

HHS Public Access

Author manuscript *Autism Res.* Author manuscript; available in PMC 2019 December 07.

Published in final edited form as:

Autism Res. 2019 December ; 12(12): 1758–1773. doi:10.1002/aur.2193.

Early Patterns of Functional Brain Development Associated With Autism Spectrum Disorder in Tuberous Sclerosis Complex

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Abstract

Tuberous sclerosis complex (TSC) is a rare genetic disorder that confers a high risk for autism spectrum disorders (ASD), with behavioral predictors of ASD emerging early in life. Deviations in structural and functional neural connectivity are highly implicated in both TSC and ASD. For the first time, we explore whether electroencephalographic (EEG) measures of neural network function precede or predict the emergence of ASD in TSC. We determine whether altered brain function (a) is present in infancy in TSC, (b) differentiates infants with TSC based on ASD diagnostic status, and (c) is associated with later cognitive function. We studied 35 infants with TSC (N= 35), and a group of typically developing infants (N= 20) at 12 and 24 months of age. Infants with TSC were later subdivided into ASD and non-ASD groups based on clinical evaluation. We measured features of spontaneous alpha oscillations (6–12 Hz) that are closely associated with neural network development: alpha power, alpha phase coherence (APC), and peak

Supporting Information

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The remaining authors (A.D., K.J.V., and C.A.N.) do not have any conflict of interests associated with this study.

Additional supporting information may be found online in the Supporting Information section at the end of the article. **Appendix S1.** Supporting Information.

alpha frequency (PAF). Infants with TSC demonstrated reduced interhemispheric APC compared to controls at 12 months of age, and these differences were found to be most pronounced at 24 months in the infants who later developed ASD. Across all infants, PAF at 24 months was associated with verbal and nonverbal cognition at 36 months. Associations between early network function and later neurodevelopmental and cognitive outcomes highlight the potential utility of early scalable EEG markers to identify infants with TSC requiring additional targeted intervention initiated very early in life.

Lay Summary:

Approximately half of infants with tuberous sclerosis complex (TSC) develop autism. Here, using EEG, we find that there is a reduction in communication between brain regions during infancy in TSC, and that the infants who show the largest reductions are those who later develop autism. Being able to identify infants who show early signs of disrupted brain development may improve the timing of early prediction and interventions in TSC, and also help us to understand how early brain changes lead to autism.

Keywords

tuberous sclerosis complex; autism spectrum disorder; functional connectivity; infancy; electroencephalography; cognitive function; alpha oscillations

Introduction

Tuberous sclerosis complex (TSC) is a rare autosomal-dominant genetic syndrome caused by the inactivation of TSC1 or TSC2 genes. The TSC1/TSC2 protein complex plays a critical role in regulating cell growth and proliferation via mechanistic target of rapamycin (mTOR) signaling. In TSC, mTOR remains constitutively active, resulting in unrestricted cell growth and proliferation throughout the body. While TSC is considered a multi-system disorder (affecting the kidneys, heart, eyes, lungs, and skin), the central nervous system is consistently and prominently affected [Curatolo, Bombardieri, & Jozwiak, 2008]. The majority of individuals with TSC have hamartomas in the brain, including cortical tubers, subependymal nodules, and/or subependymal giant-cell astrocytomas [Northrup, Krueger, & International Tuberous Sclerosis Complex Consensus Group, 2013; Roth et al., 2013], and epilepsy is estimated to occur in around 70-90% of cases [Chu-Shore, Major, Camposano, Muzykewicz, & Thiele, 2010; Curatolo et al., 2008; Devlin, Shepherd, Crawford, & Morrison, 2006; Saxena & Sampson, 2015; Wu et al., 2016]. Neurodevelopmental or neuropsychiatric disorders (referred to as TSC-associated neuropsychiatric disorders); [de Vries et al., 2015; Leclezio, Jansen, Whittemore, & de Vries, 2015] are common. About 25-60% of children with TSC will receive a diagnosis of Autism Spectrum Disorder (ASD) [Capal et al., 2017; Granader et al., 2010; Jeste, Sahin, Bolton, Ploubidis, & Humphrey, 2008; Vignoli et al., 2015], and more than 50% have some degree of cognitive impairment [Curatolo, Moavero, & de Vries, 2015; Joinson et al., 2003].

TSC is often diagnosed in utero or shortly after birth through identification of cardiac or cortical hamartomas [Datta, Hahn, & Sahin, 2008; Davis et al., 2017], facilitating the

examination of these infants well before diagnoses of ASD or intellectual disability (ID) are made. Longitudinal studies have revealed early behavioral differences between infants and toddlers with TSC who later receive an ASD diagnosis and those who do not [Jeste et al., 2016; McDonald et al., 2017]. Infants who develop ASD are found to demonstrate social communication deficits by 9 months of age [McDonald et al., 2017] and lower cognitive function by 12 months of age. Interestingly, toddlers not diagnosed with ASD demonstrate social communication profiles indistinguishable from typically developing (TD) children, despite having comparable cortical tuber burden and epilepsy severity to those with ASD [Jeste et al., 2016]. Given the high risk of neurodevelopmental disorders in TSC, with symptoms emerging in infancy, we asked whether there are neurobiological markers of atypical brain development that relate to and predict neurodevelopmental disorders in TSC.

