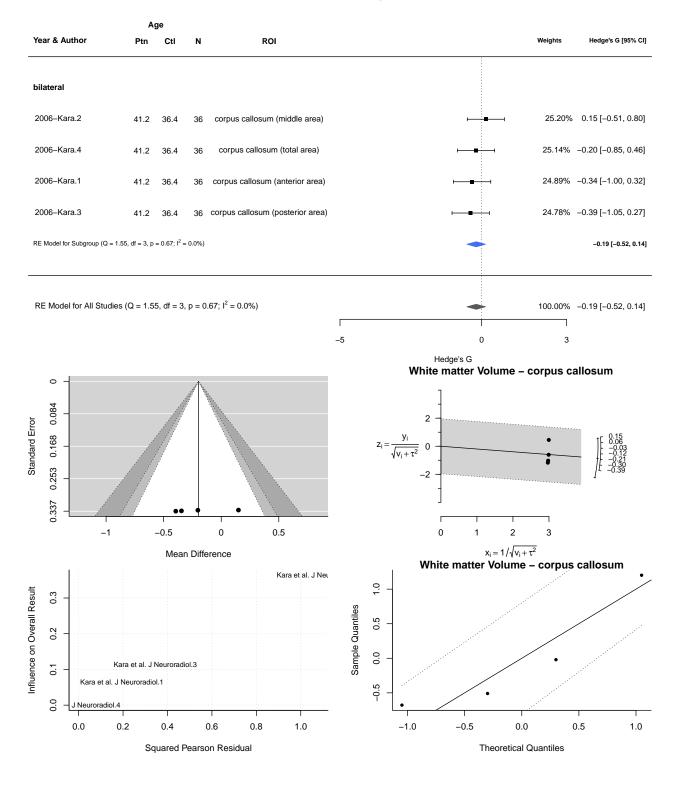
### White matter Volume - occipital



## White matter Volume - occipital



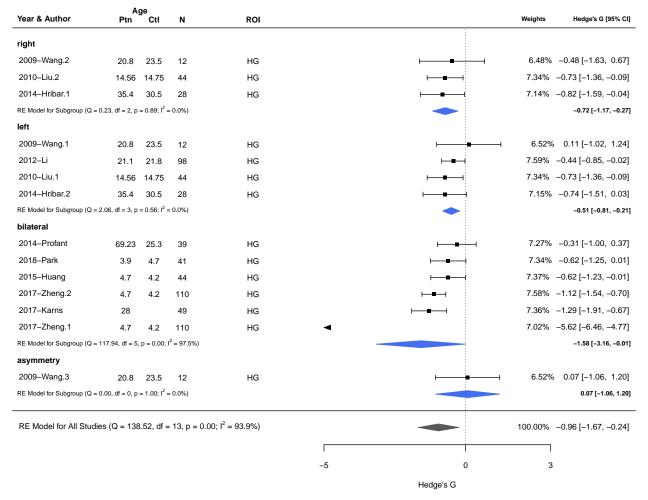
# White matter Volume - corpus callosum



# Supplementary material: Forest-plots of other Measures

# Hesch gyrus FA white matter

### White matter FA and HG



# STG Volume White matter

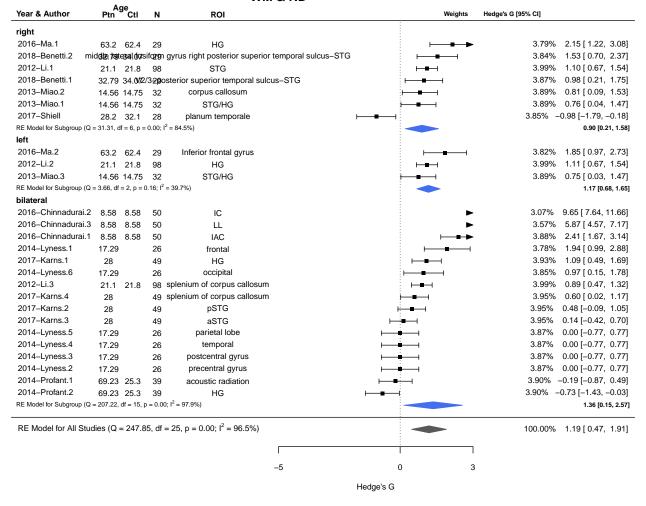
# White matter FA and STG

Year & Author	A: Ptn	ge Ctl	N	ROI			Weights	Hedge's G [95% CI]
right								
2012–Li	21.1	21.8	98	STG		<del>⊢■</del>	8.31%	-0.58 [-1.00, -0.17]
2014-Hribar	35.4	30.5	28	STG		<b>⊢</b>	6.55%	-0.82 [-1.59, -0.04]
2017-Kim.1	31.6	26.5	37	STG		<b>├──</b>	6.38%	-0.82 [-1.63, -0.01]
2009-Wang.2	20.8	23.5	12	STG		<b>├───</b>	4.55%	-1.00 [-2.22, 0.21]
2018-Zou.2	41.7	44	158	STG		<b>⊢</b> ■-1	8.65%	-1.02 [-1.35, -0.69]
RE Model for Subgroup (Q = 2.	65, df = 4, p	= 0.62; I <sup>2</sup>	= 13.4%)			•		-0.84 [-1.10, -0.58]
left								
2009-Wang.1	20.8	23.5	12	STG		<b>├</b>	4.87%	-0.21 [-1.35, 0.92]
