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Patients with autism spectrum disorders display reproducible functional connectivity alterations

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Abstract

Despite the high clinical burden little is known about pathophysiology underlying autism spectrum disorder (ASD). Recent resting state functional magnetic resonance imaging (rs-fMRI) studies have found atypical synchronization of brain activity in ASD. However, no consensus has been reached on the nature and clinical relevance of these alterations. Here we addressed these questions in four large ASD cohorts. Using rs-fMRI we identified functional connectivity alterations associated with ASD. We tested for associations of these imaging phenotypes with clinical and demographic factors such as age, sex, medication status and clinical symptom severity. Our results showed reproducible patterns of ASD-associated functional hyper- and hypo-connectivity. Hypo-connectivity was primarily restricted to sensory-motor regions whereas hyper-connectivity hubs were predominately located in prefrontal and parietal cortices. Shifts in cortico-cortical between-network connectivity from outside to within the identified regions were shown to be a key driver of these abnormalities. This reproducible pathophysiological phenotype was partially associated with core ASD symptoms related to communication and daily living skills and was not affected by age, sex or medication status. Whilst the large effect sizes in standardized cohorts are encouraging with respect to potential application as a treatment and for patient stratification, the moderate link to clinical symptoms and the large overlap with healthy controls currently limit the usability of identified alterations as diagnostic or efficacy read-out.

Overline: Autism spectrum disorder

One sentence summary: Functional connectivity alterations are conserved in multiple cohorts of patients with autism spectrum disorder.

Introduction

Autism spectrum disorder (ASD) is a neurodevelopmental disorder characterized by deficits in social interaction and communication, and by the presence of restricted and stereotyped behaviors and sensory anomalies(1). ASD occurs in up to about 1% of general child population(2) and despite its early-childhood onset and lifelong persistence, little is known about the neurobiology underlying its cause and pathogenesis(3). Based on various divergent findings several theories have been proposed to explain ASD at genetic, neuropathology, systemic and behavioral levels(4, 5). On a systemic level, the hypothesis of abnormal brain functional connectivity has been proposed more than a decade ago(6).

One of the prevailing experimental approaches to study brain functional connectivity non-invasively is by resting-state functional magnetic resonance imaging (rs-fMRI) (7). Rs-fMRI reflects spontaneous neuronal activity by detecting slow fluctuations of local oxygen demands (8). Inter-regional temporal synchronization of the intrinsic hemodynamics is then considered as an index of functional connectivity(8). A significant body of evidence now supports the existence of such functional connectivity alterations in ASD(9–15) and correlation of such alterations with autistic traits in neurotypical populations(16). However, the field has still failed to converge on the anatomy, nature, directionality and clinical relevance of the observed alterations. Whereas some reports point to an hypo-connectivity phenotype(17–19), others reported hyper-connectivity(12, 20), and some observed a mix of both phenomena(10, 21). Besides inherent disease heterogeneity, this discordance has been proposed to arise from age or methodological differences between studies(19, 22, 23), or failures to account for medication effects in the analyses(24). Yet other works highlight the importance of spatial variability in connectivity as a key source of heterogeneity in ASD(21, 25). A common caveat in interpretation of these studies is the lack of replication of respective findings. A consensus on the nature and clinical relevance of the functional connectivity changes in ASD therefore still needs to be reached. These uncertainties limit the usability of functional connectivity as a potential diagnostic or treatment biomarker for ASD.

Here we aimed to address these inconsistent findings by performing a large-scale evaluation and characterization of functional connectivity alterations in ASD in four independent cohorts. More specifically, we used the EU-AIMS LEAP (European Autism Interventions - A Multicentre Study for Developing New Medications – EU-AIMS, Longitudinal European Autism Project – LEAP) dataset for exploration and the other three cohorts (ABIDE (Autism Brain Imaging Data Exchange) I, ABIDE II and InFoR (Inserm, la Fondation FondaMental et Roche)) for replication. We further evaluated the link between these alterations and age, medication status and clinical symptoms.

Results

Clinical and demographic characteristics are comparable between patients with ASD and typically developing volunteers (TD)

Rs-fMRI data from four ASD cohorts (EU-AIMS LEAP, ABIDE I, ABIDE II and InFoR) alongside with matched TD were included in the study (Table 1). The multi-center EU-AIMS LEAP and the single center InFoR are prospective cohorts providing standardized rs-fMRI data alongside with various clinical parameters. ABIDE I and II represent large-scale retrospectively aggregated datasets provided by international brain imaging centers. Exclusion of individuals with low IQ and removal of poor image quality data resulted in exclusion of 35, 39, 33 and 14% of available EU-AIMS LEAP, ABIDE I, ABIDE II and InFoR data, respectively (Table 1 for final number of subjects (N)). For all cohorts, mean age did not differ between individuals with ASD and TD and was comparable between EU-AIMS LEAP and ABIDE I (Table 1). ABIDE II cohort was significantly younger (all $p < .001$) and the InFoR only comprised adult individuals with ASD. A significantly higher fraction of males was observed in the ABIDE I and II as compared to EU-AIMS LEAP (ABIDE I: $p < 0.001$, ABIDE II: $p = 0.005$) and InFoR (for ABIDE I only: $p = 0.042$). Individuals with ASD in all cohorts had average IQ above 100 although lower than that for TD in all cohorts except InFoR (EU-AIMS LEAP: $p = 0.033$, ABIDE I and II: $p < 0.001$, InFoR: $p = 0.392$, Table 1).

Reproducible degree centrality alterations are observed in individuals with ASD

To test for functional connectivity differences between individuals with ASD and TD, we used the EU-AIMS LEAP dataset for exploratory analyses and the other three cohorts for replication. Degree centrality (DC) was chosen as an unbiased count-based functional connectivity measure that assigns to each voxel the sum of all correlation coefficients between the time series of that voxel and all other voxels in the brain exceeding a prespecified threshold. Group comparisons in the EU-AIMS cohort revealed significant DC increases in individuals with ASD in two clusters comprising bilateral parietal ($p = 0.008$), prefrontal and anterior and posterior cingulate regions ($p = 0.001$) (Fig 1A and Table 2). Significantly decreased DC was observed in a cluster covering bilateral primary sensory motor cortices and right temporal regions, insula, amygdala and hippocampus ($p < 0.001$) (Fig 1A). These results remained qualitatively and quantitatively similar to the initial findings after controlling for between-subject differences in motion (fig. S1, table S1).

DC extracted from ABIDE I, ABIDE II and InFoR based on the EU-AIMS LEAP DC increase mask revealed significant increases in individuals with ASD in all three replication cohorts in the same areas showing increased connectivity in the exploratory cohort (ABIDE I: $p = 0.011$, ABIDE II: $p = 0.01$, InFoR: $p < 0.001$, Fig 1B). Conversely, the DC decreases were successfully replicated in two out of three cohorts (ABIDE I: $p = 0.001$, InFoR: $p = 0.002$) (Fig 1B). We further evaluated the spatial similarity of

the unthresholded DC alteration patterns by computing correlation coefficients between voxel-wise *t*-maps obtained for each cohort in the comparisons of ASD and TD (Fig 1C). These analyses revealed significant correlations of the ASD profiles across all four cohorts (all $p < 0.001$, Fig 1D).

Degree centrality alterations in ASD are driven by shifts in connectivity

Having identified replicable DC alterations in individuals with ASD we aimed to explore the nature of these connectivity differences. To this end, we compared the whole-brain histograms of correlation coefficients observed in individuals with ASD to TD revealing a nearly identical distribution of correlation strength (Fig 2A). We next computed effect sizes for differentiation between individuals with ASD and TD for four types of metrics comparing means, variances, proportions and shifts in connectivity within, outside and from within to outside the identified DC alteration clusters (fig. S2). These analyses revealed that shifts in correlations from outside to inside the DC increase mask showed the largest effect size (Fig 2B). This large effect reflects that in ASD voxels falling outside the DC increase mask show reduced connectivity to one another and an increased connectivity to voxels within the mask. This hyper-connectivity index was closely negatively correlated with the initial DC finding ($r = -0.72$; $p < 0.001$; $N = 394$) (fig. S3). Additionally, reduced overall connectivity and reduced proportion of connected voxels were observed within the DC increase mask and an increased variance of connections outside the respective mask (Fig 2B).

For the DC decrease findings, three indices revealed similar effect size as the initial DC finding: reduced mean connectivity and number of connected voxels within the DC decrease mask, and shift in connectivity from outside to inside the DC decrease mask (Fig 2B). For the latter, the proportion of voxels connected to each other outside the DC decrease mask increased and the proportion of voxels connected from outside to inside decreased. Strong correlations were observed between all these indices and the initial DC decrease findings (mean connectivity: $r = 0.81$; $p < 0.001$, proportion of connected voxels: $r = 0.84$; $p < 0.001$, shift in connectivity: $r = -0.85$; $p < 0.001$; $N = 394$) (fig. S3). As the shift in connectivity showed the strongest correlation with the initial DC findings, this hypo-connectivity index was selected for further testing of associations with clinical symptoms.