Commensurate with our understanding of the neurobiological mechanisms underlying neurodevelopmental disorders as a whole [Geschwind & Levitt, 2007; Jeste, Frohlich, & Loo, 2015], it is most likely that altered patterns of neural connectivity (outside of cortical tubers, subependymal nodules, and astrocytomas) account for atypical neurodevelopment in TSC [Ebrahimi-Fakhari & Sahin, 2015; Jülich & Sahin, 2014; Peters et al., 2012]. The critical role of TSC1/2 proteins in neural connectivity is supported by evidence from preclinical studies that demonstrate abnormalities in neuronal migration and morphology [Knox et al., 2007; Nie et al., 2010], synaptic plasticity and function [Ehninger et al., 2008; Zeng et al., 2007], excitation-inhibition balance [Bateup et al., 2013], and myelination patterns [Ercan et al., 2017] in the context of a TSC mutation. In addition, structural imaging studies in individuals with TSC demonstrate widespread alterations in structural connections [Baumer et al., 2015; Krishnan et al., 2010; Lewis et al., 2012; Peters et al., 2012], relevant to the variability seen in neurodevelopmental outcomes. Using an unbiased whole-brain connectivity analysis, Im et al. [2016] identified altered structural connectivity in TSC with markedly reduced interhemispheric connectivity (aged 3-24 years) compared to a control sample, most prominent in those with developmental delay [Im et al., 2016]. Similarly, deficits in the integrity of white matter microstructure are found to be more pronounced in individuals with TSC who also have an ASD diagnosis [Baumer et al., 2017; Lewis et al., 2012; Peters et al., 2012]. Finally, and highly relevant to the present study, a pattern of functional hypoconnectivity is identified using electroencephalography (EEG) in individuals with TSC across a wide age range, with hypoconnectivity (in some cortical areas) more prominent in individuals with both TSC and ASD [Peters et al., 2013].

EEG offers a scalable and feasible method to capture the specific changes in brain dynamics that occur during infancy. EEG measures brain function in three important dimensions: space, time, and frequency [Kida, Tanaka, & Kakigi, 2016]. Here, we focus on oscillations which occur within the alpha range (6–12 Hz), which represent the dominant signal in task-independent environments [Guggisberg et al., 2008], and have been well characterized during early infancy [Saby & Marshall, 2012]. In addition to measuring the power of alpha oscillations, we focus on two measures of alpha activity, alpha phase coherence (APC) and peak alpha frequency (PAF), that are specifically sensitive to changes in neural connectivity that occur over the course of typical brain maturation, and in clinical populations. Across the oscillatory signals that coordinate large-scale neural communication, the coherence of neural signals in the alpha range, APC, is demonstrated to be particularly sensitive to both the

healthy development of [Smit et al., 2012], and disruptions to [Hinkley et al., 2011], functional and structural connectivity. Similarly, the frequency at which maximal power occurs within the alpha band, PAF, shows well-characterized increases with chronological age in TD children [Chiang, Rennie, Robinson, van Albada, & Kerr, 2011; Cragg et al., 2011; Dustman, Shearer, & Emmerson, 1999; Miskovic et al., 2015; Somsen, van't Klooster, van der Molen, van Leeuwen, & Licht, 1997; Stroganova, Orekhova, & Posikera, 1999]. In ASD, PAF relates to cognitive function, not age [Dickinson, DiStefano, Senturk, & Jeste, 2017]. In addition to age and cognition, both APC and PAF are associated with white matter architecture (axon growth and myelination) [Jann et al., 2012; Teipel et al., 2009], further suggesting that these metrics index the development of neural networks [Klimesch, Sauseng, & Hanslmayr, 2007; Valdés-Hernández et al., 2010].

As part of a longitudinal study of TSC, we utilized high-density EEG measures of spontaneous alpha oscillations to characterize the development of functional connectivity across first 2 years of life and to determine whether these early biomarkers informed neurodevelopmental outcomes. More specifically, we asked: (a) Do alpha oscillations (power, PAF, and PAC) differ between infants with and without TSC?; (b) Do alpha oscillations (power, PAF, and PAC) differentiate infants with TSC who develop ASD from those that do not?; and (c) Does PAF measured during infancy predict later cognitive function? Data from this study could reveal the earliest manifestations of atypical brain development that may precede and predict the development of ASD and ID in TSC. Such investigations can help elucidate mechanisms underlying the distinct (and often diverse) neurodevelopmental disorders, further stratify infants by risk for neurodevelopmental disorders, and provide objective measures of change with early intervention in high-risk infants.

Method

Participants

Infants were enrolled in a longitudinal, two-site study examining early development in TSC [Jeste et al., 2016; Jeste et al., 2014; McDonald et al., 2017]. Infants with confirmed TSC diagnoses [Northrup et al., 2013] were recruited through specialty clinics, newborn nurseries, pediatricians' offices, and the tuberous sclerosis alliance. TD control infants (described below) were recruited through institutional review board-approved databases.

Infants visited one of the two study sites throughout the first 2 years of life (at 6, 9, 12, and/or 24 months of age) to undergo EEG recording. At 18, 24, and/or 36 months, infants underwent behavioral assessments to determine ASD symptoms (described in detail below), which was followed by a Best Clinical Estimate diagnostic process. The study protocol is illustrated in Figure S1. Although we collected EEG data in a small subset of infants at age 6 and 9 months, in order to achieve adequate sample sizes to interrogate statistical differences between groups, we focused on 12- and 24-month time points.

To be included in the present analyses, participants had to have EEG data for at least one of the two time points (12 and 24 months). To ensure that the EEG inclusion criteria did not

bias our representation of participants, we compared cognitive function between infants in the present sample, and those who were excluded due to a lack of 12- or 24-month EEG data (N=8). Both verbal (TD: P=0.302; ASD+: P=0.441; ASD-: P=0.312) and nonverbal (TD: P=0.635; ASD+: P=0.312; ASD-: P=0.583) cognitive function (described below) were consistent across included and excluded infants, indicating that our sample was representative of study-wide participant demographics [Jeste et al., 2016, Jeste et al., 2014; McDonald et al., 2017]. Exclusion criteria for TD infants included prematurity (<37 weeks gestational age), birth trauma, developmental concerns, or a first-degree relative with ASD or ID. It should be noted that while "typically developing" may not traditionally be used to describe infants, our later follow-up time points were used to confirm that no developmental concerns were present in the TD group at 36 months of age. The study received ethical approval from the institutional review board at both study sites. Parents provided informed written consent, in accordance with the Declaration of Helsinki.

TD (N= 22) and TSC (N= 36) infants completed an EEG recording session at 12 or 24 months. Of this sample, two TD participants and one TSC participant were excluded from the current analyses due to extensive EEG artifact. Final groups therefore consisted of 35 TSC (12 months: n = 18; 24 months: n = 28), and 20 TD infants (12 months: n = 20; 24 months: n = 12). Clinical genetics reports were available for n = 21 (60%) of the total TSC sample, with all TSC infants meeting clinical criteria for TSC [Northrup et al., 2013]. Demographic and developmental information for the sample is summarized in Table 1.