2017-Kim.2	31.6	26.5	37	STG		<b>├──</b>	6.38%	-0.82 [-1.63, -0.01]
2018-Zou.1	41.7	44	158	STG		<del> -■- </del>	8.64%	-1.09 [-1.42, -0.75]
RE Model for Subgroup (Q = 2.	29, df = 2, p	= 0.32; I <sup>2</sup>	= 11.6%)			•		-0.95 [-1.31, -0.59]
bilateral								
2016-Wu.2	2	2	77	STG		<b>├</b> - <b>=</b>	8.06%	0.72 [ 0.25, 1.19]
2016-Wu.3	4.9	3.7	87	STG		<del></del>	8.24%	0.60 [ 0.17, 1.03]
2016-Wu.1	0.5	0.5	66	STG		<del>- ■  </del>	7.80%	0.10 [-0.42, 0.63]
2017-Zheng.2	4.7	4.2	110	STG		<del>-</del>	8.41%	-0.03 [-0.42, 0.37]
2017-Zheng.1	4.7	4.2	110	STG		<del>⊢=</del>	8.27%	-1.21 [-1.63, -0.78]
RE Model for Subgroup (Q = 47	7.53, df = 4, p	p = 0.00; I	<sup>2</sup> = 91.3%)					0.03 [-0.64, 0.71]
asymmetry								
2009-Wang.3	20.8	23.5	12	STG		<b>⊢</b>	4.88%	-0.13 [-1.26, 1.00]
RE Model for Subgroup (Q = 0.	00, df = 0, p	= 1.00; I <sup>2</sup>	= 0.0%)					-0.13 [-1.26, 1.00]
RE Model for All Studies	s (Q = 96.	47, df =	13, p = 0.00;	$I^2 = 84.5\%$ )		•	100.00%	-0.44 [-0.80, -0.08]
						i		
					-5	0	3	
						Hedge's G		

