

**MACROSCALE THALAMIC FUNCTIONAL ORGANIZATION DISTURBANCES AND
UNDERLYING CORE CYTOARCHITECTURE IN EARLY-ONSET SCHIZOPHRENIA**

Running title: Expansion of thalamic functional hierarchies in schizophrenia

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1 **Abstract**

2 Schizophrenia is a polygenic mental disorder with heterogeneous positive and negative
3 symptom constellations, and is associated with abnormal cortical connectivity. The thalamus
4 has a coordinative role in cortical function and is key to the development of the cerebral
5 cortex. Conversely, altered functional organization of the thalamus might relate to
6 overarching cortical disruptions in schizophrenia, anchored in development. Here, we
7 contrasted resting-state fMRI in 99 antipsychotic-naïve first-episode early-onset
8 schizophrenia (EOS) patients and 100 typically developing controls to study whether
9 macroscale thalamic organization is altered in EOS. Employing dimensional reduction
10 techniques on thalamocortical functional connectome, we derived lateral-medial and
11 anterior-posterior thalamic functional axes. We observed increased segregation of
12 macroscale thalamic functional organization in EOS patients, which was related to altered
13 thalamocortical interactions both in unimodal and transmodal networks. Using an *ex vivo*
14 approximation of core-matrix cell distribution, we found that core cells particularly underlie
15 the macroscale abnormalities in EOS patients. Moreover, the disruptions were associated
16 with schizophrenia-related gene expression maps. Behavioral and disorder decoding
17 analyses indicated that the macroscale hierarchy disturbances might perturb both perceptual
18 and abstract cognitive functions and contribute to negative syndromes in schizophrenia,
19 suggesting a unitary pathophysiological framework of schizophrenia.

20

21 **Keywords:** *cytoarchitectural; early-onset schizophrenia; functional hierarchy; genetic;*
22 *thalamus.*

23 **Introduction**

24 Schizophrenia is a polygenic psychiatric illness characterized by a combination of
25 psychotic symptoms and motivational/ cognitive deficits, which usually emerge during early
26 adulthood (1). Over the past two decades, a wealth of neuroimaging studies has indicated
27 that schizophrenia can be associated with pathological interactions across widely distributed
28 brain regions, instead of focal brain damages. Accordingly, the overarching dysconnection
29 hypothesis posits that schizophrenia results from brain structural and functional connectivity
30 abnormalities (2). The thalamus, which is well-placed to arbitrate the interactions between
31 distributed brain organization (3), might play a pivotal role in the pathophysiological process
32 of schizophrenia (4, 5).

33

34 The thalamus is a cytoarchitecturally heterogeneous diencephalic structure that contains an
35 admixture of Parvalbumin (PVALB)-rich ‘Core’ cells and Calbindin (CALB1)-rich ‘Matrix’
36 cells (6). Whereas core cells preferentially target granular layers (Layers III and IV) of
37 unimodal primary regions, such as primary visual, auditory and somatosensory cortices,
38 matrix cells target supragranular layers (Layers I-III) over wide areas in a diffuse pattern (7).
39 This means that distinct thalamic cells may interact with cortical areas organized into
40 different topological zones (8). Cellular-scale information of the thalamus may be a critical
41 factor to understand the thalamocortical interactions that supports cognition and behavior.

42 Indeed, the thalamocortical system has been suggested to form the basis for binding multiple
43 sensory experiences into a single framework of consciousness (9). By coordinating the

44 modular architecture of cortical networks, the thalamus has been reported to be engaged in
45 integrating information processing within the whole cerebral cortex (10).

46

47 In accordance with its function coordinating cortical network organization, the thalamus
48 plays the central role in the development of the cerebral cortex (11). During brain
49 development, the thalamus changes in concordance with the cerebral cortex and disturbances
50 of this coordinated process relate to cognitive dysfunctions (12), serving as a precursor of
51 schizophrenia. Thalamocortical dysconnectivity patterns, characterized by hypoconnectivity
52 with prefrontal regions and hyperconnectivity with sensorimotor areas, have been reported
53 in both pediatric (13) and early-stage (14) patients with schizophrenia. The dysconnectivity
54 pattern has been hypothesized to arise from disturbed brain maturation, particularly during
55 the transition from youth to adulthood (15). Intriguingly, thalamo-prefrontal
56 hypoconnectivity is correlated with thalamo-sensorimotor hyperconnectivity in patients,
57 potentially implying a shared pathophysiological mechanism (14). However, few studies
58 have investigated thalamocortical connectivity in the still developing brain of schizophrenia
59 from a comprehensive perspective.

60

61 Recently, the application of dimension reduction techniques has emerged as a promising
62 strategy for holistic representations of brain connectivity. These novel data-driven methods
63 decompose high dimensional connectome into a series of low dimensional axes capturing
64 spatial gradients of connectivity variations (16, 17). The gradient framework describes a
65 continuous coordinate system, in contrast to clustering-based methods resulting in discrete

66 communities (18). Using these methods in the context of cortex-wide functional connectome,
67 previous studies observed a cortical hierarchy that spans from unimodal primary regions to
68 transmodal regions (19), which has a close link with cortical microstructure like
69 cytoarchitecture or myeloarchitecture (20). Their coupling along the unimodal-transmodal
70 axis has been reported to be genetically- and phylogenetically-controlled, supporting flexible
71 cognitive functions (21). Perturbed macroscale cortical functional hierarchies have been
72 reported in various neurological (22, 23) and psychiatric disorders (18) including
73 schizophrenia (24). Also, thalamic hierarchies have been previously derived from
74 thalamocortical connectome, identifying a lateral-medial (L-M) principal gradient and an
75 anterior-posterior (A-P) secondary gradient (25). The L-M axis captures thalamic anatomical
76 nuclei differentiation, while the A-P axis characterizes unimodal-transmodal functional
77 hierarchy. Also the coupling between core-matrix cytoarchitecture and functional connectome
78 has been shown to describe the unimodal-transmodal cortical gradients, and argued to play a
79 major role in shaping functional dynamics within the cerebral cortex (8). Given the possible
80 implication of the thalamus in schizophrenia, thalamic hierarchies may be altered during
81 brain maturation and could provide new insights into the disrupted thalamocortical
82 organization in schizophrenia.

83

84 Here, we leveraged a cohort of individuals with early-onset schizophrenia (EOS), a disorder
85 that is neurobiologically continuous with its adult counterpart (26), to examine whether
86 macroscale thalamic functional organization shows disturbances in the still developing brain
87 of schizophrenia, mirroring neocortical reports. To this end, we first evaluated functional

88 hierarchies of the thalamus by employing dimension reduction techniques on
89 thalamocortical functional connectome (27). We then embedded thalamic functional
90 hierarchies in a neurobiological context by spatially correlating the macroscale patterns with
91 gene expression maps from the Allen Human Brain Atlas (AHBA) (28). Last, we tested
92 whether the functional hierarchies could estimate clinical symptoms of patients using a
93 machine learning regression strategy.