It should be noted that males were over-represented in the TSC group. While it has been shown that many behavioral features (including the presence of ASD and ID) are distributed equally across both sexes in TSC [Curatolo, Porfirio, Manzi, & Seri, 2004; de Vries, Hunt, & Bolton, 2007], larger samples in the future (described in the discussion) will allow the exploration of sex-related differences in early brain function in TSC.

Seizures, tubers, and medication.—Thirty-two (91%) of participants with TSC had seizure data available through parent-report and/or medical records. Of these individuals, the majority (29/32; 91%) had seizures. Age of onset ranged from day of birth to 22 months, with a median age of onset at 4 months (interquartile range = 4.56 months). Of those with seizures, 62.5% (20/32) had infantile spasms, a rate that is similar to previously published estimates [Davis et al., 2017]. Information regarding the presence of cortical tubers was present for 80% of the TSC sample (n = 28). Of the participants with available tuber information, 28 (100%) had tubers evident on clinical MRI. Of infants with available medications (AEDs) at either the 9, 12, and/or 24 month assessments. Only the two infants with TSC who had no history of seizures were naïve to AEDs. No history of seizures or medication use was reported for any TD infant. See Tables S1 and S2 for additional details regarding seizure history and medication.

Behavioral Assessments

Autism symptoms and diagnosis.—Clinical ASD symptoms were assessed using Modules 1 and 2 of the Autism Diagnostic Observation Schedule (ADOS; [Lord et al.,

2000]). Calibrated severity scores (CSS) were calculated in order to facilitate comparison of scores across modules [Gotham, Risi, Pickles, & Lord, 2007]. A diagnostic outcome of ASD (ASD+) was based on both meeting the ADOS cut off score for ASD (CSS 4), and clinical best estimate (based on DSM-5 criteria) determined by a trained clinician. Infants with TSC who did not show elevated ASD symptoms are described as ASD–.

Nineteen (95%) TD infants underwent an ADOS assessment. These were carried out at either 18 (n = 3; 15.8%), 24 (n = 6; 31.6%), or 36 months (n = 10; 52.6%). No TD infants met criteria for ASD. Of the 36 TSC infants included in the present study, 4 (11%) did not complete an ADOS assessment. These infants were included in the main group comparison (TSC compared to TD infants) but not included in any sub-analyses of ASD+ compared to ASD–. Of the four children with TSC who did not complete an ADOS assessment, one could not have the ADOS administered in their native language, and three were lost to follow-up. Of the 31 TSC infants that received an ADOS assessment at either 18 (n = 1; 3%), 24 (n = 6; 19%) or 36 months (n = 24; 77%), 17 were deemed to have met ASD criteria based on Best Clinical Estimate (55%), and 14 did not (45%). This is consistent with previously reported prevalence rates of ASD in TSC [Jeste et al., 2008].

Cognitive function.—Cognitive function was assessed at 36 months of age using a standardized measure of cognitive and motor development, the Mullen Scales of Early Learning [Mullen, 1995]. A verbal developmental quotient (VDQ) was calculated using the average age equivalent score of the Receptive Language and Expressive Language subscale scores (divided by chronological age), and a nonverbal developmental quotient (NVDQ) was calculated using the average age equivalent score of the Visual Reception and Fine Motor subscale scores (divided by chronological age) [Akshoomoff, 2006]. Cognitive function measures were only available for 50% of the TD group and 66% of the TSC group, due to a large number of participants being lost to follow up before cognitive function was assessed at 36 months (see Table 1).

EEG Acquisition and Processing

EEG recording.—Continuous EEG data were recorded using a high-density 128-channel HydroCel Geodesic Sensor Net (Electrical Geodesics Inc., Eugene, OR). EEG was recorded for 5 min in a dimmed, sound-attenuated room while infants sat in their caregiver's lap. Consistent with many other studies in developmental populations, a research assistant aimed to keep the infant settled and content for the duration of the recording by blowing bubbles [Dawson, Klinger, Panagiotides, Lewy, & Castelloe, 1995; McEvoy, Hasenstab, Senturk, Sanders, & Jeste, 2015; Stroganova et al., 1999; Tierney, Gabard-Durnam, Vogel-Farley, Tager-Flusberg, & Nelson, 2012; Webb et al., 2015; Gabard-Durnam, 2015; Levin, 2017]. Net Station 4.4.5 software was used to record from a Net Amps 300 amplifier with a lowpass analog filter cutoff frequency of 6 kHz. Data were sampled at 500 Hz and referenced to vertex at the time of recording. Electrode impedances were kept below 100 kΩ.

EEG data preprocessing.—All offline data processing and analyses were performed using EEGLAB [Delorme & Makeig, 2004], and in-house MATLAB scripts. Video recordings synchronized to EEG collection were first inspected to ensure that any periods of

EEG data reflecting periods during which the participant was not sat relatively still and looking at the bubbles (due to behavioral dysregulation, participant distraction, or emotional dysregulation) were omitted from analysis. This was carried out to ensure that the "spontaneous" EEG data analyzed were as consistent as possible across participants. Data were high pass filtered to remove frequencies below 1 Hz and low pass filtered to remove frequencies above 50 Hz, using a finite impulse response filter implemented in EEGLAB. Continuous data were then visually inspected, and noisy channels were removed. Following channel removal, data were interpolated to the international 10-20 system 25 channel montage [Jasper, 1958], in order to retain an adequate k factor for independent component analysis (ICA) decomposition (described below) [Onton & Makeig, 2006]. Based on visual inspection, sections of data that showed nonstereotyped bursts of electromyographic (EMG) signal, or other non-stereotyped artifacts, were then removed from the recording. ICA, a statistical blind source separation technique [Makeig, Jung, Bell, Ghahremani, & Sejnowski, 1997], was implemented to remove electrooculogram (EOG) and other stereotyped artifacts from the data. After decomposing the data into maximally independent components (IC), the scalp topography and time course of each IC was visually inspected. Any IC that represented a non-neural source (including EMG and EOG) was removed from the data. The experimenter was blinded to all participant details throughout the data cleaning process.