# Measures of White matter Integrity

## White matter: RD

WM & RD



# White matter: MD

WM & MD

Year & Author	Aç Ptn	ge Ctl	N	ROI			Weights	Hedge's G [95% CI]
right								
2016-Ma.1	63.2	62.4	29	HG			<del>- 5.38</del> %	2.14 [ 1.22, 3.07]
2017-Shiell	28.2	32.1	28	planum temporale		<b>⊢</b>	5.72% -	-0.85 [-1.65, -0.06]
RE Model for Subgroup (Q =	23.21, df = 1, p	o = 0.00; I <sup>2</sup>	<sup>2</sup> = 95.7%)					<b>0.63</b> [ <b>-2.30</b> , <b>3.57</b> ]
left								
2016-Ma.2	63.2	62.4	29	Inferior frontal gyrus			<del>- 5.1</del> 9%	1.81 [ 0.93, 2.68]
RE Model for Subgroup (Q =	0.00, df = 0, p	= 1.00; I <sup>2</sup>	= 0.0%)					1.81 [0.93, 2.68]
bilateral								
2014-Lyness.1	17.29		26	frontal			<del>■ 5.2</del> 8%	1.94 [ 0.99, 2.88]
2014-Lyness.6	17.29		26	occipital		<b>—</b>	5.65%	0.97 [ 0.15, 1.78]
2014-Lyness.5	17.29		26	parietal lobe		ı .	5.74%	0.61 [-0.18, 1.40]
2015-Huang.3	4.7	4.2	44	IC		<b>⊢</b> ■	→ 6.27%	0.28 [-0.32, 0.88]
2015-Huang.1	4.7	4.2	44	ТВ		<b>⊢</b> ■	→ 6.27%	0.20 [-0.40, 0.79]
2015-Huang.2	4.7	4.2	44	SON		<b>⊢</b>	6.27%	0.09 [-0.51, 0.68]
2015-Huang.6	4.7	4.2	44	HG		· ·	6.28%	0.00 [-0.59, 0.59]
2015-Huang.5	4.7	4.2	44	AR		<b>⊢</b>	6.28%	0.00 [-0.59, 0.59]
2015-Huang.4	4.7	4.2	44	MGB		<b>⊢</b>	6.28%	0.00 [-0.59, 0.59]
2014-Lyness.4	17.29		26	temporal		ı	→ 5.79%	0.00 [-0.77, 0.77]
2014-Lyness.2	17.29		26	precentral gyrus		<b>⊢</b>	→ 5.79%	0.00 [-0.77, 0.77]
2014-Profant.1	69.23	25.3	39	acoustic radiation		<b>⊢</b>	6.04%	-0.04 [-0.72, 0.64]
2014-Profant.2	69.23	25.3	39	HG		<b>├──</b>	6.00%	-0.64 [-1.34, 0.06]
2014-Lyness.3	17.29		26	postcentral gyrus		<b>⊢</b>	5.52% -	-1.37 [-2.23, -0.51]
RE Model for Subgroup (Q =	37.15, df = 13,	p = 0.00;	$I^2 = 69.6\%$ )			•		0.12 [-0.21, 0.46]
RE Model for All Stud	ies (Q = 75.0	05, df =	16, p = 0.	.00; I <sup>2</sup> = 83.4%)		•	100.00%	0.28 [-0.15, 0.70]
						<u> </u>		
					-5	0	3	
						Hedge's G		