94

95 **Materials and Methods**

96 **Participants**

97 Ninety-nine antipsychotic-naïve first-episode EOS patients and 100 typically developing
98 (TD) controls were recruited from the First Hospital of Shanxi Medical University and the
99 local community through advertisements, respectively. All pediatric participants were 7–17
100 years old. The diagnosis of schizophrenia was in accordance with the Structured Clinical
101 Interview for DSM-IV, and was confirmed by at least one senior psychiatrist (Y.X.) through
102 a structured clinical interview after at least 6-months of follow-up. The psychiatric
103 symptomatology of 71 patients was evaluated using the Positive and Negative Syndrome
104 Scale (PANSS). Exclusion criteria for all subjects included 1) ≥ 18 years old; 2)
105 neurological MRI anomalies; 3) any electronic or metal implants; or 4) substance abuse. In
106 addition, EOS patients were excluded if they suffered from the illness for > 1 year, and TD
107 controls were excluded if they and their first-degree relatives had any history of psychiatric
108 disorder. This retrospective study was approved by the Ethics Committee of the First
109 Hospital of Shanxi Medical University. Written consent was obtained from every participant
110 and their parents or legal guardians.

111

112 Four patients and two controls were excluded from the study because of incomplete
113 scanning data, nine patients and six controls due to excessive head motion [mean frame-wise
114 displacement, (FD) > 0.2 mm or outliers $> 50\%$], and one control due to poor quality of
115 intrasubject brain registration. Ultimately, 86 EOS patients and 91 demographically-matched
116 TD controls were included in the analysis (See **Table 1** for detailed demographic data).

118

Table 1. Demographics and clinical data.

Characteristic	EOS	TD	Group comparisons	
			Statistic-Values	P-Values
Demographic sample	n=86	n=91		
Sex (male/female)	31/55	33/58	0.0009 ^a	0.98
Age (years)	14.56 ± 1.95	14.34 ± 2.02	3711 ^b	0.55
Handedness (right/left)	86/0	91/0	–	–
Mean FD (mm)	0.07 ± 0.03	0.07 ± 0.03	3843 ^b	0.84
Clinical sample	n=65	–		
PANSS total scores	65.4 ± 17.41	–		
PANSS general scores	31.68 ± 8.71	–		
PANSS positive scores	15.14 ± 5.11	–		
PANSS negative scores	14.00 ± 5.55	–		

119

Note: Mean ± SD; ^a The χ^2 value for gender distribution was obtained by chi-square test; ^b

120

The U values were obtained by Mann–Whitney tests.

121

Abbreviations: EOS, early-onset schizophrenia; TD, typically developing; FD, frame-wise

122

displacement; PANSS, Positive and Negative Symptom Scale.

123

124 **MRI acquisition**

125 Multimodal MRI data were collected using a 3 Tesla MRI scanner (MAGNETOM Verio;
126 Siemens, Germany) in the First Hospital of Shanxi Medical University. T1-weighted data
127 were acquired using a three-dimensional fast spoiled gradient-echo sequence [repetition time
128 (TR) = 2,300 ms; echo time (TE) = 2.95 ms; flip angle = 9°; matrix = 256 × 240; slice
129 thickness = 1.2 mm (no gap); and voxel size = 0.9375 × 0.9375 × 1.2 mm³, with 160 axial
130 slices]. Resting-state functional MRI (rs-fMRI) data were obtained using a two-dimensional
131 echo-planar imaging sequence [TR = 2,500 ms; TE = 30 ms; flip angle = 90°; matrix = 64 ×
132 64; number of volumes = 198; slice thickness = 3 mm (1 mm gap); and voxel size = 3.75 ×
133 3.75 × 4 mm³, with 32 axial slices].

134

135 **MRI processing**

136 T1-weighted structural data were preprocessed with FreeSurfer (v7.1.0,
137 <http://surfer.nmr.mgh.harvard.edu/>), which included cortical segmentation and surface
138 reconstruction. Rs-fMRI functional data were preprocessed with the CBIG pipeline
139 (<https://github.com/ThomasYeoLab/CBIG>) based on FSL [v5.0.9, (29)] and FreeSurfer
140 (v7.1.0), which included removal of the first four volumes, slice-timing, motion correction,
141 and boundary-based registration to structural images. See **Supplement 1** for further details.
142 Preprocessed images were then registered to MNI152 template and resampled to the cortical
143 surface using Ciftify package [v2.3.3, (30)]. The thalamus was localized using the Gordon
144 333 Atlas (31), including 2536 voxels across both hemispheres.

145

146 **Macroscale thalamocortical gradient identification**

147 Analogous to previous work (25), gradients of thalamocortical functional connectome were
148 generated to describe thalamic functional organization using the diffusion embedding
149 algorithm in BrainSpace Toolbox (32). Thalamocortical functional connectivities were first
150 calculated based on Pearson correlations between the thalamic and cortical rs-fMRI
151 time-series for each subject (16, 23), and then converted into cosine similarity matrices (16,
152 17). Subsequently, nonlinear dimensionality reduction techniques were employed on
153 similarity matrices to resolve connectome gradient, i.e., spatial axis in connectome
154 variations (27). See **Supplement 1** for detailed analyses. We selected the first two gradients
155 to represent the macroscale thalamic connectome space, which explained 44% of the total
156 eigenvariance in functional connectome (**Figure S2**). The relative positioning of thalamic
157 voxels along each organizational axis describes similarity of their functional connectivity
158 profiles. To quantify the dispersion of each thalamic voxel in the two-dimensional gradient
159 space, we computed eccentricity, i.e., the square root of the Euclidian distance from each
160 thalamic voxel to the center of mass in the two-dimensional gradient space (33). For each
161 individual, global eccentricity was calculated by averaging eccentricity values across all
162 thalamic voxels, indicating overall dispersion of the gradient space. We additionally
163 explored cortical-thalamic gradients by generating cortical similarity matrices from
164 cortical-thalamic connectivity profile (see **Supplement 2** for further details).

165

166 **Thalamic functional community division**

167 To characterize the functional relevance of macroscale thalamic gradient space, we created a
168 thalamic functional atlas using winner-take-all representation approach (25, 34). Partial
169 correlations were computed between rs-fMRI time-series of each thalamic voxel and six
170 cortical functional networks (35) including the visual network (VIS), sensorimotor network
171 (SMN), dorsal attention network (DAN), ventral attention network (VAN), frontoparietal
172 network (FPN), and default mode network (DMN). The limbic network was excluded for
173 low signal quality in corresponding cortical areas of our data. Each thalamic voxel was
174 labeled by the functional network showing the highest partial correlation coefficient.
175 Additionally, to explore cortical correspondences of thalamic gradients, we projected
176 thalamic gradients onto the cerebral cortex. For each cortical vertex, gradient projection was
177 calculated by correlating its cortical-thalamic connectivity profiles with thalamic gradients.
178 These cortical maps were down sampled into 400 cortical parcels and grouped into
179 functional networks (36) to further validate thalamic functional community division.
180

181 **The core-matrix cytoarchitecture**

182 To delineate the core-matrix cytoarchitecture in the thalamus, we used the spatial maps of
183 mRNA expression levels for two calcium-binding proteins (CALB1 and PVALB)
184 (<https://github.com/macshine/corematrix>) (8) generated from post-mortem Allen Human
185 Brain Atlas (28). Thalamic voxels with positive CP values (CALB1-PVALB values) related
186 to matrix projection cells, and voxels with negative values related to core populations.
187 Interregional correlations between CP map and differential eccentricity map were employed
188 to reveal an association between the thalamic cytoarchitecture and disturbed functional

189 organization in EOS patients. The variogram-matching model was used to correct for the
190 spatial autocorrelation of brain maps (37). Subsequently, we projected CP maps onto the
191 cerebral cortex to reveal couplings between the core-matrix cytoarchitecture and functional
192 connectome. Cortical parcels with positive coupling values indicated as preferential
193 associations with matrix thalamic populations, and negative values suggested core
194 populations. To evaluate cognitive terms associated with gene-connectome coupling maps,
195 we further conducted topic-based behavioral decoding using NeuroSynth meta-analytic
196 database (38). See **Supplement 1** for detailed analysis steps.