Data were then separated into three-second epochs. The minimum amount of artifact free data available across participants was 46 sec, thus the first 45 sec (15 epochs) of data from every participant underwent analyses. This data length represents an appropriate minimum threshold to gain reliable estimates of the characteristics of spontaneous EEG [Dickinson et al., 2017; Gudmundsson, Runarsson, Sigurdsson, Eiriksdottir, & Johnsen, 2007].

EEG data processing: Alpha power.—Welch's method, implemented using 2-sec Hamming windows with 50% overlap, was used to compute spectral power from cleaned EEG for each of the 25 channels. The resulting power spectra had approximately 0.5 Hz frequency resolution. Relative power was calculated by determining the proportion of total spectral power (1–50 Hz) accounted for by each frequency bin. Relative power values were then summed for frequencies within the alpha (6–12 Hz) range.

EEG data processing: Peak alpha frequency.—PAF was computed for three regions of interest: frontal, central, and occipital, in line with previous descriptions of atypical PAF in nonsyndromic ASD [Dickinson et al., 2017]. Using the computed power spectra (described above) from six channels (F3, F4, C3, C4, O1, and O2), PAF for each region was defined by averaging the power spectra from two channels (i.e., F3, F4 = frontal; C3, C4 = central; O1, O2 = occipital). A robust curve fitting procedure [fully described in Dickinson et al., 2017] was used to quantify PAF in an objective manner that is free of bias toward lower frequencies [Neto, Allen, Aurlien, Nordby, & Eichele, 2015], as is illustrated in Figure 1. If the curve fitting procedure failed, data were omitted from further analysis. Chi-square analyses revealed that there were no group differences in peak presence between TSC and TD groups, with 9.63% of participants at 12 months, and 3.3% of participants at 24 months having at least one region omitted from analysis.

EEG data processing: Alpha phase coherence.—Cleaned EEG data underwent surface Laplacian transform using realistic head geometry, based on the average head radius of participants in the present sample (12 months = 7.3 cm; 24 months = 7.8 cm). The surface Laplacian is the second spatial derivative of the scalp-recorded potentials for each electrode, thus transforming the scalp-recorded EEG into estimates of current source density (CSD). We employed a spline flexibility constant of m = 3, which represents adequate flexibility to prevent distortion of the original data [Kayser & Tenke, 2015].

Coherence analyses were then conducted on CSD values. CSD values were decomposed into frequency-time domain using Fast Fourier Transform with a fixed window size of 1,024 samples to generate a sequence of amplitude and phase components for the frequency bins between 6 and 12 Hz. Due to the hypothesis of the present study, coherence analysis was restricted to the alpha range, defined here as 6–12 Hz, a commonly used range in young children [McLaughlin, Fox, Zeanah, & Nelson, 2011; Oberman, Ramachandran, & Pineda, 2008].

Coherence was then computed in the form of event-related phase coherence from the aforementioned epochs using the newcrossf function provided by EEGLAB [Delorme & Makeig, 2004]. For each channel pair, a single-phase coherence value was calculated by averaging the phase coherence of all frequency bins encompassed by alpha band, across each time point, resulting in 300 unique average alpha coherence values for every possible electrode pair.

Statistical Analysis

In order to retain participants who only contributed EEG at one time point, analyses of EEG differences were conducted separately at 12 and 24 months of age. We confirmed that relevant EEG measurements did not significantly differ between the two sites of acquisition.

Statistical analysis: Group comparisons.—Two main group comparisons were performed: (a) TSC versus TD (b) ASD+ versus ASD–. Group means, *P*-values, *q*-values, and Cohen's *d* effect sizes are presented in Tables S3–S6. As a reminder, all infants included in ASD group comparisons had a TSC diagnosis. Identical analysis approaches (described below) were used for both of the group comparisons. Due to the large number of comparisons, type I errors were controlled using false discovery rate (FDR) methods (FDR = 0.15) [Benjamini & Hochberg, 1995].

Power: Instead of selecting predefined regions of interest, we took an unbiased, data-driven approach to examine power across the entire scalp. Group differences (in all 25 channels) were examined using test statistics, which represented normalized differences in group means. A permutation test, where group labels were randomly assigned to subjects in each resample, was used to estimate the distribution of the test statistics. Two separate permutation tests (100,000 permutations each) were used to study group differences at 12 and 24 months.

<u>PAF</u>: Two separate repeated measure ANCOVA's (with site entered as a covariate) were used to determine whether there were group differences in PAF at 12 and 24 months. To

include all possible data and study regional differences, each region was investigated separately in follow-up analyses.

APC: We also applied nonparametric permutation testing to study group differences in APC. We examined APC for every possible electrode pair combination across the scalp (300 electrode pairs). Group differences in APC were examined using two separate permutation tests (described above) at 12 and 24 months. Following preliminary analyses, analysis of covariance (ANCOVA) models (with VDQ and NVDQ entered as covariates) were used to assess group differences. It should be noted that while these models were used to determine whether group differences remained significant after controlling for cognitive function, these analyses were restricted to the portion of the sample for whom cognitive data were available.

Statistical analysis: PAF and cognitive function.—In order to maximize both sample size and range of cognitive function, correlations between cognitive function and PAF were collapsed across all infants (TD and TSC). A nonparametric Kendall's tau correlation was employed to examine relationships between cognitive function and PAF for each separate region of interest (frontal, central, and occipital), so as to make no assumptions about the distribution of the data.

Results

TSC/TD Group Comparison

Alpha power.—There were no alpha power differences between TSC and TD infants at 12 months (P > 0.39; q > 0.384) or 24 months (P > 0.006; q > 0.153). See Table S3 for means and effect sizes.