The further exploration in the EU-AIMS LEAP dataset of the choice of correlational threshold for DC computation on the observed differentiation between ASD and TD revealed a maximum differentiation for the hyper- and hypo-connectivity index at $r = 0.15$ and $r = 0.4$, respectively (Fig 2C). For both indices the effect size for differentiation between ASD and TD obtained at the initially chosen DC threshold ($r = 0.25$) recommended by the literature (26) was very close to the maximum observed with those optimized thresholds. These results support the use of the recommended threshold for DC-based analyses of the ASD population.

To better understand the spatial specificity of the observed shifts from outside to inside the respective DC masks for both hyper- and hypo-connectivity indices we have additionally recomputed both indices by splitting the outside regions into cortical, subcortical or cerebellar regions of interest. Only the cortico-cortical shifts in connectivity revealed significant ($p < 0.001$) and numerically larger effect sizes for differentiation between ASD and TD than the initial whole-brain hyper- and hypo-connectivity indices (Table 3). No significant differences were observed for both indices when using subcortical and cerebellar outside regions, with effect sizes close to zero (Table 3).

Connectivity differences are linked to demographic and clinical factors

We next evaluated if and how these hyper- and hypo-connectivity indices relate to clinical characteristics, age, sex, medication and psychiatric comorbidity status in individuals with ASD. First, we used t-tests to evaluate if the hyper- and hypo-connectivity alterations persisted when splitting the cohorts into children, adolescents and adults. As InFoR only comprised adults, these analyses were restricted to the other cohorts. For the hyper-connectivity index, significant differences and similar effect sizes were observed for all age groups in the EU-AIMS LEAP cohort (all $p < 0.001$, Fig 3A). In other cohorts significant differences between individuals with ASD and TD were observed for adolescents (ABIDE I only, $p = 0.028$) and adults (ABIDE I: $p = 0.014$, ABIDE II: $p = 0.015$) but not for children (Fig 3A). The hypo-connectivity index showed consistently larger effect sizes in the adolescent and adult populations as compared to children in all three cohorts (Fig 3B). In formal testing for age category by diagnosis interactions on the hyper- and hypo-connectivity indices using analysis of variance (ANOVA) no significant interactions were observed (table S2).

In ANOVA testing for the effects of sex on the observed hyper- and hypo-connectivity indices no significant sex-by-diagnosis group interactions were observed for any of the cohorts (all $p > 0.1$) (table S2). A significant effect of psychotropic medication on functional connectivity was only found for DC increases in the EU-AIMS LEAP cohort with individuals with ASD on medication being closer to the TD population (table S3). Psychiatric comorbidity status was only available for a subsample of the ABIDE II cohort. There was no significant difference between ASD with and without comorbidity diagnoses with respect to DC increases ($t(187) = 0.1; p = 0.903$), decreases ($t(187) = -1.0; p = 0.335$) or derived hyper- ($t(187) = 0.6; p = 0.582$) and hypo-connectivity ($t(187) = 1.5; p = 0.147$) indices (fig. S4). Psychiatric comorbidity and medication status were only weakly associated in that sample (Kendall's $\tau = 0.18; p = 0.014$) (table S4).

Following a similar exploration and replication strategy as for group comparisons, twelve general linear models were computed in the EU-AIMS LEAP dataset to evaluate if the obtained hyper- and hypo-connectivity indices are linked to clinical severity (table S5). These analyses revealed overall two significant relationships between the hyper-connectivity index and Autism Diagnostic Interview (ADI) communication subscale ($p = 0.026$) and Vineland Adaptive Behavior Scale (VABS, (27)) daily living

skills standard score ($p=0.024$) (Table 4 and table S5). Stronger connectivity alterations were associated with stronger clinical impairment (Fig 3C). In replication analyses, both relationships found in EU-AIMS LEAP were also significant in the ABIDE I cohort (ADI communication: $p=0.004$, VABS daily living skills: $p=0.048$) but not in the substantially younger ABIDE II subset (Table 4, table S6). In an additional exploratory analysis of the EU-AIMS LEAP cohort we examined the correlational structure between the identified ADI communication subscale and the VABS subscales. This analysis was performed to better understand why VABS daily living skills (but not the corresponding communication subdomain) were associated with the hyper-connectivity index. ADI communication subscale was numerically more strongly associated with VABS daily living skills ($r=-.29$; $p=0.006$) than with the VABS communication ($r=-.24$; $p=0.029$) subdomain (Fig 3D).

Understanding relationship to underlying structure

Lastly, we aimed to evaluate if the observed DC alterations or the extracted ASD hyper- and hypo-connectivity indices were related to underlying grey matter structure or to previously reported white matter alterations (whole cerebral white matter and corpus callosum subregions) (28). For this, we first compared the respective volumetric information between ASD and TD and then correlated the respective structural information with the identified functional connectivity alterations. Overall white ($p=0.017$) and splenium of corpus callosum matter ($p=0.014$) were reduced in ASD displaying yet small effect sizes (table S7). Grey matter volume in the selected regions and genu and body of corpus callosum were not significantly different between ASD and TD. Only white matter in genu of corpus callosum but none of the other structural information was significantly correlated with the DC increase values ($r=-0.1$; $p=0.043$) and the respective ASD hyper-connectivity index ($r=0.1$; $p=0.04$) with lower genu volume being associated with stronger rs-fMRI alterations (table S8).

Discussion

Here we provide evidence of reproducible functional hyper- and hypo-connectivity alterations in individuals with ASD as demonstrated across several large single- and multi-center cohorts. We further show that these alterations persisted when controlling for motion and medication effects and were linked to clinical symptoms as captured by several clinical and behavioral scales.

We reconcile previous divergent literature findings and report evidence of both functional hypo- and hyper-connectivity in individuals with ASD, with spatially-varying signatures that are consistent with some of the recent single cohort evidence (10, 12, 19, 20). The identified hypo- and hyper-connectivity patterns that are based on a positive correlation threshold show spatial similarity across all cohorts with the hypo-connectivity being primarily restricted to sensory-motor regions and hyper-connectivity hubs being predominately located in prefrontal and parietal cortices. Both anatomical networks are consistent

with some previous work reporting connectivity alterations in respective networks(10, 21, 29). Moreover, our results suggest that the identified alterations are primarily driven by cortico-cortical connectivity shifts with little to no cerebellar or subcortical contribution. These findings contrast with some of the previous literature suggesting alterations in cerebro-cerebellar and cerebro-subcortical connectivity as a key alteration in ASD(30, 31). Substantial differences in preprocessing and methodology may account for some of these discrepancies. Consistent with some previous reports, we find the overall global connectivity to be preserved in ASD(21).

Anatomically, the DC increase regions closely correspond to the well-established central executive network(32–34). This network has been closely implicated in high-level cognitive functions such as planning, decision making, and the control of attention and working memory(33). All of these functions have been consistently shown to be affected in ASD(35–37). Moreover, we demonstrate that the DC increases originate specifically from increased engagement of this network by regions outside the network. Similarly, DC decreases are mainly driven by disengagement of the sensory-motor network from other regions. However, we also observe a similar magnitude of reduced connectivity within the sensory-motor network suggesting that both processes contribute to DC decreases. Biologically, these connectivity shifts may support the idea that individuals with ASD are unable to engage or disengage specific networks to the extent of TD, and may therefore underlie deficits in mental switching and cognitive flexibility(35, 36).

Estimates of effect size range from small to very large across cohorts. The ABIDE I and II datasets are retrospectively established cohorts with large variation in diagnostic and inclusion criteria, sequence quality, scanning duration and other parameters such as resting state acquisition instructions. This variability may explain the observed low effect sizes. In contrast, the standardized multi-center EU-AIMS LEAP and the single center InFoR cohort show a large effect size for the hyper-connectivity and a moderate to large effect size for the hypo-connectivity alterations, suggesting that these measures may indeed have promise as stratification or treatment biomarkers for ASD. Whilst these results provide a valuable first step demonstrating the existence of a consistent functional connectivity phenotype associated with ASD they also at the current stage point to the limited usability of the respective alterations as a diagnostic biomarker considering the large overlap between ASD and TD. As a key focus of our study primarily in identification and replication of the respective alteration, a large amount of possible optimizations remains to be explored including deployment of improved sequences but also evaluation of other than default processing parameters and further spatial dissection of respective alterations. All of these aspects may result in further increases in effect size potentially lifting it to a clinically relevant magnitude.

The above results contribute to a growing consensus on the specific networks characterized by hyper- vs. hypo-connectivity profiles in ASD, but also extend this work to demonstrate a key driver of ASD's

connectivity profile: the larger proportion of frontoparietal voxels within our "DC Increase" mask showing suprathreshold connectivity with voxels outside this mask, and the larger proportion of somatosensory voxels within our "DC Decrease" mask showing subthreshold connectivity with voxels outside this mask. Our somatosensory findings especially concord with emerging ICA-based analyses of the EU-AIMS LEAP and other cohorts (38), in which connectivity alterations in ASD are generally most pronounced when considering connectivity between networks. This concordance, despite different preprocessing and analysis methodologies, suggests ample robustness to the results reported here, and motivate further characterization of the etiology, development and clinical symptoms associated with these phenotypes in ASD. At the same time, we note that the most consistent ICA-based findings of between-network differences involve cerebellum, which was only inconsistently acquired in the broader range of cohorts analyzed here. Clearly, the role of cerebellar function in the broader ASD phenotype, and its relation to frontoparietal connectivity in particular, may be an important direction for future work(39). We do not find consistent age-by-diagnosis interactions on the identified hyper- and hypo-connectivity indices. Nonetheless, numerically all cohorts show smaller effect size for the hypo-connectivity index in children suggesting a potential increase of these alterations over age. As this index is not associated with any clinical scales, it may serve a more compensatory role that increases with age. In line with this hypothesis, previous studies reported age-related improvement among older individuals with ASD in terms of cognitive flexibility(40). The absence of the functional connectivity alterations in two out of the three children cohorts may also point to a potentially secondary role of the respective alterations, for example emerging due to prolonged social interaction. However, also other possible explanations such as differential neurodevelopmental trajectories or long term exposure to medication are also possible. Prospective longitudinal studies are therefore essential to further dissect the role of the identified functional connectivity alterations in ASD.