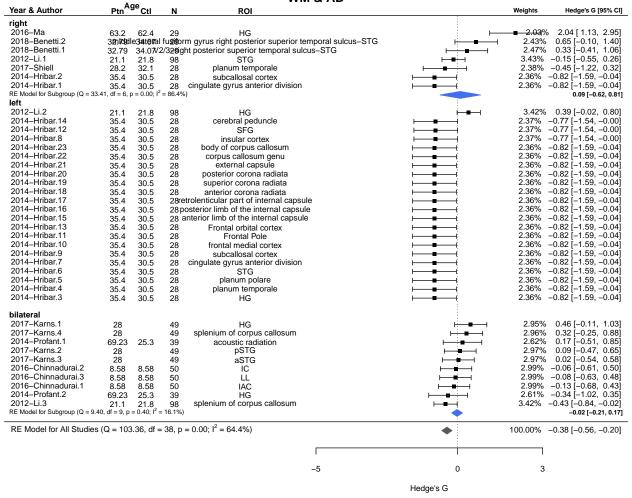
# White matter: Mean Kurtosis

WM & Mean Kurtosis

Year & Author	A Ptn	ge Ctl	N	ROI		Weights	Hedge's G [95% CI]
bilateral							
2017-Zheng.9	4.7	4.2	110	STG	<b>⊢=</b> I	3.77%	0.60 [ 0.20, 1.00]
2017-Zheng.19	4.7	4.2	110	IFG	<b>⊢=</b> -1	3.79%	0.38 [-0.01, 0.78]
2017-Zheng.17	4.7	4.2	110	HG	: [─■─┤	3.79%	0.38 [-0.02, 0.78]
2017-Zheng.18	4.7	4.2	110	MFG	<b>⊢=</b> →	3.80%	0.22 [-0.17, 0.62]
2017-Zheng.13	4.7	4.2	110	acoustic radiation	<del>⊢i≡</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	<b>⊢</b>	3.81%	0.00 [-0.39, 0.39]
2017-Zheng.23	4.7	4.2	110	supramarginal gyrus	⊢÷	3.81%	0.00 [-0.39, 0.39]
2017-Zheng.22	4.7	4.2	110	Angular gyrus	<b>⊢</b>	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> = <del>:</del> -1	3.81%	-0.11 [-0.50, 0.28]
2017-Zheng.20	4.7	4.2	110	MTG	<del> -■</del> -1	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	ТВ	<b>⊢=</b> i	3.79%	-0.46 [-0.86, -0.07]
2017-Zheng.4	4.7	4.2	110	MGB	<b>⊢=</b> -1	3.79%	-0.47 [-0.86, -0.07]
2016-Chinnadurai.1	8.58	8.58	50	IAC	<del>  ■  </del>	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	<b>⊢=</b> —i	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	⊢ <del>≡</del> ⊣	3.78%	-0.58 [-0.98, -0.18]
2017-Zheng.3	4.7	4.2	110	SON	⊢■→	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	⊢■→	3.74%	-0.84 [-1.25, -0.44]
2016-Chinnadurai.3	8.58	8.58	50	LL	<b>├─ड</b> ─┤	3.01%	-0.92 [-1.51, -0.34]
2017-Zheng.10	4.7	4.2	110	Angular gyrus	<b>⊢=</b> ⊣	3.70%	-1.07 [-1.49, -0.65]
RE Model for Subgroup (Q = 1	10.59, df = 20	6, p = 0.00	$I^2 = 76.6\%$	s)	•		-0.29 [-0.45, -0.12]
RE Model for All Studies	s (Q = 110	D.59, df =	= 26, p = 0	$0.00; I^2 = 76.6\%)$	*	100.00%	-0.29 [-0.45, -0.12]
					İ		
					<b>-</b> 5 0	3	
					Hedge's G		

### White matter: AD

WM & AD

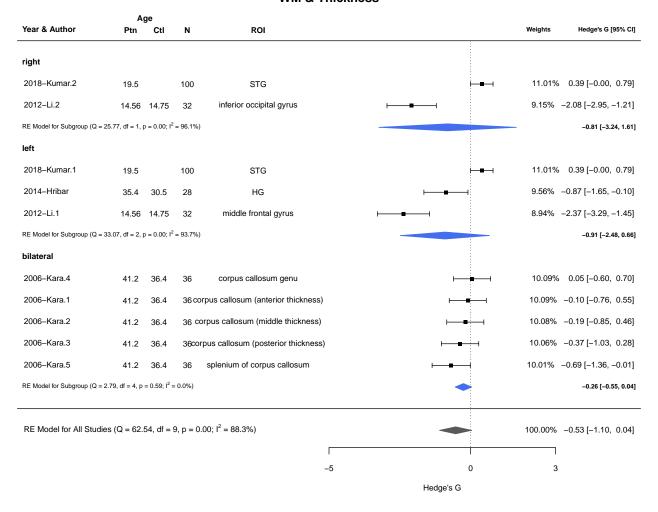


Error in rma(yi = hedgesG, vi = varG, data = meta.mod, measure = "MD", : Fisher scoring algorithm did not converge. See 'help(rma)' for possible remedies.

# 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