Peak alpha frequency.—At 12 months, there was no significant effect of group (F(1, 27) = 2.322, P = 0.139), or region (F(2, 54) = 0.379, P = 0.69), and no group × region interaction effect (F(2, 54) = 0.257, P = 0.774) on PAF. At 24 months, there was a significant effect of group (F(1, 34) = 5.909, P = 0.020), where PAF values were significantly lower in the TSC group compared to the TD group. There was also a significant effect of region (F(2, 68) = 3.549, P = 0.034), but no group × region interaction effect (F(2, 68) = 0.042, P = 0.959). PAF values for each region of interest are described in Table 2.

Alpha phase coherence.—At 12 months, infants with TSC (M = 0.24, SD = 0.03) showed significantly decreased APC compared to TD infants (M = 0.31, SD = 0.09) for a long-range temporal–parietal interhemispheric connection between electrodes T7 and C4 (P < 0.00013, q = 0.075, d = 1.058; see Fig. 2). ANCOVA demonstrated that group differences in APC for this connection (T7-C4) remained significant (in a subset of the sample) when covarying VDQ and NVDQ (F(1, 16) = 4.85, P = 0.04). This group difference remained unaltered when the two TD infants with the highest APC were removed from analyses (P = <0.022). There were no differences in APC for this connection the two sites of acquisition (P > 0.232). At 24 months, there were no significant differences in APC (P > 0.002; q > 0.561; see Table S4 for means and effect sizes).

ASD+/ASD- Group Comparison

Alpha power.—There were no power differences between ASD+ and ASD– infants at 12 months (P > 0.107; q > 0.617) or 24 months (P > 0.059; q > 0.649). See Table S5 for means and effect sizes.

Peak alpha frequency.—At 12 months, there was no significant effect of group (R(1, 8) = 1.172, P = 0.311), or region (R(2, 16) = 0.239, P = 0.790), and no group × region interaction effect (R(2, 54) = 0.822, P = 0.457). At 24 months, there was no significant effect of group (R(1, 21) = 0.248, P = 0.624). There was a significant effect of region (R(2, 42) = 3.549, P = 0.026), but no group × region interaction effect (R(2, 42) = 1.787, P = 0.180).

Alpha phase coherence.—There were no significant differences in APC between ASD+ and ASD– groups at 12 months (P > 0.002, q > 0.457). Decreased long-range interhemispheric APC (between electrodes T9 and C4) in ASD+ (M = 0.23, SD = 0.03) was found compared to ASD– infants at 24 months (M = 0.28, SD = 0.04; P = 0.00045, q =0.117, d = 1.571; See Fig. 3). Interestingly, this is a highly similar connection to that, which differentiated TSC and TD participants at 12 months. ANCOVA demonstrated that group differences in APC for this connection (T9-C4) remained significant when covarying VDQ and NVDQ (R(1, 17) = 5.54, P = 0.031). There were no differences in APC for this connection between the two sites of acquisition (P > 0.112). See Table S6 for means and effect sizes.

PAF and Cognitive Function

Kendall's Tau correlations were performed to test the relationships between PAF (at 12 and 24 months) and cognitive function at 36 months across the entire sample (i.e., TD and TSC). Twenty-four month PAF across all three regions of interest was found to be positively associated with NVDQ at 36 months, and frontal and central PAF at 24 months were found to be positively associated with VDQ at 36 months (see Table 3 and Fig. 4). These analyses were repeated within the TSC group only, in order to determine whether PAF was related to cognitive function within the clinical sample (both ASD+ and ASD–). While similar relationships were seen (see Table 3 and Fig. 4), a significant association was only seen for occipital PAF (24 months) and NVDQ (36 months).

Additional Statistical Comparisons

Although limited in scope due to the high number of participants who had tubers and seizures, the identified APC functional connections and PAF values were not associated with the presence of seizures or tubers, or the age of seizure onset (see Table S7). In addition, the identified APC functional connections at both 12 and 24 months did not show associations with PAF in any region (P > 0.08).

Discussion

Many infants with TSC show developmental delays and emerging signs of ASD by 12 months of age, with the most robust differences occurring between 12 and 24 months of age [Capal et al., 2017; Jeste et al., 2014; McDonald et al., 2017]. Recent evidence from older

children and adults with TSC suggests that widespread alterations in neural connectivity are associated with the presence of neurodevelopmental disorders in TSC [Im et al., 2016; Lewis et al., 2012; Peters et al., 2012, 2013]. Here, we report differences in network disruption (quantified using EEG measures of PAF and APC) across a clinically representative (high incidence of epilepsy and associated medications as well as cortical tuber burden) cohort of participants with TSC, thus reinforcing the value of these EEG biomarkers even in the context of extensive clinical comorbidities. Consistent with previous reports of PAF and APC measured within the same sample [Dickinson et al., 2018; Dickinson et al., 2017], PAF and cognitive function were closely associated in the present sample, whereas APC appears to represent changes between groups that go beyond differences in cognitive function.

TSC/TD Group Comparison

Using an unbiased, data-driven statistical approach, we found that infants with TSC demonstrate reduced interhemispheric APC between the left temporal and right parietal regions compared to TD infants at 12 months of age. Notably, these differences were identified despite the absence of group differences in alpha power, suggesting that phase coherence provides increased sensitivity to functional disruptions at this early time point. In addition to APC differences, we demonstrate that PAF is associated with cognitive function in TSC, which likely drives the PAF differences observed between TSC and TD infants at 24 months of age. To the extent that APC and PAF index the development of neural networks [Klimesch et al., 2007; Valdés-Hernández et al., 2010], these findings suggest that functional networks show disruption from a young age in TSC.