In line with the idea of differential neurodevelopmental trajectories, a previous large diffusion tensor imaging study in toddlers with ASD demonstrated alterations in axon pathways connecting primarily prefrontal and other brain regions(41) and these findings are also supported by an earlier postmortem study in ASD reporting altered white matter composition below prefrontal regions (42). The respective imaging study did not find any alterations in white matter connections in parietal regions. In line with that, the visualization of voxel-wise degree centrality provided in our study illustrates a more dominant prefrontal pattern in both ABIDE cohorts whilst the parietal effects are primarily observed in both EU-AIMS and the adult only InFoR cohort. This discrepancy may point to the rather later emergence of the parietal alterations and also to some extent explain the low effect size in the younger ABIDE II cohort when extracting the signal from the whole fronto-parietal network identified in the EU-AIMS LEAP dataset. Supportively, the only associations between structural and functional information identified in our study was the correlation of fronto-parietal DC and the respective hyper-connectivity index with white matter in genu of corpus callosum connecting specifically prefrontal regions to each other. The

directionality of these correlations points to larger alterations in both resting state measures being associated with lower genu volume. However, the rather weak nature of this association combined with the much lower effect size for the white matter alterations also point to a largely independent contribution of both endophenotypes with stronger relevance of the identified functional connectivity alterations going beyond changes in underlying structure. Our results also support the idea of an altered topology in ASD with a spatially distinct pattern of between network connections with both hyper- and hypo-connectivity being present(43).

When evaluating the link between imaging and clinical features, we find that the extent of hyper-connectivity abnormalities is correlated with autism symptoms in two out of three cohorts. Hyper-connectivity was associated with ADI communication functioning as well as VABS daily living skills but not the respective VABS communication subscale. Greater connectivity alterations are thereby associated with greater severity of respective symptoms. Additional exploratory analyses illustrated the rather weak relationship between deficits captured by the corresponding ADI and VABS subscales suggesting that VABS daily living skills are more closely correlated with ADI communication scores. A potential reason for this closer link to VABS daily living skills may be due to ADI collecting information about ASD-specific communication deficits historically observed in daily living rather than differentiating these subdomains in terms of current functioning, as assessed by the VABS. These findings further support the idea that these alterations may reflect pathological processes related to skills needed for development of personal independence and social communication abilities. We also failed to replicate this association in the on average six to ten years younger ABIDE II subpopulation. Considering that ABIDE II children diagnosed with ASD do not show any detectable connectivity abnormalities, the lack of significant correlation with clinical symptoms in this cohort is not surprising. Overall, further research is needed to understand the specific reasons for these discrepant findings. Potential confounding effects of medication on functional connectivity have been raised for previous ASD studies(24). Here, we did not find consistent effects of psychotropic medication on the functional connectivity measures with the only effect pointing to reduced functional connectivity alterations with psychotropic medication use. On one side, the rather low effect size of those associations limits their immediate usability as potential surrogate efficacy read-outs. On the other side, the large effect size in the standardized cohorts, the link to the clinical symptoms and, if replicated, the preliminary evidence in one of the cohorts that ASD medication may move the networks closer to TD warrant further exploration of these alterations as a potentially more sensitive and earlier pharmacodynamic read-out for treatment evaluation. Similarly, in the only cohort (ABIDE II) where psychiatric comorbidity status was consistently provided, we did not find any evidence of its impact on the observed functional connectivity alterations. Although these results are consistent with the notion that the observed imaging alterations are specific to ASD, further analysis of broader cohorts are needed to fully address the question of specificity.

For statistical and data quality reasons we have excluded for all analyses data of individuals with ASD with low IQ or where imaging quality including excessive motion did not meet prespecified criteria. More specifically, motion is critically discussed in the literature as a potential contributor or even driver of functional connectivity alterations observed in previous studies(44, 45). Similarly, low IQ is often associated with specific factors such as genetic etiologies or prenatal events including exposure to specific drugs(46–49). Those factors are likely to increase variability of the overall population potentially reducing the sensitivity to detect specific between group differences. Additionally, the low prevalence of low IQ subjects was not equally distributed across cohorts and between ASD and TD in the EU-AIMS dataset inducing an additional statistical bias which we aimed to avoid. It was therefore critical for interpretation of our findings to minimize any possible confounding effects associated with both factors by excluding these subjects. Nonetheless, whilst these exclusion criteria are important to enable more clear interpretation of the outcomes, they also limit the generalizability of our findings to the excluded patient populations. In addition, also the role of other potential confounding factors such as presence of specific neurological conditions or congenital malformations may have contributed to observed differences and needs to be explored in future studies collecting detailed information on the respective aspects.

Lastly, we did not use the INFOR dataset for replication of the clinical associations; considering the rather weak nature of the observed associations in the EU-AIMS LEAP cohort, the InFoR dataset did not provide sufficient power (only 12% power based on power-analysis using G-Power) to detect the respective associations.

In conclusion, we provide evidence of reproducible hyper- and hypo-connectivity alterations in individuals with ASD. These alterations are mainly driven by shifts in connectivity from within to outside the respective networks and are partially associated with core clinical deficits observed in ASD.

Materials and methods

Study design

The primary objective of the study was to identify and independently replicate rs-fMRI-based functional connectivity alterations associated with ASD as compared to TD. Further, we aimed to understand the specific nature of these alterations and if they are associated with methodological, demographic or clinical factors known to affect rs-fMRI derived measures. Lastly, we explored if these functional connectivity alterations are linked to core clinical symptoms observed in ASD. To enable an accurate identification and replication of functional connectivity alterations in ASD four cross-sectional datasets were included into the study (Table 1, Supplementary materials and methods, table S9-11): EU-AIMS LEAP (www.eu-aims.com), ABIDE I, ABIDE II and InFoR (10, 50–52). All cohorts acquired rs-fMRI

data (Supplementary materials and methods, table S12-15) in individuals with ASD and TD along with other cohort-specific measures. Only individuals with ASD with average range IQ (>70) were included for further analyses to reduce variability associated with low functioning ASD (53). A further statistical reason for exclusion of this population was in the strong imbalance in the EU-AIMS dataset with more than twice as many individuals with ASD meeting this criterion as compared to TD. Additionally, only few subjects meeting this criterion were available in all replication cohorts. In brief, EU-AIMS LEAP cohort is a large well characterized multicenter cohort of individuals with ASD and TD. EU-AIMS LEAP recruitment was performed using existing local databases, clinic contacts, and local and national support groups. TD were recruited via mainstream schools, flyers, and existing databases. A BIDE I and II provide large, retrospectively aggregated multi-center data with varying diagnostic and recruitment criteria, rs-fMRI sequences and clinic scales. InFoR is a smaller single center cohort with standardized imaging assessments. As the only large cohort with standardized imaging assessments and inclusion/exclusion criteria, the EU-AIMS LEAP dataset was selected for exploratory analyses. The three other cohorts were used for replication of the identified functional connectivity alterations. All studies were approved by local ethics committees. Further details on rs-fMRI data collection and pre-processing are provided in Supplementary materials and methods.

Statistical analyses

Identifying functional connectivity alterations

Given the divergent prior findings on functional connectivity alterations in ASD, we followed an unbiased exploratory approach by computing a simple connectivity metric (degree centrality – DC, (56)) to compare individuals with ASD and TD. DC is a count-based measure that assigns to each voxel the sum of all correlation coefficients between the time series of that voxel and all other voxels in the brain exceeding a prespecified threshold ($r>0.25$) (26). Only positive correlations were included to avoid spurious negative correlations that may arise from global signal regression (58). This threshold was recommended in previous studies using DC to eliminate counting voxels that had low temporal correlation attributable to signal noise (26). All further group comparison and correlational analyses were controlled for site, sex, age and IQ effects. As compared to many other connectivity approaches proposed in the literature DC does not require specific assumptions about the location of the signal and can be computed on a voxel-wise basis. Though dimensionality reduction techniques such as independent component analysis are also often applied in the literature to evaluate functional connectivity, they are less optimal for the key focus of our study which was in replication across different cohorts. Solutions provided by such algorithms are often highly susceptible to differences across study populations such as sample size, demographics and scanner and sequence differences. Also there is no consistent approach for matching the identified components across cohorts. The DC approach was therefore chosen for the replication focus of this study.