197

198 **Clinical correspondences**

199 To investigate clinical significance of thalamic functional organization disturbances, we
200 further associated the macroscale functional phenotype with schizophrenia-related genetic
201 expression. Based on previous work (39), we selected out 28 protein-coding genes
202 particularly implicated in schizophrenia etiology or treatment (See **Table S1** for detailed
203 information). Within an overlapping thalamic mask (1969 voxels), estimated mRNA levels
204 were extracted from Allen Human Brain Atlas and then normalized using a robust sigmoid
205 function (40). Gene expression maps were then spatially correlated with differential
206 eccentricity map between the EOS and TD groups, corrected by the variogram-matching
207 model (1000 surrogate maps) as abovementioned.

208

209 Second, we estimated the relationship between thalamic functional hierarchies and clinical
210 presentations in EOS patients. Following a machine learning pipeline, we used the elastic

211 net regression model to predict clinical symptoms in EOS (41). Eccentricity values of the
212 two-dimensional gradient space were defined as input features, and PANSS positive and
213 negative scores were used as predictors. The model performance was evaluated by
214 comparison of observed and predicted clinical scores. See **Supplement 1** for detailed
215 analyses.

216

217 **Group comparisons between the EOS and TD groups**

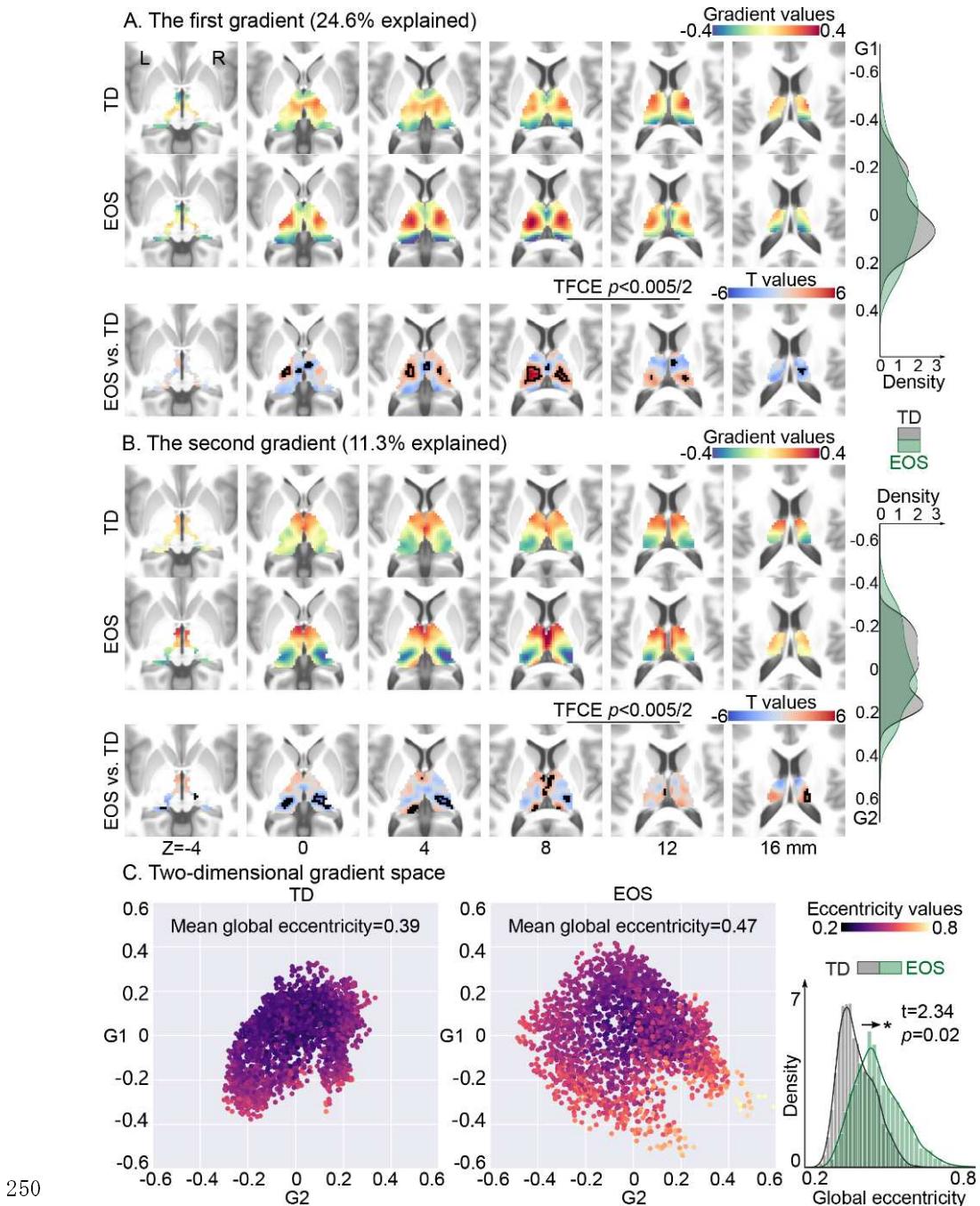
218 Between-group differences on all kinds of measurements were assessed by using two-sample
219 t-tests with covariates including age, gender, and mean FD. The multiple comparison
220 corrections were conducted using three methods for different spatial scales: voxel-wise,
221 parcel-wise, and global-wise. For voxel-wise thalamic gradient and eccentricity values,
222 multiple comparison corrections were employed using the permutation-based threshold-free
223 cluster enhancement (TFCE, 10,000 permutations, $p < 0.005$) method (FSL-PALM), which
224 could improve sensitivity and interpretable than cluster-based thresholding method (42). For
225 parcel-wise gene-connectome coupling values, false discovery rate (FDR) corrections ($p <$
226 0.005) were used to control the effect of false positives. For global-wise index like
227 network-level gradient values and global eccentricity values, Bonferroni corrections were
228 conducted with a significant level of $p < 0.05$.