The present findings specifically implicate aberrant long-range functional connectivity in TSC. One of the processes crucial to the integrity of large-scale functional networks is the myelination of long-range connections, which facilitates the efficient transfer of information over distributed brain regions [Fornari, Knyazeva, Meuli, & Maeder, 2007]. The integrity of long-range structural connections is reflected in measurements of functional network properties, with oscillations in the alpha range found to be particularly reflective of changes in white matter structure [Jann et al., 2012; Picci, Gotts, & Scherf, 2016; Teipel et al., 2009; Valdés-Hernández et al., 2010]. Conversely, disorders of white matter are associated with disruptions in alpha oscillations, with decreased PAF demonstrated in vascular dementia [Moretti et al., 2004], multiple sclerosis [Leocani et al., 2000; Tewarie et al., 2014; Van der Meer et al., 2013], and white (but not gray) matter tumors [Gloor, Kalabay, & Giard, 1968; Goldensohn, 1979]. Decreased PAF and reduced APC may therefore indicate delayed or atypical maturation of white matter during the infancy period, which is consistent with the disrupted interhemispheric white matter connections that have been identified later in TSC [Baumer et al., 2017; Im et al., 2016; Krishnan et al., 2010; Peters et al., 2012].

ASD+/ASD- Group Comparison

A strikingly similar pattern (to that differentiating TSC/TD at 12 months of age) of decreased interhemispheric connectivity emerged at 24 months in ASD+, compared to ASD – infants. This pattern is consistent with the fact that ASD is largely associated with long range decreased connectivity during childhood and adulthood [Wass, 2011], and closely resembles connectivity disturbances observed in idiopathic ASD using identical recording

procedures and analyses [Dickinson et al., 2018]. However, although neurobiological mechanisms that potentially explain 12–24 changes in group differences, it should also be noted that a larger sample size at 24 months may have provided the statistical power to detect group differences that were obscured at 12 months.

The timing of differences in connectivity patterns between ASD+ and ASD- infants indicates that changes in connectivity are not static, and likely change over development, as we would expect given the dynamic interplay of genes, environment, and clinical comorbidities that affect neural function in these infants. In fact, in other high-risk infant populations, it is the change in neural connectivity over the first year of life that distinguishes infants who develop ASD from those that do not [Shen & Piven, 2017]. Moreover, in non-syndromic ASD, imaging studies have demonstrated a pattern of *under*connectivity during childhood that seems to be preceded by structural *over*-connectivity [Wolff et al., 2012] and early cortical overgrowth [Courchesne, 2004; Courchesne & Pierce, 2005; Hazlett et al., 2017] in the first months of life. Additionally, although reports of increased functional connectivity (quantified using APC) at 14 months of age in participants who later receive an ASD diagnosis [Orekhova et al., 2014] have not been replicated at a group level, APC at 14 months is associated with the severity of later restricted and repetitive behaviors [Haartsen, Jones, Orekhova, Charman, & Johnson, 2019].

Therefore, it is possible that the lack of significant connectivity differences at 12 months may reflect a transitional period in the infants who go on to develop ASD (see Fig. 5). While robust longitudinal analyses were not possible in the present data, qualitative inspection of ASD+ and ASD- infants who have both 12- and 24-month EEG data supports that APC between 12 and 24 months increases in ASD- infants, but decreases in ASD+ infants (see Fig. S2). In nonsyndromic ASD, several neurobiological mechanisms have been attributed to underlie the atypical "crossover" trajectory captured in ASD, including differences in axonal pruning and myelination [Wolff et al., 2015; Wolff et al., 2012], differences in large scale anatomy [Courchesne, 2004; Hazlett et al., 2017], and a complex interplay between gray and white matter changes [Bos et al., 2015; Ecker et al., 2016]. It should be noted that this is the first study to examine such change in neural function in TSC, and it supports the need for longitudinal investigation in neurodevelopmental disorders, regardless of underlying genetic etiology. Even though the impact of the TSC mutation impacts brain structure and function before birth, there are ongoing changes in the functional brain network that may be able to be modulated by early interventions (both pharmacological and behavioral). For instance, pharmacological intervention using an mTOR inhibitor, everolimus, appears to improve the microstructural integrity of the corpus callosum in patients with TSC [Peters et al., 2019; Tillema, Leach, Krueger, & Franz, 2012].

Cognition and PAF

Finally, we asked whether early measures of network function (PAF) related to individual differences in later clinically meaningful dimensional outcomes, which we quantified using measures of cognitive function. We found that across the entire sample, 24 month PAF was associated with cognitive functioning measured at a later timepoint (36 months). This is consistent with cross sectional work demonstrating that PAF is related to cognitive function

in children with ASD [Dickinson et al., 2017], and neurotypical adults [Grandy et al., 2013; Klimesch, Schimke, & Pfurtscheller, 1993; Richard Clark et al., 2004], and extends these findings to demonstrate that PAF also holds value as an early prognostic marker related to later developmental outcomes. Restricting the sample to just TSC infants, similar associations between 24 month PAF and 36 month NVDQ were observed, highlighting the potential for PAF to stratify individuals within the TSC group. However, it should be noted that the limited sample size in the present study precluded within-subgroup TSC analyses (ASD+, ASD–). Furthermore, the TD group in the present study demonstrates above average NVDQ and VDQ, indicating that they do not adequately represent the range of cognitive function seen during typical development. Studying larger and more representative samples of these subgroups in the future will allow us to determine whether the relationship between PAF and later cognition extends from ASD, to wider non-ASD developmental outcomes.

Limitations

Structural lesions such as cortical and subcortical hamartomas (tubers) may alter EEG measures of functional connectivity through volume conduction [Nunez & Srinivasan, 2006; Nunez et al., 1997]. We applied surface Laplacian transform in an attempt to mitigate the effects of tubers [Kayser & Tenke, 2015]. The surface Laplacian isolates the source activity under each electrode. Therefore, CSD estimates capture the unique properties of each electrode while minimizing activity that is broadly distributed across multiple electrodes thereby reducing individual differences in volume conduction introduced through cortical differences, including tuber burden. The special consideration given to this specific signal processing issue resulted in the detection of individual differences in subgroups of TSC infants (ASD+/ASD–), two groups equivalently impacted by tuber burden. Additional controls for possible EEG differences contributed by behavioral differences during the recording session could be provided through a concurrent analysis of each participants behavior, as we have implemented in more recent data collection procedures [DiStefano, Dickinson, Baker, & Jeste, 2019].