The resulting DC maps for EU-AIMS were entered into a voxel-wise general linear model (GLM) comparing individuals with ASD and TD. A liberal voxel-wise uncorrected threshold of $p < 0.05$ combined with an exact permutation based (1000 permutations) family-wise error corrected cluster threshold of $p < 0.05$ was used in those analyses to test for significance (59). All analyses were controlled for site, sex, age and IQ effects. Identical GLM designs were then created for ABIDE I, ABIDE II and InFoR data. Weighted mean DC signals (first eigenvariate) were extracted for the three replication cohorts based on EU-AIMS LEAP findings. Effect sizes (Cohen's d) and independent sample t -tests were computed comparing DC values between individuals with ASD and TD ($p < .05$). We then tested if the whole-brain DC alteration patterns observed in ASD are similar across the cohorts by computing correlations between the un-thresholded t -maps (degrees of freedom for p -value computation are based on number of resolution elements in the images provided by SPM12) (Figure 1C, D). As residual motion effects have been repeatedly criticized as potential sources of observed functional connectivity differences between ASD and TD (54), we have computed mean translational and rotational motion and frame-wise displacement and compared it between ASD and TD using t -tests (table S1). As the mean frame-wise displacement was indeed significantly higher in ASD as compared to TD, we have also recomputed all analyses by additionally controlling for the effects of the motion parameters in the between group comparison. For this we added mean translational and rotational motion and mean translational and rotational frame-wise displacement for each subject as additional covariates of no interest and recomputed the above analyses.

Understanding the nature of functional connectivity alterations

Whilst DC provides evidence of altered functional connectivity, it does not allow conclusions on the nature of those alterations. For example, DC changes could arise from alterations in mean strength and/or variance of the underlying connectivity, or shifts in connectivity from within to outside of the respective networks. To better understand the observed functional connectivity alterations, we first computed subject-specific pair-wise correlations between time courses of all voxels in the brain. Based on these correlations we computed the following four types of connectivity indices within and outside regions showing DC alterations as defined in figure S2: 1) mean connectivity of all voxels (Fisher's z -transformed), 2) variance of connectivity of all voxel-wise correlations (Fisher's z -transformed) 3) proportion of connected voxels, and 4) shifts in connectivity from inside to outside the respective regions.

Effect sizes and t -tests comparing individuals with ASD and TD were then computed for each of the above indices. To test if these indices indeed reflect the initial DC changes we computed correlations between both in the EU-AIMS LEAP dataset. Indices showing strongest correlation with initial DC increase and decrease findings and a similar or higher effect size were selected for further evaluations. These two indices providing a more refined view on functional connectivity alterations were extracted

for the three replication cohorts and used for further analyses. Both indices are further referred to as hyper- and hypo-connectivity indices.

We further examined in how far the initial choice of the correlational threshold used for DC computation affected the effect sizes observed with both indices. For this, we systematically varied in the EU-AIMS LEAP dataset the correlational threshold from -0.25 to 0.7 computing the effect sizes for differentiation between ASD and TD.

Lastly, as previous studies reported potential differences in between network connectivity alterations in cortical, subcortical and cerebellar regions we have additionally computed the effect sizes and t-tests comparing ASD and TD for the above two indices by re-computing them using for outside regions either cortical, subcortical or cerebellar masks as described in Supplementary methods (30, 31).

Evaluation of cortical, subcortical and cerebellar hyper- and hypo-connectivity indices

To re-compute the identified hyper- and hypo-connectivity indices splitting the outside regions into cortical, subcortical and cerebellar subregions we masked the outside regions (outside the respective DC increase and decrease masks) using the Automated Anatomical Labeling – AAL atlas. Thereby, an inclusive mask was computed between the outside DC masks and a cortical mask pooling all cortical AAL regions. Similarly, an inclusive mask was computed between the DC outside masks and subcortical regions pooling caudate, putamen and thalamic AAL regions. For cerebellum mask, all AAL cerebellum and vermis regions were used. The hyper- and hypo-connectivity indices as defined in equation $\pi 5$ in Figure 2 of the main manuscript were recomputed for the obtained cortical, subcortical and cerebellar masks.

Understanding the link between imaging findings and clinical and demographic factors

Having identified the hyper- and hypo-connectivity indices as replicable ASD biomarkers we next investigated their relationship to clinical symptoms, medication status, age and sex. To test for differential effects of age, we split the cohorts into three age categories: children (below 12), adolescents (age between 12 and 18) and adults (age above 18). Hyper- and hypo-connectivity indices between ASD patient and respective TD populations were compared by computing effect sizes and t-tests (controlling for overall age effects, sex, site and IQ). As based on the outcomes of the overall cohort we had clear directionality hypotheses a one-sided $p < 0.05$ was applied. Additionally, we formally tested for age-by-diagnosis interactions (three age categories) using ANOVAs controlling for sex, site and IQ. Similarly, to test for the effects of sex we computed ANOVAs testing for sex-by-diagnosis interactions controlling for age, site and IQ. We further evaluated for each cohort if psychotropic medication interferes with the observed functional connectivity alterations. Because detailed medication information on specific drugs, dose or duration was not available we restricted the analyses to t-tests

comparing DC and hyper- and hypo-connectivity indices between patients on and off psychotropic medication. Additionally, we aimed to evaluate whether psychiatric comorbidity and concomitant medication status might also contribute to the observed functional connectivity alterations. For a subcohort of ABIDE II the psychiatric comorbidity status was also available (N=189, 72 with comorbidities). For this subcohort, we computed t-tests comparing the obtained DC and respective hyper- and hypo-connectivity indices (adjusted for age, sex, site and IQ) and explored the relationship between medication and comorbidity status using Kendall's tau.

To test for potential relationships between hyper- and hypo-connectivity indices and clinical scales, we followed a similar exploration and replication strategy as for the group comparisons. Because of few overlaps regarding clinical scales and a much smaller number of patients than in the exploration sample InFoR was not used for these analyses. Separate GLMs (as implemented in SPSS) were computed using the hyper- and hypo-connectivity indices to predict clinical severity in the EU-AIMS LEAP individuals with ASD (table S6). Significant ($p < 0.05$, two-sided) relationships identified in the EU-AIMS cohort were selected for replication in ABIDE I and II using identical GLMs. Additionally, we explored the interrelationship of the ADI and VABS subscales showing a significant association with the hyper- and hypo-connectivity indices in the EU-AIMS LEAP dataset by computing Pearson correlations among the respective measures.

Understanding relationship to underlying structure

Lastly, we aimed to evaluate if the observed functional connectivity alterations are related to potential underlying grey matter structural changes, as between-group structural differences may confound the effects seen on functional connectivity (60). To address this question, two-types of analyses were performed using eigenvariates of grey matter volumes extracted using both DC masks and adjusted for age, sex, site, IQ and total intracranial volume: 1) group comparisons of ASD and TD using two-sample t-tests, and 2) Pearson correlations between grey matter volumes and respective DC and hyper- and hypo-connectivity indices. Additionally, as previous studies have reported reduced white matter and specifically corpus callosum volumes in ASD, we repeated both types of analyses using white matter volumes extracted from the respective regions using the Jülich white matter atlas (whole white matter, corpus callosum, genu, body and splenium) (28, 61).

List of Supplementary Materials

Supplementary materials and methods

figure S1. Significant degree centrality differences between ASD and TD observed in the EU-AIMS LEAP dataset with and without control for motion in the group comparisons

figure S2. Connectivity indices computed based on the identified degree centrality alterations

figure S3. Correlations between the initial degree centrality findings and derived connectivity indices showing the largest effect sizes for differentiation between ASD and TD

figure S4. Plots of comorbidity status versus DC and hyper- and hypo-connectivity indices from ABIDE II cohort

table S1. Mean motion in ASD and TD

table S2. Results of analyses of variance testing for age category- and sex-by-diagnosis interactions on the hyper- and hypo-connectivity indices

table S3. Effects of medication on functional connectivity measures

table S4. Contingency table of psychiatric comorbidity versus current medication status in ASD from ABIDE II

table S5. Outcomes of EU-AIMS LEAP general linear model analysis using ASD indices to predict clinical scores

table S6. Age characteristics of the ADI and VABS subpopulations used for correlation with clinical scales

table S7. Volumetric comparisons between ASD and TD in the EU-AIMS LEAP cohort

table S8. Associations between structural and functional connectivity measures in the EU-AIMS LEAP cohort

table S9. Additional clinical characteristics of the EU-AIMS LEAP population

table S10. Additional clinical characteristics of the ABIDE I population

table S11. Additional clinical characteristics of the ABIDE II population

table S12. Summary of scanning parameters for each participating site in the EU-AIMS LEAP consortium

table S13. Summary of scanning parameters for each participating site in the ABIDE I cohort

table S14. Summary of scanning parameters for each participating site in the ABIDE II cohort