229

230 **Results**

231 **Macroscale thalamic gradients in TD and EOS**

232 The principal gradient (G1, 24.6% explained) of the thalamus revealed a L-M axis, and the
233 second gradient (G2, 11.3% explained) described an A-P axis, in line with previous work
234 (25). In **Figure S2**, we also showed the third gradient pattern running in ventral-dorsal
235 direction (9% explained). In EOS patients, we observed expansions at both anchors of the
236 G1; the lateral portions including the ventral lateral and ventral posterior thalamic nuclei, the
237 medial dorsal areas compared to TD controls (**Figure 1A**, TFCE, $p < 0.005/2$). We also
238 observed expansions along the G2 axis including the pulvinar and anterior nuclear groups
239 (**Figure 1B**, TFCE, $p < 0.005/2$). Combing G1 and G2 axes, we computed an eccentricity
240 score using the square root of the Euclidian distance from each thalamic voxel to the center
241 of mass in the two-dimensional gradient space (33). Global eccentricity was assessed for
242 each participant by averaging eccentricity values across all thalamic voxels. We found
243 significantly increased global eccentricity for thalamic voxels in EOS compared to TD ($t =$
244 2.34, $p = 0.02$), indicating a segregation of macroscale thalamic functional organization in
245 patients (**Figure 1C**). Additionally, we explored the spatial pattern of cortical-thalamic
246 gradients, reflecting the organization of cortical-thalamic connectivity in the cortex (**Figure**
247 **S3**). The first cortical-thalamic gradient was similar with previous cortical functional
248 gradients, showing a unimodal-transmodal transition pattern. See **Supplement 2** for further
249 discussions.



251 **Figure 1. Thalamic gradients in typically developing (TD) controls and early-onset**
252 **schizophrenia (EOS) patients.** (A) The group-level primary gradients (G1) for TD and
253 EOS, and their between-group differences. The G1 depicts a transition from lateral to medial
254 portions of the thalamus. Thalamic voxels showing significant G1 score differences were
255 surrounded by black contours [t-test, EOS vs. TD; threshold-free cluster enhancement
256 (TFCE), $p < 0.005$]. The density map represents the distribution of G1 loading for EOS
257 (green) or TD (gray). (B) The group-level secondary gradients (G2) for TD and EOS, and
258 their differences. G2 separates the anterior thalamic portions from the posterior portions.
259 Thalamic voxels with significant G2 differences were surrounded by black contours. The

260 density map represents the G2 loading for EOS (green) or TD (gray). **(C)** Gradient spaces
261 built on the group-level G1 and G2, separately for TD and EOS. Each point represents a
262 thalamic voxel embedded in the gradient space. Voxels are color coded based on their mean
263 eccentricity scores across subjects. Eccentricity score was computed by the square root of
264 the Euclidian distance from each thalamic voxel to the center of mass in the
265 two-dimensional gradient space. Higher eccentricity indicates greater segregation, e.g.,
266 larger dissimilarity of thalamocortical connectivity, in the gradient space. The density plot
267 depicts the distribution of eccentricity scores in EOS and TD.

268 **Functional relevance**

269 Utilizing a whole-brain functional network parcellation (35), thalamocortical functional
270 connectome was distributed into six functional communities (**Figure 2A**). We then averaged
271 gradient loadings within each community for each subject and compared them between the
272 EOS and TD groups (**Figure 2B**, Bonferroni correction, $p < 0.05/6$). EOS patients had
273 significantly higher mean G1 scores in the SMN ($t = 5.37, p < 0.0001$) and DAN ($t = 2.94, p$
274 = 0.004), and lower G1 scores in the VAN ($t = -2.85, p = 0.005$), FPN ($t = -3.69, p =$
275 0.0003), and DMN ($t = -3.03, p = 0.003$) relative to TD controls. For the G2 axis, patients
276 showed decreased gradient scores in the VIS ($t = -4.08, p < 0.0001$), and increased in the
277 DMN ($t = 5.45, p < 0.0001$). In the two-dimensional gradient space showing functional
278 relevance (**Figure 2C**), EOS patients had apparent dissociation of SMN-related thalamic
279 voxels along the G1 axis, and a larger dissociation of VIS- and DMN-related regions along
280 the G2 axis.

281

282 Moreover, rather than dividing the thalamus into functional communities, we also projected
283 thalamic gradients onto the cerebral cortex and evaluated their correspondence with cortical
284 functional networks (**FigureS4**). Both cortical projections of G1 and G2 tended to follow the
285 unimodal-transmodal transition. However, the VAN was located in the unimodal part of G1
286 projection, while the transmodal portion of G2 projection. Nevertheless, the dissociation of
287 SMN-related thalamic voxels in EOS was supported by results of G1 projection, and the
288 VIS-associated dissociation of G2 axis was also replicated.

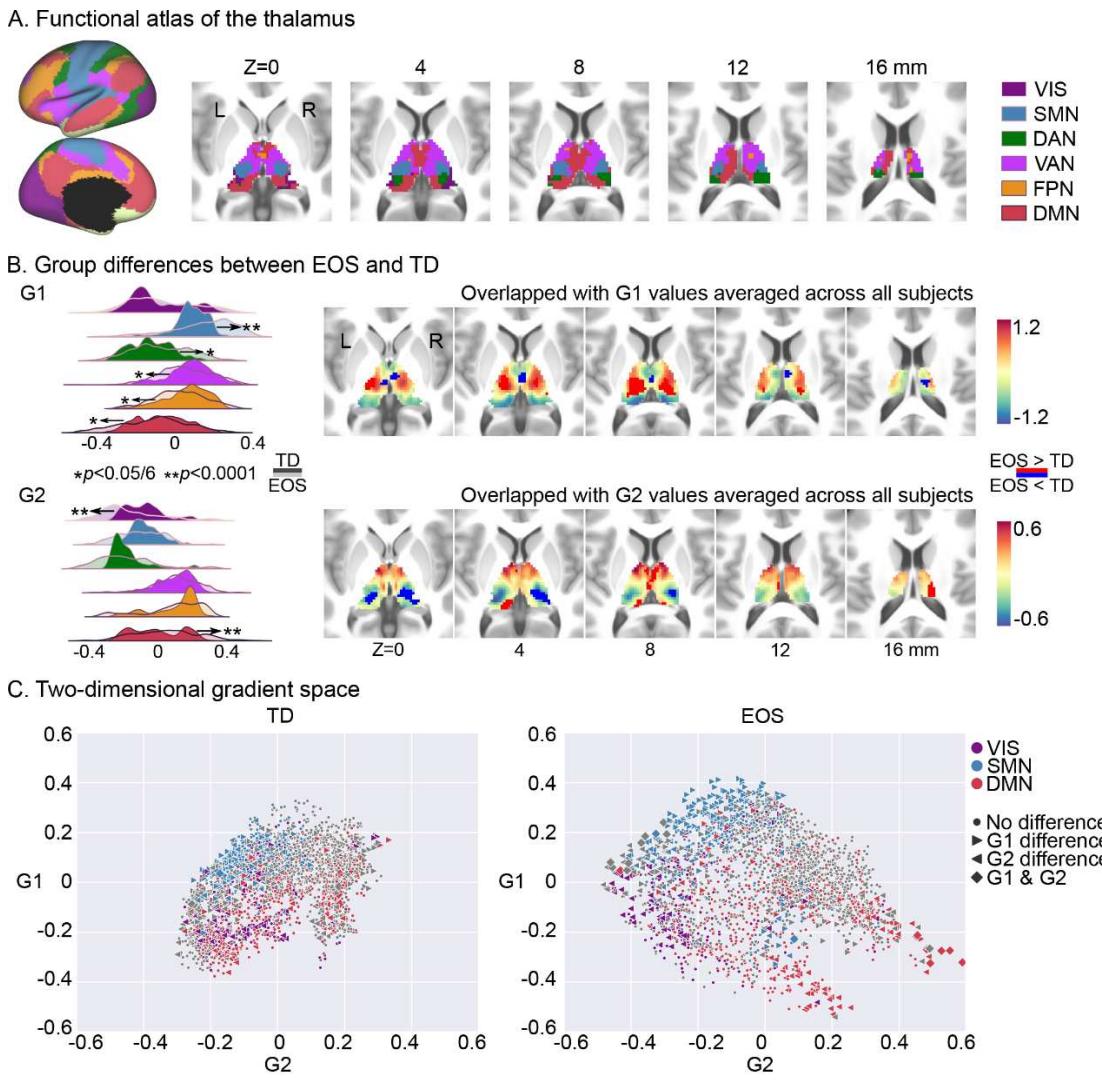


Figure 2. Thalamic gradients distributed into functional communities. (A)