The presence of interictal EEG abnormalities in infants with epilepsy also can alter the EEG signal being measured [Schevon et al., 2007]. However, the presence of seizures is more often associated with EEG hypersynchrony [Mormann, Lehnertz, David, & Elger, 2000; Schevon et al., 2007], which would manifest as increased phase coherence. Therefore, the fact that TSC participants in the present study show decreased phase coherence would suggest that these differences are not spuriously introduced through the presence of seizures. In addition, our goal was to include a clinically representative sample of infants with TSC, and the findings reinforce the fact that EEG biomarkers in early infancy, as in other high-risk populations, hold potential value for early risk stratification, prediction, and as possible biomarkers of change with intervention. However, in future studies we aim to fully explore the potential effect of AED's and seizure severity on the EEG measures collected here, which will be facilitated through larger samples.

Future Directions: APC and PAF as Markers of Altered Neural Connectivity in TSC

The present results indicate that both APC and PAF may hold utility in early life in TSC. Both of these metrics not only are able to differentiate TSC from TD infants, but are also

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sensitive to differences in infants who later develop ASD, and show early differences according to later cognitive function. APC and PAF have been previously associated with measures of structural network development, particularly white matter connections. In addition, a large literature of child and adult studies in TSC find disruptions in white matter integrity. Considering both of these bodies of literature, altered PAF and APC in the present study may be capturing early divergences in white matter development.

The demonstration that functional connectivity deficits in TSC can be measured as early as 12 months and relate to neurodevelopmental outcomes suggests that EEG could serve as a robust tool not only to predict seizures (which is already being done clinically) but also to inform our understanding of neurodevelopmental disorders in TSC. Future research is needed to examine functional EEG measures alongside white matter maturation in TSC in infancy in order to better understand the neurobiological mechanisms underlying early network disruption. Studying infants from earlier in life will also allow us to determine whether TSC ASD+ infants show similar patterns of long-range hyperconnectivity to those seen during the first months of life in infants at familial risk for ASD. The in utero diagnosis of TSC certainly facilitates this research avenue. However, due to the rare nature of TSC, and further reduction of sample sizes that occurs when comparing TSC infants based on ASD outcomes, sufficiently large samples will only be achieved through multisite studies.

Conclusions

Here, we find that EEG measures of brain network function are altered in TSC and those infants with TSC who develop ASD in the first 2 years of life. Specifically, we find (a) decreased long-range coherence in TSC at 12 months and lower PAF at 24 months, (b) decreased long-range coherence at 12 months in infants who develop ASD, and (c) an association between PAF (measured at 24 months) and cognitive function (measured at age 3). The identification of objective, quantifiable markers of atypical development within the first years of life in TSC can improve the timing and accuracy of identification of those infants requiring additional, targeted developmental interventions from very early in life. The availability of scalable objective markers of altered neural development is particularly advantageous during infancy, as this is a period of peak neural plasticity during which developmental trajectories could potentially be steered. Thus, while a formidable challenge, particularly in the context of a rare disorder marked by high heterogeneity, this early detection approach is highly warranted considering the potential clinical and translational applications associated with the identification of objective (and potentially, prognostic), biological markers of neurodevelopmental outcomes.

Supplementary Material

Refer to Web version on PubMed Central for supplementary material.

Acknowledgments

The authors wish to thank all of the infants and families who generously participated in the study. The present research was supported by grants from the Department of Defense (DOD CDMRP TSCRP: 2011–2014) and the UCLA CTRC (UL1TR000124). The authors are also grateful to Scott Huberty and Lauren Baczewski for their role in the study, and to Manjari Daniel for her helpful comments on the manuscript.

S.S.J. serves as a consultant for Roche Pharmaceuticals and on the professional advisory board for the Tuberous Sclerosis Alliance. M.S. receives research support from Novartis, Roche, Pfizer, LAM Therapeutics and Quadrant Biosciences, and has served on the Scientific Advisory Board of Sage Therapeutics, Roche, Celegne and Takeda.x

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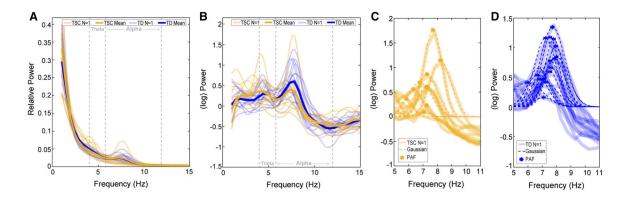


Figure 1.

Example of the PAF fitting routine across all 12-month data (orange = TSC, blue = TD). (A) Individual (faded) and group average (bold) power spectra highlight the difficulty of extracting a clear alpha peak from raw spectra. (B) After the removal of the 1/f trend, clearly separable modulations occurring in the alpha (6–12 Hz), and theta (4–6 Hz) ranges can be identified. (C) Gaussian peak fits for TSC, and (D) TD participants.

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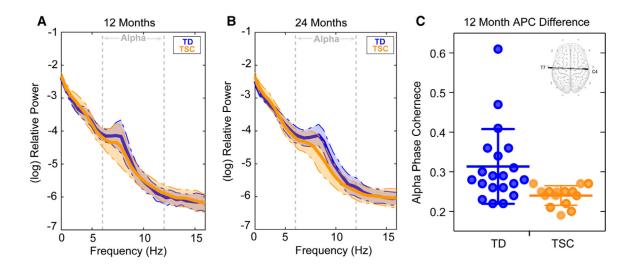


Figure 2.

This figure illustrates TSC and TD infants. (A) 12 month and (B) 24 months logtransformed power spectral density (average of all 25 channels) for both TSC (orange) and TD participants (blue). Shaded regions represent the median absolute deviation. (C) Dot plot demonstrating APC for electrode pair T7-C4 (with group mean and error bars representing *SD*) for both TD and TSC groups (the group difference remained significant following removal of the two TD participants with the highest APC, P < 0.002). The scalp map of the interhemispheric connection that differentiated TD from TSC infants at 12 months is depicted (top right).