table S15. Summary of scanning parameters used in the InFoR cohort

References

1. A. P. Association, *Diagnostic and statistical manual of mental disorders (DSM-5®)* (American Psychiatric Pub, 2013).
2. G. Baird, E. Simonoff, A. Pickles, S. Chandler, T. Loucas, D. Meldrum, T. Charman, Prevalence of disorders of the autism spectrum in a population cohort of children in South Thames: the Special Needs and Autism Project (SNAP), *The lancet* **368**, 210–215 (2006).
3. C. J. Newschaffer, L. A. Croen, J. Daniels, E. Giarelli, J. K. Grether, S. E. Levy, D. S. Mandell, L. A. Miller, J. Pinto-Martin, J. Reaven, A. M. Reynolds, C. E. Rice, D. Schendel, G. C. Windham, The Epidemiology of Autism Spectrum Disorders, *Annu. Rev. Public Health* **28**, 235–258 (2007).
4. S. H. Baum, R. A. Stevenson, M. T. Wallace, Behavioral, perceptual, and neural alterations in sensory and multisensory function in autism spectrum disorder, *Prog. Neurobiol.* **134**, 140–160 (2015).
5. M. Sahin, M. Sur, Genes, circuits, and precision therapies for autism and related neurodevelopmental disorders, *Science* **350**, aab3897 (2015).
6. M. K. Belmonte, G. Allen, A. Beckel-Mitchener, L. M. Boulanger, R. A. Carper, S. J. Webb, Autism and abnormal development of brain connectivity, *J. Neurosci. Off. J. Soc. Neurosci.* **24**, 9228–9231 (2004).
7. M. P. Van Den Heuvel, H. E. Hulshoff Pol, Exploring the brain network: a review on resting-state fMRI functional connectivity, *Eur. Neuropsychopharmacol.* **20**, 519–534 (2010).
8. M. D. Fox, M. E. Raichle, Spontaneous fluctuations in brain activity observed with functional magnetic resonance imaging, *Nat. Rev. Neurosci.* **8**, 700–711 (2007).
9. S.-J. Weng, J. L. Wiggins, S. J. Peltier, M. Carrasco, S. Risi, C. Lord, C. S. Monk, Alterations of resting state functional connectivity in the default network in adolescents with autism spectrum disorders, *Brain Res.* **1313**, 202–214 (2010).
10. A. Di Martino, C.-G. Yan, Q. Li, E. Denio, F. X. Castellanos, K. Alaerts, J. S. Anderson, M. Assaf, S. Y. Bookheimer, M. Dapretto, B. Deen, S. Delmonte, I. Dinstein, B. Ertl-Wagner, D. A. Fair, L. Gallagher, D. P. Kennedy, C. L. Keown, C. Keysers, J. E. Lainhart, C. Lord, B. Luna, V. Menon, N. J. Minshew, C. S. Monk, S. Mueller, R.-A. Müller, M. B. Nebel, J. T. Nigg, K. O’Hearn, K. A. Pelphrey, S. J. Peltier, J. D. Rudie, S. Sunaert, M. Thioux, J. M. Tyszka, L. Q. Uddin, J. S. Verhoeven, N. Wenderoth, J. L. Wiggins, S. H. Mostofsky, M. P. Milham, The autism brain imaging data exchange: towards a large-scale evaluation of the intrinsic brain architecture in autism, *Mol. Psychiatry* **19**, 659–667 (2014).
11. M. Assaf, K. Jagannathan, V. D. Calhoun, L. Miller, M. C. Stevens, R. Sahl, J. G. O’boyle, R. T. Schultz, G. D. Pearlson, Abnormal functional connectivity of default mode sub-networks in autism spectrum disorder patients, *Neuroimage* **53**, 247–256 (2010).
12. C. L. Keown, P. Shih, A. Nair, N. Peterson, M. E. Mulvey, R.-A. Müller, Local functional overconnectivity in posterior brain regions is associated with symptom severity in autism spectrum disorders, *Cell Rep.* **5**, 567–572 (2013).
13. E. A. von dem Hagen, R. S. Stoyanova, S. Baron-Cohen, A. J. Calder, Reduced functional connectivity within and between ‘social’ resting state networks in autism spectrum conditions, *Soc. Cogn. Affect. Neurosci.* **8**, 694–701 (2012).

14. C. S. Monk, S. J. Peltier, J. L. Wiggins, S.-J. Weng, M. Carrasco, S. Risi, C. Lord, Abnormalities of intrinsic functional connectivity in autism spectrum disorders, *Neuroimage* **47**, 764–772 (2009).
15. J.-J. Paakki, J. Rahko, X. Long, I. Moilanen, O. Tervonen, J. Nikkinen, T. Starck, J. Remes, T. Hurtig, H. Haapsamo, Alterations in regional homogeneity of resting-state brain activity in autism spectrum disorders, *Brain Res.* **1321**, 169–179 (2010).
16. A. Di Martino, Z. Shehzad, C. Kelly, A. K. Roy, D. G. Gee, L. Q. Uddin, K. Gotimer, D. F. Klein, F. X. Castellanos, M. P. Milham, Relationship between cingulo-insular functional connectivity and autistic traits in neurotypical adults, *Am. J. Psychiatry* **166**, 891–899 (2009).
17. I. Dinstein, K. Pierce, L. Eyler, S. Solso, R. Malach, M. Behrmann, E. Courchesne, Disrupted neural synchronization in toddlers with autism, *Neuron* **70**, 1218–1225 (2011).
18. J. S. Anderson, T. J. Druzgal, A. Froehlich, M. B. DuBray, N. Lange, A. L. Alexander, T. Abildskov, J. A. Nielsen, A. N. Cariello, J. R. Cooperrider, E. D. Bigler, J. E. Lainhart, Decreased interhemispheric functional connectivity in autism, *Cereb. Cortex N. Y. N 1991* **21**, 1134–1146 (2011).
19. R.-A. Müller, P. Shih, B. Keehn, J. R. Deyoe, K. M. Leyden, D. K. Shukla, Underconnected, but how? A survey of functional connectivity MRI studies in autism spectrum disorders, *Cereb. Cortex N. Y. N 1991* **21**, 2233–2243 (2011).
20. K. Supekar, L. Q. Uddin, A. Khouzam, J. Phillips, W. D. Gaillard, L. E. Kenworthy, B. E. Yerys, C. J. Vaidya, V. Menon, Brain hyperconnectivity in children with autism and its links to social deficits, *Cell Rep.* **5**, 738–747 (2013).
21. A. Hahamy, M. Behrmann, R. Malach, The idiosyncratic brain: distortion of spontaneous connectivity patterns in autism spectrum disorder, *Nat. Neurosci.* **18**, 302–309 (2015).
22. D. R. Dajani, L. Q. Uddin, Local brain connectivity across development in autism spectrum disorder: A cross-sectional investigation, *Autism Res. Off. J. Int. Soc. Autism Res.* **9**, 43–54 (2016).
23. L. Q. Uddin, K. Supekar, V. Menon, Reconceptualizing functional brain connectivity in autism from a developmental perspective, *Front. Hum. Neurosci.* **7**, 458 (2013).
24. A. C. Linke, L. Olson, Y. Gao, I. Fishman, R.-A. Müller, Psychotropic Medication Use in Autism Spectrum Disorders May Affect Functional Brain Connectivity, *Biol. Psychiatry Cogn. Neurosci. Neuroimaging* **0** (2017), doi:10.1016/j.bpsc.2017.06.008.
25. A. Rausch, W. Zhang, C. F. Beckmann, J. K. Buitelaar, W. B. Groen, K. V. Haak, Connectivity-Based Parcellation of the Amygdala Predicts Social Skills in Adolescents with Autism Spectrum Disorder, *J. Autism Dev. Disord.* , 1–11 (2017).
26. R. L. Buckner, J. Sepulcre, T. Talukdar, F. M. Krienen, H. Liu, T. Hedden, J. R. Andrews-Hanna, R. A. Sperling, K. A. Johnson, Cortical hubs revealed by intrinsic functional connectivity: mapping, assessment of stability, and relation to Alzheimer’s disease, *J. Neurosci.* **29**, 1860–1873 (2009).
27. S. S. Sparrow, in *Encyclopedia of clinical neuropsychology*, (Springer, 2011), pp. 2618–2621.
28. A. Lefebvre, A. Beggiato, T. Bourgeron, R. Toro, Neuroanatomical Diversity of Corpus Callosum and Brain Volume in Autism: Meta-analysis, Analysis of the Autism Brain Imaging Data Exchange Project, and Simulation, *Biol. Psychiatry* **78**, 126–134 (2015).