Network-level representations of the thalamocortical connectome. **(B)** G1 and G2 scores within functional communities. Density maps indicate gradient scores (top left: G1, bottom left: G2) within six networks for EOS and TD. Network level differences between EOS and TD were accessed by t-tests, and significant differences are depicted by * (Bonferroni correction, $p < 0.05$) and ** ($p < 0.0001$). Thalamic G1 (top right) and G2 (bottom right) representations across all participants. Voxels showing higher gradient values in EOS are colored red, whereas voxels with lower gradient values are colored blue. **(C)** Gradient space representation of the thalamus together with functional communities for TD (left) and EOS (right). Thalamic voxels are situated based on their G1 (x-axis) and G2 (y-axis) scores, and are colored according to their network assignment (purple-VIS; blue-SMN; red-DMN; grey-all other networks). ► stands for the G1 score differences between EOS and TD; ▲ for the G2 score differences; ◆ for both G1 and G2 score differences; ● for no difference. VIS, the visual network; SMN, the sensorimotor network; DAN, the dorsal attention network; VAN, the ventral attention network; FPN, the frontoparietal network; DMN, the default mode network.

306 **The core-matrix cytoarchitectural basis**

307 Having established macroscale thalamic connectome gradients, we further investigated
308 whether connectome differences between patients and controls were specific to thalamic
309 cytoarchitectural features, namely core vs. matrix cells, motivated by a previous work (8).
310 Thus, mRNA expression levels for CALB1 (core type cells) and PVALB (matrix type cells)
311 were separately assessed at the thalamic voxel level (**Figure S5A**). The differential
312 expression level between CALB1 and PVALB, i.e., CP scores, was used to delineate
313 thalamic core/matrix type cell distribution, where positive/negative CP value indicated with
314 matrix/core cell type, respectively. Eccentricity was used to evaluate the gradient
315 space-embedded position of each thalamic voxel (**Figure S5B**). EOS patients showed
316 increased eccentricity in medial dorsal, ventral lateral, and ventral anterior nucleus compare
317 to TD controls, indicating their increased dispersion from the center of the gradient space
318 (TFCE, $p < 0.005$).

319

320 We then mapped the core-matrix cytoarchitectural features onto the two-dimensional
321 gradient space (**Figure 3A**). Core cells showed mildly increased global eccentricity in EOS
322 patients compared with core cells of TD ($t = 2.38, p = 0.02$), but matrix cells did not ($t = 1.62$,
323 $p = 0.11$). No significant difference was found between global eccentricity of core and
324 matrix populations within EOS ($t = 1.07, p = 0.29$) or TD ($t = 0.36, p = 0.72$). Next,
325 Pearson's correlations were used to quantify group-level associations between the
326 cytoarchitecture and the connectome gradient disturbances, and the variogram methods that
327 control for the spatial autocorrelations were used to determine statistical significance levels

328 (**Figure 3B**). A negative correlation was found between CP values and t values of
329 eccentricity ($r = -0.29$, $p_{vario} = 0.002$). In detail, increased dispersion of macroscale
330 connectome gradient space in patients was particularly related to core cells ($r = 0.31$, $p_{vario} <$
331 0.0001), rather than matrix cells ($r = 0.02$, $p_{vario} = 0.84$).

332

333 **Cognitive relevance**

334 Given the association between CP map and thalamocortical connectome, we further assessed
335 whether gene-connectome coupling contributes to cognitive processing. Thus, we projected
336 the core-matrix cytoarchitecture to the cerebral cortex and then conducted a behavioral
337 decoding using the NeuroSynth database (38). Couplings between functional connectome
338 and gene expression were computed and further down sampled into 400 cortical parcels. As
339 shown in **Figure S6**, core cell populations were mainly associated with unimodal regions
340 that subserve primary sensory and multisensory functions. Matrix cell populations were
341 associated with transmodal cortices characterized by more abstract cognition.

342

343 Compared with TD controls, EOS patients showed increased gene-connectome couplings in
344 the DMN including the inferior parietal lobule, posterior cingulate cortex, inferior prefrontal
345 cortex, temporal areas, and the FPN including the precuneus and inferior parietal sulcus, and
346 temporoparietal junction network (**Figure 3C**, FDR, $p < 0.005$). These abnormal coupling
347 increases were associated with seven cognitive topics including declarative memory,
348 autobiographical memory, working memory, verbal semantics, social cognition, language,
349 and visuospatial (z -statistic > 3.1), indicating its implications in higher-level cognitive

350 processes. Additionally, reduced gene-connectome couplings were observed in the VIS
351 including the extrastriate and inferior extrastriate, and the DAN including postcentral regions
352 and superior parietal lobule, as well as the SMN. These reductions were characterized by
353 low-level visual sensory and motor functions involving visual perception, visual attention,
354 motor, action, eye movements, visuospatial, multisensory processing, and reading (**Figure**
355 **3D**).

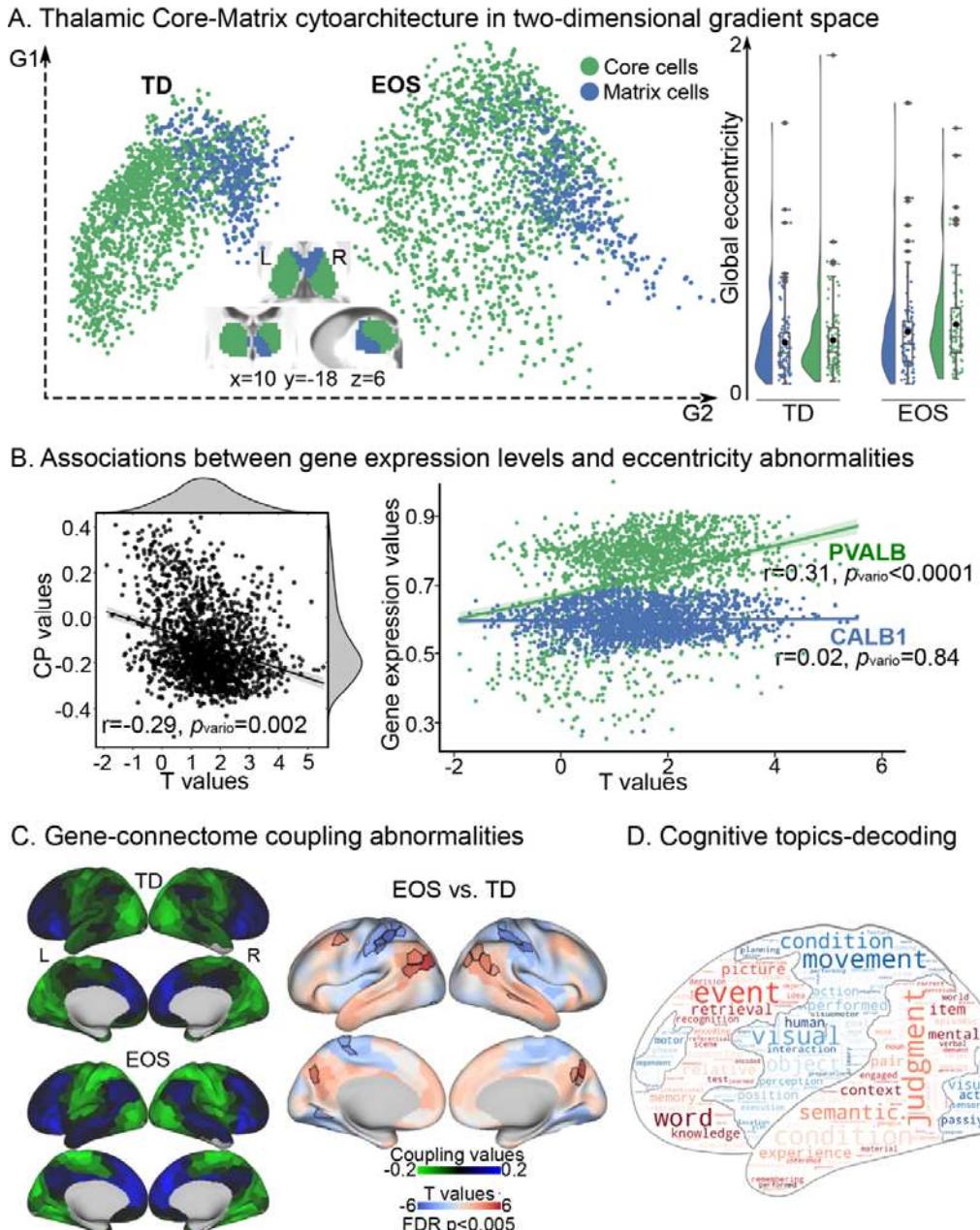


Figure 3. The core-matrix cytoarchitecture of the thalamus. (A) Core-matrix cytoarchitectural features of thalamic voxels projected onto the two-dimensional gradient space for TD and EOS. Rainclouds present group comparisons of global eccentricity in four fashions: global eccentricity values across core cells in EOS vs. global eccentricity values across matrix cells in EOS; core cells in TD vs. matrix cells in TD; core cells in EOS vs. core cells in TD; matrix cells in EOS vs. matrix cells in TD. (B) Associations between eccentricity differences (t values; EOS vs. TD) and differential gene expression levels (CP, left) as well as single gene expression levels (CALB1 and PVALB, right). Correlations were obtained across thalamic voxels (Pearson r values) and their significances were tested using variogram approach (p_{vario} values). (C) Parcel-wise couplings between functional connectome and gene expression for TD and EOS, and their differences (t-test, EOS vs. TD).

367 Cortical parcels with positive coupling values (blue) indicate as preferential associations
368 with matrix thalamic cells, and negative coupling values (green) suggested core cells.
369 Parcels with significantly different coupling patterns in patients are surrounded by black
370 contours [false discovery rate (FDR) corrections, $p < 0.005$]. **(D)** Topic-based behavioral
371 decoding of regions with abnormal couplings in EOS. Cognitive terms in warm color
372 correspond to brain regions showing hyper-couplings in patients relative to controls, and
373 cool color represent hypo-couplings. In the word cloud, the size of a cognitive term is
374 proportional to its loading strength for decoding an input brain mask.

375 **Clinical relevance**

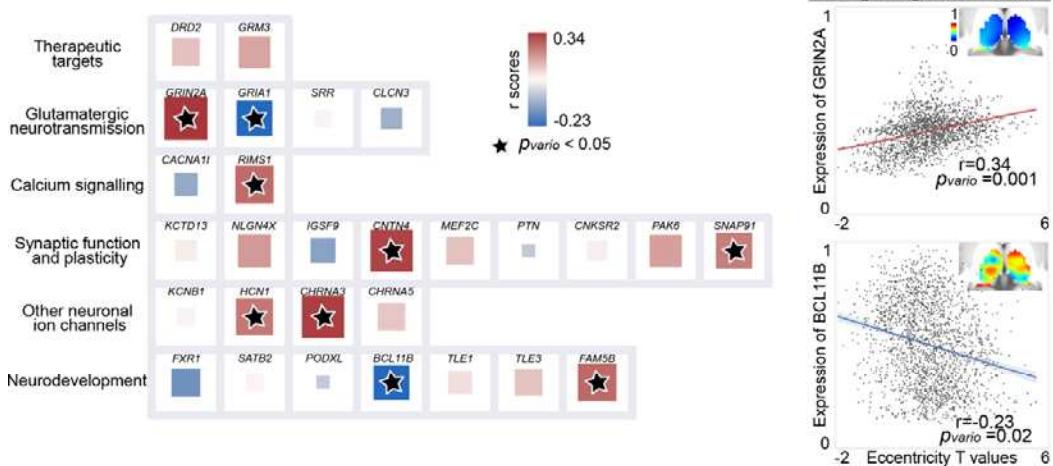
376 To reveal clinical significance of the macroscale functional gradients, we first explored its
377 associations with schizophrenia-related gene expression maps, based on a previous work (39)
378 (**Figure 4A**). In particular, spatial correlations were conducted to quantify their relationships,
379 significance levels of which were evaluated by the variogram methods. The difference map
380 of eccentricity scores between patients and controls was significantly correlated with mRNA
381 expression levels for Glutamatergic neurotransmission-related genes including GRIN2A
382 ($r=0.34$, $p_{vario} =0.001$), GRIA1 ($r=-0.23$, $p_{vario} =0.03$), Calcium signaling-related genes
383 RIMS1 ($r=0.26$, $p_{vario} =0.005$), synaptic function and plasticity-related genes including
384 CNTN4 ($r=0.32$, $p_{vario} =0.001$) and SNAP91 ($r=0.24$, $p_{vario} =0.02$), other neuronal ion
385 channels-related genes including HCN1 ($r=0.25$, $p_{vario} =0.03$) and CHRNA5 ($r=0.33$, p_{vario}
386 $=0.001$), neurodevelopment-related genes including BCL11B ($r=-0.23$, $p_{vario} =0.02$) and
387 FAM5B($r=0.27$, $p_{vario} =0.006$). No significant correlation was observed in the other 19 genes,
388 including therapeutic target-related genes.