29. E. W. Dickie, S. H. Ameis, S. Shahab, N. Calarco, D. E. Smith, D. Miranda, J. D. Viviano, A. N. Voineskos, Personalized intrinsic network topography mapping and functional connectivity deficits in Autism Spectrum Disorder, *Biol. Psychiatry* (2018).
30. L. Cerliani, M. Mennes, R. M. Thomas, A. D. Martino, M. Thioux, C. Keyzers, Increased Functional Connectivity Between Subcortical and Cortical Resting-State Networks in Autism Spectrum Disorder, *JAMA Psychiatry* **72**, 767–777 (2015).
31. A. J. Khan, A. Nair, C. L. Keown, M. C. Datko, A. J. Lincoln, R.-A. Müller, Cerebro-cerebellar Resting-State Functional Connectivity in Children and Adolescents with Autism Spectrum Disorder, *Biol. Psychiatry* **78**, 625–634 (2015).
32. M. D. Greicius, B. Krasnow, A. L. Reiss, V. Menon, Functional connectivity in the resting brain: a network analysis of the default mode hypothesis, *Proc. Natl. Acad. Sci.* **100**, 253–258 (2003).
33. V. Menon, Large-scale brain networks and psychopathology: a unifying triple network model, *Trends Cogn. Sci.* **15**, 483–506 (2011).
34. W. W. Seeley, V. Menon, A. F. Schatzberg, J. Keller, G. H. Glover, H. Kenna, A. L. Reiss, M. D. Greicius, Dissociable intrinsic connectivity networks for salience processing and executive control, *J. Neurosci.* **27**, 2349–2356 (2007).
35. E. Courchesne, J. Townsend, N. A. Akshoomoff, O. Saitoh, R. Yeung-Courchesne, A. J. Lincoln, H. E. James, R. H. Haas, L. Schreibman, L. Lau, Impairment in shifting attention in autistic and cerebellar patients., *Behav. Neurosci.* **108**, 848 (1994).
36. B. A. Corbett, L. J. Constantine, R. Hendren, D. Rocke, S. Ozonoff, Examining executive functioning in children with autism spectrum disorder, attention deficit hyperactivity disorder and typical development, *Psychiatry Res.* **166**, 210–222 (2009).
37. E. M. Barendse, M. P. Hendriks, J. F. Jansen, W. H. Backes, P. A. Hofman, G. Thoonen, R. P. Kessels, A. P. Aldenkamp, Working memory deficits in high-functioning adolescents with autism spectrum disorders: neuropsychological and neuroimaging correlates, *J. Neurodev. Disord.* **5**, 14 (2013).
38. M. Oldehinkel, M. Mennes, A. Marquand, T. Charman, J. Tillmann, C. Ecker, F. Dell’Acqua, D. Brandeis, T. Banaschewski, S. Baumeister, C. Moessnang, S. Baron-Cohen, R. Holt, S. Bölte, S. Durston, P. Kundu, M. V. Lombardo, W. Spooren, E. Loth, D. G. M. Murphy, C. F. Beckmann, J. K. Buitelaar, J. Ahmad, S. Ambrosino, B. Auyeung, T. Banaschewski, S. Baron-Cohen, S. Baumeister, C. F. Beckmann, S. Bölte, T. Bourgeron, C. Bours, M. Brammer, D. Brandeis, C. Brogna, Y. de Bruijn, J. K. Buitelaar, Altered Connectivity Between Cerebellum, Visual, and Sensory-Motor Networks in Autism Spectrum Disorder: Results from the EU-AIMS Longitudinal European Autism Project, *Biol. Psychiatry Cogn. Neurosci. Neuroimaging* **0** (2018), doi:10.1016/j.bpsc.2018.11.010.
39. C. J. Stoodley, A. M. D’Mello, J. Ellegood, V. Jakkamsetti, P. Liu, M. B. Nebel, J. M. Gibson, E. Kelly, F. Meng, C. A. Cano, Altered cerebellar connectivity in autism and cerebellar-mediated rescue of autism-related behaviors in mice, *Nat. Neurosci.* **20**, 1744 (2017).
40. F. Happé, R. Booth, R. Charlton, C. Hughes, Executive function deficits in autism spectrum disorders and attention-deficit/hyperactivity disorder: examining profiles across domains and ages, *Brain Cogn.* **61**, 25–39 (2006).
41. S. Solso, R. Xu, J. Proudfoot, D. J. Hagler Jr, K. Campbell, V. Venkatraman, C. C. Barnes, C. Ahrens-Barbeau, K. Pierce, A. Dale, Diffusion tensor imaging provides evidence of possible axonal

- overconnectivity in frontal lobes in autism spectrum disorder toddlers, *Biol. Psychiatry* **79**, 676–684 (2016).
42. B. Zikopoulos, H. Barbas, Changes in prefrontal axons may disrupt the network in autism, *J. Neurosci.* **30**, 14595–14609 (2010).
43. D. H. Geschwind, P. Levitt, Autism spectrum disorders: developmental disconnection syndromes, *Curr. Opin. Neurobiol.* **17**, 103–111 (2007).
44. E. B. Torres, K. Denisova, Motor noise is rich signal in autism research and pharmacological treatments, *Sci. Rep.* **6**, 37422 (2016).
45. J. D. Power, K. A. Barnes, A. Z. Snyder, B. L. Schlaggar, S. E. Petersen, Spurious but systematic correlations in functional connectivity MRI networks arise from subject motion, *Neuroimage* **59**, 2142–2154 (2012).
46. A. M. Maillard, A. Ruef, F. Pizzagalli, E. Migliavacca, L. Hippolyte, S. Adaszewski, J. Dukart, C. Ferrari, P. Conus, K. Männik, The 16p11. 2 locus modulates brain structures common to autism, schizophrenia and obesity, *Mol. Psychiatry* (2014).
47. R. A. Mukherjee, P. A. Cook, S. H. Norgate, A. D. Price, Neurodevelopmental outcomes in individuals with fetal alcohol spectrum disorder (FASD) with and without exposure to neglect: Clinical cohort data from a national FASD diagnostic clinic, *Alcohol* (2018).
48. A. Kolevzon, R. Gross, A. Reichenberg, Prenatal and Perinatal Risk Factors for Autism: A Review and Integration of Findings, *Arch. Pediatr. Adolesc. Med.* **161**, 326–333 (2007).
49. A. Hogart, D. Wu, J. M. LaSalle, N. C. Schanen, The comorbidity of autism with the genomic disorders of chromosome 15q11. 2-q13, *Neurobiol. Dis.* **38**, 181–191 (2010).
50. E. Loth, W. Spooren, L. M. Ham, M. B. Isaac, C. Auriche-Benichou, T. Banaschewski, S. Baron-Cohen, K. Broich, S. Bölte, T. Bourgeron, T. Charman, D. Collier, F. de Andres-Trelles, S. Durston, C. Ecker, A. Elferink, M. Haberkamp, R. Hemmings, M. H. Johnson, E. J. H. Jones, O. S. Khwaja, S. Lenton, L. Mason, V. Mantua, A. Meyer-Lindenberg, M. V. Lombardo, L. O’Dwyer, K. Okamoto, G. J. Pandina, L. Pani, A. M. Persico, E. Simonoff, S. Tauscher-Wisniewski, J. Llinares-Garcia, S. Vamvakas, S. Williams, J. K. Buitelaar, D. G. M. Murphy, Identification and validation of biomarkers for autism spectrum disorders, *Nat. Rev. Drug Discov.* **15**, 70–70 (2016).
51. E. Loth, T. Charman, L. Mason, J. Tillmann, E. J. H. Jones, C. Wooldridge, J. Ahmad, B. Auyeung, C. Brogna, S. Ambrosino, T. Banaschewski, S. Baron-Cohen, S. Baumeister, C. Beckmann, M. Brammer, D. Brandeis, S. Bölte, T. Bourgeron, C. Bours, Y. de Bruijn, B. Chakrabarti, D. Crawley, I. Cornelissen, F. D. Acqua, G. Dumas, S. Durston, C. Ecker, J. Faulkner, V. Frouin, P. Garces, D. Goyard, H. Hayward, L. M. Ham, J. Hipp, R. J. Holt, M. H. Johnson, J. Isaksson, P. Kundu, M.-C. Lai, X. L. D’ardhuy, M. V. Lombardo, D. J. Lythgoe, R. Mandl, A. Meyer-Lindenberg, C. Moessnang, N. Mueller, L. O’Dwyer, M. Oldehinkel, B. Oranje, G. Pandina, A. M. Persico, A. N. V. Ruigrok, B. Ruggeri, J. Sabet, R. Sacco, A. S. J. Cáceres, E. Simonoff, R. Toro, H. Tost, J. Waldman, S. C. R. Williams, M. P. Zwiers, W. Spooren, D. G. M. Murphy, J. K. Buitelaar, The EU-AIMS Longitudinal European Autism Project (LEAP): design and methodologies to identify and validate stratification biomarkers for autism spectrum disorders, *Mol. Autism* **8**, 24 (2017).
52. A. Di Martino, D. O’Connor, B. Chen, K. Alaerts, J. S. Anderson, M. Assaf, J. H. Balsters, L. Baxter, A. Beggiato, S. Bernaerts, L. M. E. Blanken, S. Y. Bookheimer, B. B. Braden, L. Byrge, F. X. Castellanos, M. Dapretto, R. Delorme, D. A. Fair, I. Fishman, J. Fitzgerald, L. Gallagher, R. J. J. Keehn, D. P. Kennedy, J. E. Lainhart, B. Luna, S. H. Mostofsky, R.-A. Müller, M. B. Nebel, J. T. Nigg, K. O’Hearn, M. Solomon, R. Toro, C. J. Vaidya, N. Wenderoth, T. White, R. C. Craddock, C.