389

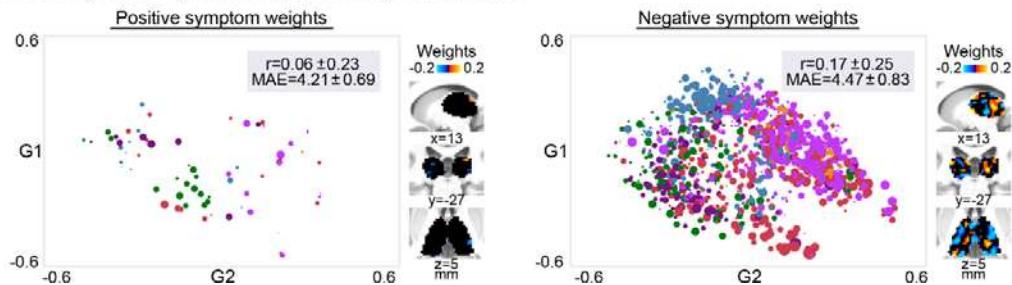
390 Second, we investigated whether thalamic gradients could predict clinical symptoms in EOS.
391 We used eccentricity scores as input features in a linear regression model to estimate PANSS
392 positive and negative scores of patients. The optimal model parameter (L1 ratio) was 0.3 for
393 positive symptoms prediction, and 0.1 for negative symptoms (**Figure S7**). Eccentricity
394 scores of thalamic voxels performed moderately when predicting the severity of negative
395 symptoms ($r = 0.17 \pm 0.25$, MAE = 4.47 ± 0.83), while poor in the prediction of positive
396 symptoms ($r = 0.06 \pm 0.23$, MAE = 4.21 ± 0.69). For positive and negative symptoms,

397 predictive models with median performance were separately reported in **Figure 4B**. In the
398 model predicting negative symptoms, weights were heavier in the thalamic voxels regarding
399 to transmodal networks such as the VAN and DMN.

A. Correlations with 28 schizophrenia-related gene expressions



B. Predicting clinical symptoms by eccentricity maps in EOS



400 **Figure 4. Clinical relevance of thalamic gradients in EOS.** (A) The relationship between
401 the macroscale functional phenotype and 28 schizophrenia-related gene expressions.
402 Interregional correlations (Pearson r coefficients) were employed between eccentricity
403 differences (t values; EOS vs. TD) and gene expression levels, and their significances were
404 evaluated by variogram approach (★ represents p_{vario} value < 0.05). In the left panel, the
405 size of a square is in proportion to absolute value of corresponding Pearson r coefficient,
406 and its color is coded by the sign of r value (red-positive; blue-negative). The highest positive
407 correlation with eccentricity differences was found at a glutamatergic neurotransmission
408 protein-coding gene, i.e., GRIN2A, whose expression pattern is shown in the top corner. The
409 highest negative correlation was found at BCL11B, a neurodevelopment-related gene. (B)
410 The relationship between thalamic functional organization and clinical presentations.
411 Eccentricity values in patients were used to predict PANSS positive and negative scores
412 based on a linear regression model. The model performance was evaluated by comparing
413 observed and predicted clinical scores. This procedure including model learning and testing
414 was repeated 101 times, generating the distributions of Pearson r coefficients and mean
415 absolute error (MAE). The model with median performance was reported here, and absolute
416 value of feature coefficient was used as weight for each thalamic voxel (a point in the

417 two-dimensional gradient space). The size of a point is coded by predictive weight, and is
418 colored according to its network assignment (purple-VIS; blue-SMN; green-DAN;
419 rose-VAN; orange-FPN; red-DMN).

420

421 **Discussion**

422 In the current study we investigated macroscale thalamic functional organization in EOS
423 through dimensionality reduction techniques on thalamocortical functional connectivity. We
424 found both expansions along L-M principal axis and A-P secondary axis of thalamic
425 hierarchies in EOS, indicating connectivity profiles between both anchors showed higher
426 dissimilarities in patients versus controls. Disordered functional hierarchies of the thalamus
427 were related to altered thalamocortical interactions both in unimodal and transmodal
428 networks. To evaluate the cytoarchitectural underpinnings of the macroscale functional
429 organization disturbances in EOS, we compared alterations in functional organization within
430 core- and matrix-cells that derived from an independent transcriptomic atlas. We found that
431 in particular, functional organization of thalamic core cells was altered in EOS. Patients'
432 abnormal coupling patterns between a-priori map of the core-matrix cytoarchitecture and
433 functional connectome characterized a spectrum from perceptual to abstract cognitive
434 functions. Moreover, transcriptomic-informed analyses suggested a close relationship
435 between macroscale functional organization and schizophrenia etiology-related gene
436 expressions in the thalamus. Employing a machine learning strategy, we found that thalamic
437 functional organization was able to predict negative symptoms in EOS. In sum, the current
438 findings provide mechanistic evidence for disrupted thalamocortical system in EOS, and
439 point to alterations in functional networks associated with both perceptual and cognitive
440 functions, suggesting a unitary pathophysiology of heterogeneous symptoms in
441 schizophrenia.

442

443 In line with previous work on thalamic hierarchies (25), the principal functional gradient
444 described continuous transition from the ventral lateral nucleus to the anterior and pulvinar
445 groups, and the second axis delineated gradual transition from the anterior nuclei to the
446 pulvinar. Whereas L-M thalamic axis has been reported to correspond to the distribution of
447 gray matter morphology, i.e., low-to-high intensity of neural mass, and the A-P gradient was
448 strongly related to the intrinsic geometry of the thalamus. These findings suggested an
449 association between functional thalamic hierarchies and its structure. Albeit not directly
450 shown, L-M and A-P axes may reflect functional relevance in different dimensions, i.e., two
451 kinds of transitions across functional networks. Indeed, we found that the L-M principal axis
452 functionally segregated the VIS and SMN, similar to the second gradient of cortical
453 connectome (16). Conversely, the A-P axis of thalamic hierarchies segregated unimodal and
454 transmodal networks, in accord with previous findings (25). Taken together, beyond
455 supporting macroscale thalamic hierarchical framework, the current findings further broaden
456 our knowledge of functional specialization of thalamic L-M and A-P axes.

457
458 Thalamic ventral lateral and ventral posterior nuclei, as one end of the L-M organizational
459 axis, exhibited evident dissociation in EOS. Both nuclei receive neuronal input from the
460 sensory periphery, and project to the motor and somatosensory cortices, respectively (43).
461 The etiology of schizophrenia has been suggested to damage refinement of
462 motor/somatosensory-thalamic connectivity patterns that occurs during brain maturation
463 (15). Indeed, in adolescent patients relative to adult patients, structural abnormalities in the
464 sensorimotor cortex are reported to be particularly salient (44), but may gradually fade out

465 within a longitudinal period of observation (45). Moreover, motor performance has been
466 reported worse in adolescent patients relative to adult patients when accounting for
467 developmental factors (46). In line with this observation, a meta-analysis suggested that
468 motor deficits may precede the onset of schizophrenia and may constitute robust antecedents
469 of this mental disorder (47). In the context, we postulate sensorimotor-related segregation
470 along thalamic L-M hierarchy might underlie premorbid disturbances in motor development,
471 a marker distinct to schizophrenia.