- Lord, B. Leventhal, M. P. Milham, Enhancing studies of the connectome in autism using the autism brain imaging data exchange II, *Sci. Data* **4**, 170010 (2017).
53. J. L. Matson, M. Shoemaker, Intellectual disability and its relationship to autism spectrum disorders, *Res. Dev. Disabil.* **30**, 1107–1114 (2009).
54. A. D. Cox, J. Virues-Ortega, F. Julio, T. L. Martin, Establishing motion control in children with autism and intellectual disability: Applications for anatomical and functional MRI, *J. Appl. Behav. Anal.* **50**, 8–26 (2017).
55. K. J. Friston, S. Williams, R. Howard, R. S. Frackowiak, R. Turner, Movement-related effects in fMRI time-series, *Magn. Reson. Med.* **35**, 346–355 (1996).
56. X.-W. Song, Z.-Y. Dong, X.-Y. Long, S.-F. Li, X.-N. Zuo, C.-Z. Zhu, Y. He, C.-G. Yan, Y.-F. Zang, REST: a toolkit for resting-state functional magnetic resonance imaging data processing, *PLoS One* **6**, e25031 (2011).
57. J. Ashburner, A fast diffeomorphic image registration algorithm, *Neuroimage* **38**, 95–113 (2007).
58. K. Murphy, R. M. Birn, D. A. Handwerker, T. B. Jones, P. A. Bandettini, The impact of global signal regression on resting state correlations: Are anti-correlated networks introduced?, *NeuroImage* **44**, 893–905 (2009).
59. J. Dukart, F. Sambataro, A. Bertolino, Distinct Role of Striatal Functional Connectivity and Dopaminergic Loss in Parkinson's Symptoms, *Front. Aging Neurosci.* **9** (2017), doi:10.3389/fnagi.2017.00151.
60. J. Dukart, A. Bertolino, When Structure Affects Function—The Need for Partial Volume Effect Correction in Functional and Resting State Magnetic Resonance Imaging Studies, *PLoS One* **9**, e114227 (2014).
61. T. P. DeRamus, R. K. Kana, Anatomical likelihood estimation meta-analysis of grey and white matter anomalies in autism spectrum disorders, *NeuroImage Clin.* **7**, 525–536 (2015).

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Authors’ contributions:

JD designed the overall study, processed the data, performed all analyses related to derived hyper- and hypo-connectivity indices and all analyses testing for associations with structural, clinical and demographic measures. SH performed the initial DC analyses. JD and SH wrote the manuscript. AB, JFH, GDH, FB, JS and CHC contributed to interpretation of the EU-AIMS LEAP data. TC, JT, CFB, EL, DM, MO, CB, AR, JKB, WS and PG contributed to design, collection and interpretation of the EU-AIMS LEAP data. CC, CB, XL and AB contributed to design and interpretation of the InFoR data. MD, ML, IS, AG, RD, CL, JH, MLL, SG, JD, SH, MC, AA, MB and AnB contributed to design and collection of the InFoR data. All authors reviewed and commented on the manuscript.

Competing interests:

SH, JFH, CHC, PG, WS, XL, CB, AB, CC, FB, GDH, JS and JD are current or former employees of F. Hoffmann-La Roche Ltd. and received support in form of salaries. JKB has been in the past 3 years a consultant to / member of advisory board of / and/or speaker for Janssen Cilag BV, Eli Lilly, Lundbeck, Shire, Roche, Novartis, and Servier. He is not an employee of any of these companies, and not a stock shareholder of any of these companies. He has no other financial or material support, including expert testimony, patents, and royalties. CB, AR, MO, MB, AA, MC, SG, MLL, JH, CFB, EL, DM, TC, JT, CL, RD, AnB, AG, IS, ML and MD have no conflict of interests.

Data and materials availability: All the data used for this study are present in the main text or in the supplementary material and methods file

Table 1 Clinical and demographic characteristics

	EU-AIMS LEAP			ABIDE I			ABIDE II			InFoR		
	ASD	TD	Stats (test value, df, p-value)	ASD	TD	Stats (test value, df, p-value)	ASD	TD	Stats (test value, df, p-value)	ASD	TD	Stats (test value, df, p-value)
N	202	192	-	299	376	-	306	391	-	34	25	-
Male/female	142/60	124/68	1.5,1.,26	268/31	313/63	5.7,1.,017	262/44	263/127	30.4,1<.001	26/8	19/6	0.0,1.,967
Age±SD	17.5±5.3	17.4±5.7	0.1,392,915	17.5±7.7	17.7±7.8	.3,673.,776	14.0±6.8	13.6±6.2	.8,695,428	29.5±8.9	30.6±8.3	.5,57.,638
Child/Adol/Adult	35/76/9	43/71/78	1.7,2.,434	69/118/112	85/147/144	.1,2.,974	147/85/74	234/77/80	10.3,2,006	0/0/34	0/0/25	-
IQ (mean±SD, N)	106±14.9	109±12.6	2.1,392,033	106.3±16.0	112.0±12.1	5.3,673<.001	107.0±16.0	115.7±12.5	8.0,695,<.001	104,3±18.7	.6±17.5	.9,54.,392
DSM IV diag (none/ ASD/ Asperger/ PDD-NOS)	-	-	-	16/204/60/16	-	-	121/55/78/52	-	-	-	-	-
On medication (N)	54	2	-	61	1	-	81	17	-	-	-	-
ADOS total (mean±SD, N)	10.1±4.9	170	-	11.9±3.7	1.3±1.4	15.4,287,<.001	10±3.7,167	1.8±1.7,38	13.4,203,<.001	-	-	-
ADI social (mean±SD, N)	15.7±6.7	191	-	19.9±5.4	1.84	-	18.7±5.8	2,17	-	16.2±6.4	25	-
ADI communication (mean±SD, N)	12.9±5.8	191	-	15.8±4.7	1.85	-	14.9±4.6	2,17	-	9.4±5.3	25	-
ADI RRB (mean±SD, N)	4.1±2.6	191	-	6.1±2.5	1.85	-	5.5±2.5	2,17	-	2.2±1.3	22	-
SRS t-score	68.9±1.9	45.9±7.5	16.6,253,<.001	-	-	-	75.4±1.9	3.5,256	-	-	-	-
VABS adaptive behavior composite (mean±SD, N)	70.0±1.9	3.87	-	77.3±1.9	3.7,60	-	82.8±1.9	0.9,61	-	-	-	-
VABS daily living skills (mean±SD, N)	71.0±1.9	3.87	-	83.2±1.9	5.2,60	-	87.4±1.9	0.4,61	-	-	-	-
VABS socialization (mean±SD, N)	69.3±1.9	6.88	-	75.0±1.9	6.5,60	-	82.5±1.9	2.5,61	-	-	-	-
VABS communication (mean±SD, N)	77.3±1.9	4.9,88	-	79.3±1.9	5.5,60	-	87.3±1.9	3.9,61	-	-	-	-

ADI – Autism Diagnostic Interview, ADOS – Autism Diagnostic Observation Schedule, ASD – Autism spectrum disorder patients, PDD-NOS – Pervasive developmental disorder not otherwise specified, RRB – Restricted and repetitive behaviors, SD – Standard deviation, SRS – Social Responsiveness Scale, TD – Typically developing controls

Table 2. Regions showing significant DC differences in the EU-AIMS LEAP cohort between individuals with ASD and TD

Contrast	Anatomical region	Cluster size	Exact cluster p-value	T-value	MNI coordinates
ASD > TD	Bilateral: Parietal, posterior cingulate, precuneus, primary visual	3066	0.008	5.35*	-33 54 39
ASD > TD	Bilateral: Lateral and medial prefrontal, premotor, supplementary motor, anterior cingulate Right: Anterior insula, caudate	4851	0.001	5.33*	36 42 3
TD > ASD	Bilateral: Primary sensory, primary motor, middle cingulate Right: Posterior insula, temporal cortex, amygdala, hippocampus, entorhinal	5777	<.001	5.41*	12 -33 69

* Significant at a whole-brain voxel-wise family-wise error corrected threshold of $p < .05$

Table 3. Effect sizes and p-values for the hyper- and hypo-connectivity indices computed using cortical, subcortical and cerebellar subregions comparing ASD and TD

"Outside" region	Hyper-connectivity index (Cohen's d;p)	Hypo-connectivity index (Cohen's d;p)
Cerebellar	-0.12;.233	0.05;.619
Cortical	-0.76;<.001	0.46;<.001
Subcortical	0.05;.604	-0.10;.332

ASD – Autism spectrum disorder, TD – typically developing controls

Table 4 Results of correlations between hyper- and hypo-connectivity indices and clinical scales

Scale	Connectivity measure	EU-AIMS LEAP	ABIDE I	ABIDE II
ADI communication total	hyper-conn	$F(1,182)=5; p=0.026$	$F(1,171)=8.7; p=0.004$	$F(1,204)=0.3; p=0.602$
VABS daily living standard	hyper-conn	$F(1,80)=5.3; p=0.024$	$F(1,55)=4.1; p=0.048$	$F(1,55)=0.1; p=0.75$

ADI – Autism Diagnostic Interview, ASD – Autism spectrum disorder, DC – Degree centrality, conn – connectivity index, VABS – Vineland Adaptive Behavior Scales, Significant ($p < 0.05$) relationships between functional connectivity indices and respective clinical scales are indicated in bold.

Figure 1. Outcomes of functional connectivity comparisons between individuals with ASD and TD. (A) Representative reconstruction of a brain illustrating regions showing significant DC increases (red) and decreases (blue) in individuals with ASD in the reference EU-AIMS LEAP cohort. **(B)** Distribution of DC values (first eigenvariate over the masks' voxels) across subjects for the hyper-connected (top row, red) and hypo-connected (bottom row, blue) network separately for individuals with ASD and TD. **(C)** Unthresholded *t*-maps showing regions with increased (warm colors) and decreased (cold colors) DC in individuals with ASD as compared to TD. **(D)** Spatial correspondence of the *t*-maps between the reference data set (EU-AIMS LEAP) and all remaining replication data sets (ABIDE I, ABIDE II, InFoR) in form of a pair-wise correlation matrix and individual correlation plots. A – anterior, ASD – autism spectrum disorder, *d* – Cohen's *d* effect size, DC – degree centrality, FWE – family-wise error, L – left, P – posterior, R – right, TD – typically developing. **p* < 0.05, ***p* < 0.01, ****p* < 0.001 as obtained using independent samples *t*-tests (b) and Pearson correlation analyses (d).

Figure 2. Functional connectivity indices and their effect size (A) Distribution of the Pearson correlation coefficients between pairs of voxel-wise blood oxygen level dependence (BOLD) activity time courses observed in ASD and TD in the EU-AIMS LEAP cohort. **(B)** Effect sizes of the functional connectivity indices based on DC masks for the contrasts ASD > TD (left) and TD > ASD (right). Indices: mean connectivity (μ), variance of connectivity (σ) and proportion of connected voxels (π) within (π_1), outside (π_2), and from within to outside (π_3) the DC alteration masks and shifts in connectivity from within to outside (π_4) and from outside to within (π_5) the DC alteration mask. Details on computation are provided in figure S3. ASD – autism spectrum disorder, DC – degree centrality, o – outside, *r* – correlation coefficient, *r_z* – Fisher's *z*-transformed correlation coefficient, TD – Typically developing, w – within, prop – proportion, Std – standard deviation. **p* < 0.05, ***p* < 0.01, ****p* < 0.001 as obtained using independent samples *t*-tests, -- indicates the original negative sign of the respective index. **(C)** Effect size for the *hyper- and hypo-connectivity indices* obtained in the EU-AIMS LEAP dataset plotted versus the applied correlational threshold. Dashed line indicates the correlational threshold used for the initial degree centrality computation

Figure 3. Relationships between hyper- and hypo-connectivity indices, age and clinical outcomes. (A) Effect size of the *hyper-connectivity index* for all patients and split by age groups (children [<12 years], adolescents [12–18 years], adults [>18 years]) for contrast ASD > TD (red). **(B)** The equivalent of (a) for the hypo-connectivity index (blue). **(C)** Correlations of the hyper-connectivity indices with ASD clinical scales in the EU-AIMS LEAP and ABIDE I data sets (adjusted for the effects of age, sex, site and IQ). **(D)** Correlations between ADI communication and VABS daily living skills and communication subscales, ADI – Autism Diagnostic Interview, VABS – Vineland Adaptive Behavior Scale, ASD – autism spectrum disorder, comm – communication, dls – daily living skills, **p* < 0.05, ***p* < 0.01, ****p* < 0.001 as obtained using independent samples *t*-tests.

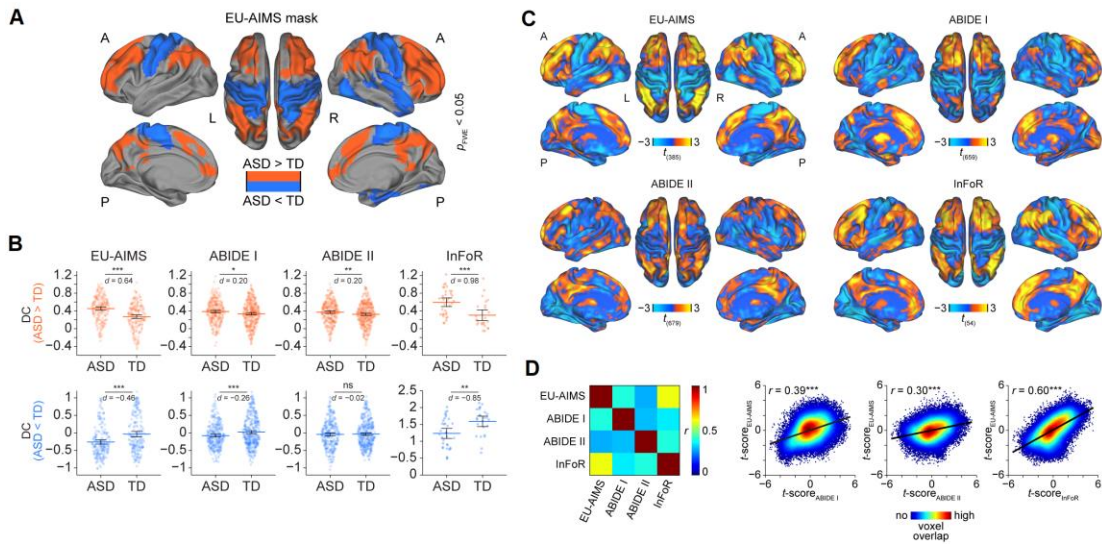


Figure 1

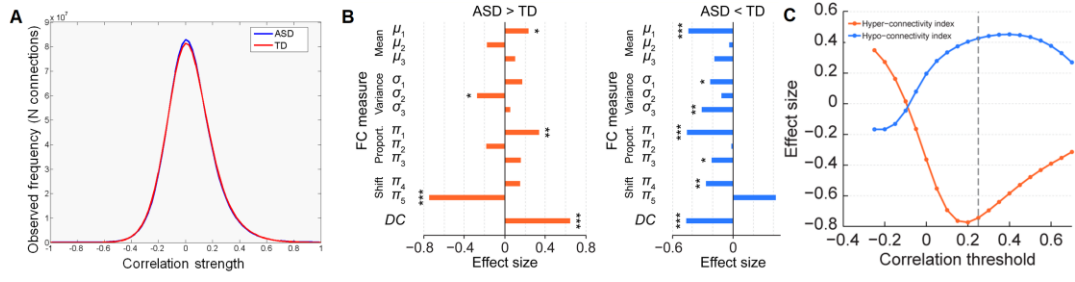


Figure 2

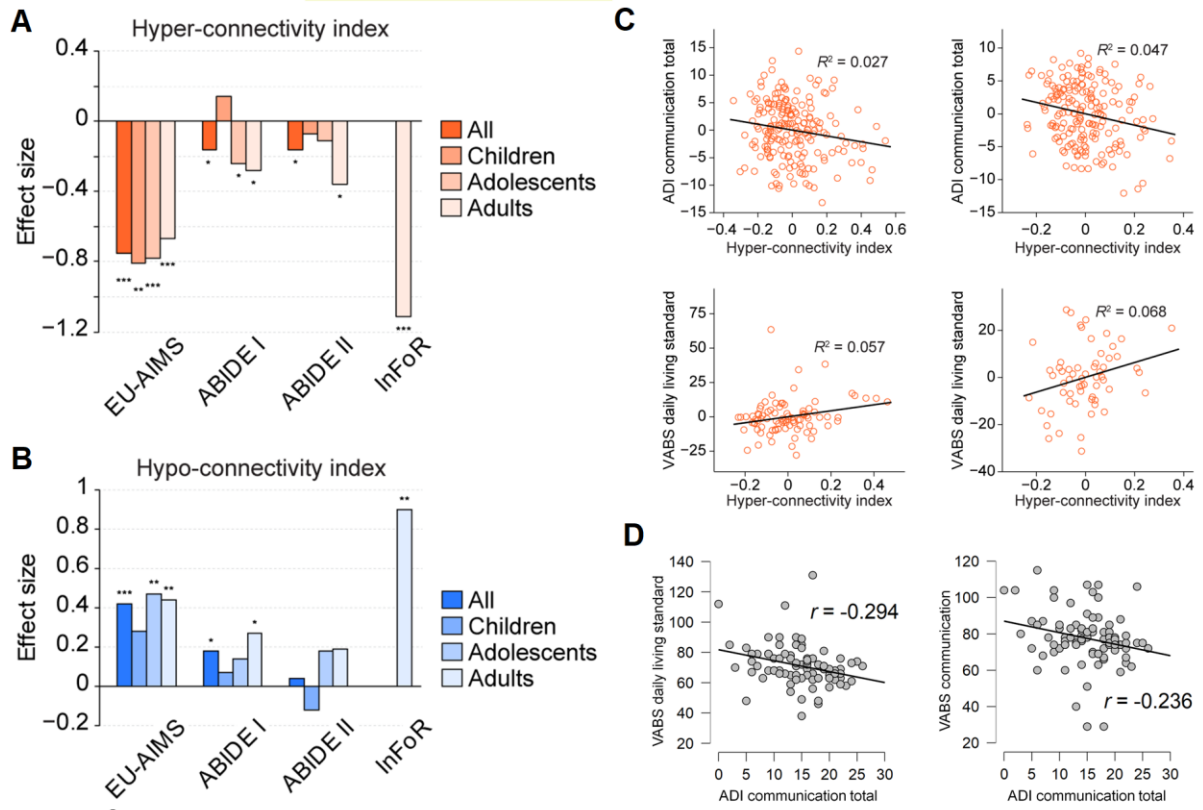


Figure 3