472

473 Compared with TD, EOS patients had increased segregation in two extremes of the A-P axis,
474 i.e., visual-related pulvinar nucleus and default mode-associated anterior nuclear group.
475 Weaker functional connectivity between the VIS and DMN has been previously reported in
476 EOS (48, 49). In fact, given the central role of the thalamus in the development of the
477 cerebral cortex (11), abnormalities of the cerebral cortex in schizophrenia might occur
478 secondary to thalamic pathology (4, 50). Thus, the unimodal-transmodal thalamic hierarchy
479 expansion might further result in disturbed cortical differentiation of unimodal and
480 transmodal regions in EOS. A compression of the unimodal-to-transmodal cortical hierarchy
481 was recently found in chronic adult-onset schizophrenia (51), contrasting with our
482 observation of cortical-thalamic hierarchy expansion. Given age-dependent shifts in the
483 macroscale cortical hierarchy (52), this inconsistency might be due to their disparate stage of
484 the illness and age of onset, or their usage of antipsychotic drugs. Further longitudinal works
485 are needed to chart functional organization abnormalities of the thalamus and the cerebral
486 cortex during the course of schizophrenia. Nevertheless, the current findings embed

487 thalamus into a cortical functional organization linked to differentiation of sensory from
488 abstract cognitive functions, paving the way to comprehensively reveal cognitive defects,
489 another well documented precursor of schizophrenia excepts for motor deficits (47).

490

491 Leveraging our observations of alterations of thalamic functional organization against a
492 proxy map of core/matrix cells based on post-mortem transcriptomic data (8), we observed
493 thalamic core cells to underlie expansive functional hierarchies in EOS, rather than matrix
494 cells. Core and matrix cells are two primary types of thalamic relay neurons which
495 separately exhibit immunoreactivity to the calcium binding proteins Parvalbumin and
496 Calbindin (9). Thalamic nuclei differ in the ratio of core and matrix neurons (53).
497 Specifically, sensory and motor relay nuclei, as well as the pulvinar nuclei and the
498 intralaminar nuclei are chiefly composed of core cells (54). Compared with matrix cells, the
499 larger core cells innervate middle cortical layers in a more area-restricted and
500 topographically-organized fashion (9). Functionally, Parvalbumin-rich core cells have been
501 reported to act as drivers of feed-forward activity, while Calbindin-rich matrix cells fulfil a
502 more modulatory function (7). Together, this may suggest that thalamic hierarchy
503 disturbances in EOS may relate to the “feed-forward” pathway that transmits information
504 from the sensory periphery, not the “feed-back” pathway (55).

505

506 A further inspection of gene-connectome coupling abnormalities in EOS by evaluating the
507 selective connection of the core-matrix cytoarchitecture with the cortex could show that,
508 patients’ core thalamus dys-connected with both unimodal and transmodal cortices. In

509 healthy adults, core regions have preferential connections with unimodal primary regions,
510 and matrix areas with transmodal cortices incorporating the DMN, FPN, VAN and the limbic
511 network (8). Conversely in the current pediatric sample, we observed core-related
512 connections with wide-spread cortical regions including both unimodal and posterior
513 transmodal cortices for TD, whereas EOS had lager similarity with the previously reported
514 adult pattern, i.e., stronger core-related couplings in the VIS, SMN and DAN and weaker
515 couplings in the DMN and FPN. Together, our results imply putatively excessive maturation
516 in the thalamocortical feedforward pathway of schizophrenia. However, future
517 molecular-level work is undoubtedly needed to elaborate on the feedforward pathway
518 alterations in the still developing brain of schizophrenia.

519

520 It has been suggested that schizophrenic brain may not form connections according to gene
521 encoded blueprints which have been phylogenetically determined to be the most efficient
522 (56). In line with this, we observed that impaired thalamic hierarchy in EOS was highly
523 associated with schizophrenia-related gene expressions, especially genes encoding
524 Glutamatergic neurotransmission and neurodevelopmental proteins (39). Our findings reveal
525 a gene-connectome correspondence in the thalamocortical system of EOS, adding new
526 evidence for genetics of schizophrenia. However, there is an obvious shortage in our
527 gene-related analyses. The gene expression levels were assessed from post-mortem brain
528 tissue of adults (28), which might be different from the pediatric human brain. Limited by
529 the lack of pediatric transcriptional atlas, our findings about the association between
530 macroscale connectome topology and gene architecture should be carefully considered.

531

532 Behavioral decoding of the gene-connectome coupling pattern derived a sensory-cognitive
533 architecture, describing functions associated with primary sensory and multisensory
534 processing to working memory, cognitive control and motivation. Along the continuous
535 behavioral spectrum, perception (especially visual sensory), motor, and higher cognition
536 such as memory are particularly affected by schizophrenia. Consistently, thalamic hierarchy
537 abnormalities were sensorimotor-related, as well as visual/ default mode-associated along a
538 second axis. As the “cognitive dysmetria” theory suggested, the multitude and diversity of
539 behavior deficits in schizophrenia might be tied to an impaired fundamental cognitive
540 process mediated by the thalamus (56, 57). This impairment, i.e., cognitive dysmetria
541 referred to a disruption of the fluid and coordinated sequences of both thought and action,
542 leading to a decreased coordination of perception, retention, retrieval, and response
543 functions (58). In particular, our study suggested that abnormal thalamic hierarchy was
544 closely related to negative symptoms of schizophrenia, i.e., a diminution of functions related
545 to motivation and interest. Compared to positive symptoms (such as delusions), negative
546 symptoms are more complex and likely to be the result of systematic disruption. Effective
547 treatment of negative symptoms has long been a clinical challenge for its poor outcomes.
548 The current study provides a thalamic hierarchy framework for heterogeneous behavior
549 deficits related to negative symptoms in schizophrenia, which might denote future therapy of
550 the resistant symptoms.

551

552 In sum, the current study describes thalamic functional organization abnormalities in EOS,

553 which could be related to “feed-forward” core thalamocortical pathway. The macroscale
554 disruptions were related to schizophrenia-related genetic factors. Crucially, it might perturb
555 behaviors involving both low-level perception and high-level cognition, resulting in diverse
556 negative syndromes in schizophrenia.

557

558 **Data and code availability**

559 The data that support our findings are available from the corresponding author upon
560 reasonable request. The estimated spatial maps of mRNA expression levels were
561 downloaded at: <https://www.meduniwien.ac.at/neuroimaging/mRNA.html>. The code for
562 functional gradient analysis was adapted from the MICA lab (<http://mica-mni.github.io>) and
563 is available at <https://github.com/Yun-Shuang/Thalamic-functional-gradient-SZ>. The code
564 for behavioral decoding was adapted from
565 https://github.com/NeuroanatomyAndConnectivity/gradient_analysis/blob/master/05_metaanalysis_neurosynth.ipynb. Statistical analyses were carried
566 out using PALM (<https://fsl.fmrib.ox.ac.uk/fsl/fslwiki/PALM>) and BrainSMASH (<https://brainsmash.readthedocs.io/>). Machine learning analyses were based on scikit-learn package
567 (https://scikit-learn.org/stable/modules/generated/sklearn.linear_model.ElasticNetCV.html).
568 Results were visualized using Connectome Workbench (<https://www.humanconnectome.org/software/connectome-workbench>), and Seaborn (<https://seaborn.pydata.org/>) in
569 combination with ColorBrewer (<https://github.com/scottclowe/cbrewer2>).
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584

585 **Disclosures**

586 The authors declare that they have no conflict of interest.

587

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