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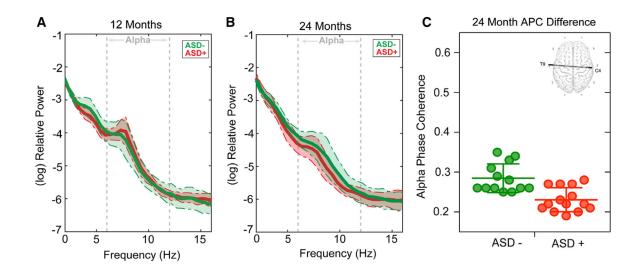


Figure 3.

This figure illustrates ASD+ and ASD– infants. (A) 12 month and (B) 24 months logtransformed power spectral density (average of all 25 channels) for both TSC (orange) and TD participants (green). Shaded regions represent the median absolute deviation. (C) Dot plot demonstrating APC for electrode pair T9-C4 (with group mean and error bars representing SD) for both ASD+ and ASD– groups. The scalp map of the interhemispheric connection that differentiated ASD+ from ASD– infants at 24 months is depicted (top right).

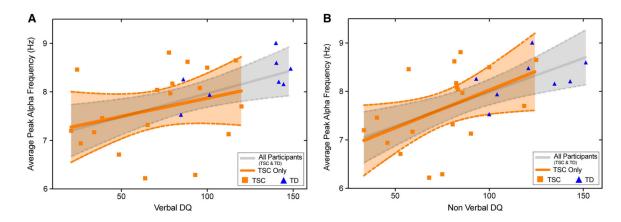


Figure 4.

Scatter plots demonstrating the relationship between 24-month average PAF (averaged over frontal; central and occipital regions) and 36 months (**A**) VDQ, and (**B**) NVDQ. Plots represent the association (and 95% confidence intervals) for both the entire sample (gray), and restricted to TSC participants (orange).

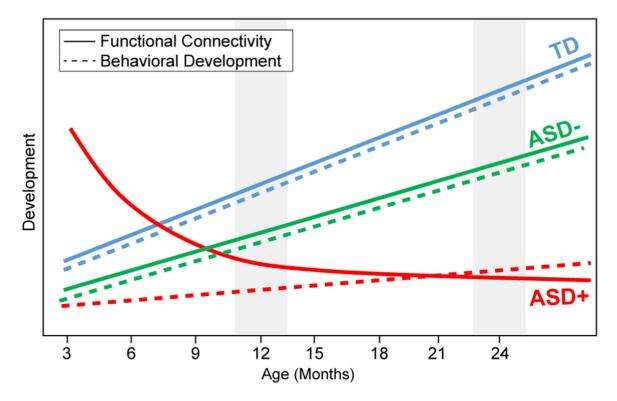


Figure 5.

Theoretical (simplified) trajectories of functional connectivity and behavioral development across the first 2 years of life, with gray patches indicating timing of present study. Crucially, the hypothetical trajectories suggested here would elucidate why behavioral measures may discriminate ASD and no ASD in TSC at 12 months of age, but EEG measures do not.

	TD Control $(n = 20)$	TSC $(n = 35)$	ASD- $(n = 14)$ ASD+ $(n = 17)$	$ASD+\left(n=17\right)$
Sex				
Number of female (% female)	11 female (55%)	9 female (26%)	9 female (26%) 4 female (29%) 4 female (24%)	4 female (24%)
12 month EEG				
Number completed (percentage completed).	$n = 20 \; (100\%)$	n = 18 (51%)	n = 7 (50%)	<i>n</i> = 8 (47%)
24 month EEG				
Number completed (percentaage completed).	$n = 12 \ (60\%)$	$n = 28 \ (80\%)$	n = 13 (93%)	<i>n</i> = 14 (82%)
36 months verbal DQ	115.20 (30.68)	68.92 (30.38)	82.74 (22.57)	47.42 (29.25)
Mean (SD) ; Number completed (percentage completed)	$n = 10 \; (50\%)$	<i>n</i> = 23 (66%)	$n = 14 \ (100\%)$	n = 9 (53%)
36 months nonverbal DQ	113.87 (24.34)	73.32 (24.27)	84.40 (20.78)	56.09 (19.24)
Mean (SD); Number completed (percentage completed)	n = 10~(50%)	<i>n</i> = 23 (66%)	$n = 14 \ (100\%)$	n = 9 (53%)
ASD outcome				
Mean calibrated severity score (<i>SD</i>); Number completed (percentage completed)	1.05 (0.52) n = 19 (95%)	4.3 (2.77) n = 30 (86%)	$\begin{array}{c} 1.71 \; (0.82) \\ n = 14 \; (100\%) \end{array}$	6.24 (2.05) n = 17 (100%)

Note. Overall N(top row) indicates the total number of unique participants in each group.

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Peak Alpha Frequency Values and Group Comparison Results

12-months	TD Mean (SD)	TSC Mean (SD)	TSC Mean (SD) Test statistic (U) P-value Cohen's d	<i>P</i> -value ^{<i>a</i>}	Cohen's d
Frontal	7.42 (0.53) n = 18	7.42 (0.53) n = 18 7.16 (0.59) n = 16	U=102	0.154	0.464
Central	7.49 (0.47) n = 19	7.36(0.48) n = 16	U = 124.5	0.367	0.274
Occipital	7.25 (0.58) n = 18	6.9 (0.63) n = 16	U = 102.5	0.154	0.580
24-months	Mean (SD)	Mean (SD)			
Frontal	8.31 (0.39) n = 12	8.31 (0.39) n = 12 7.71 (0.71) n = 27	<i>U</i> =75	0.007 ^a	1.047
Central	8.42 (0.44) n = 12	7.8 (0.87) n = 27	U=86	0.026 ^a	0.899
Occipital	8.08 (0.73) n = 12	8.08 (0.73) $n = 12$ 7.56 (0.91) $n = 26$	U = 112.5	0.174	0.630

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Table 3.

Associations Between Peak Alpha Frequency and Cognitive Function (Measured at 36 Months) Across All Participants

	All Participants (TD and TSC)	(TD and TSC)	TSC Gro	TSC Group Only
	VDQt(P)	NVDQ $t(P)$	VDQt(P)	NVDQ $t(P)$
12 months				
Frontal	0.013~(0.940)	-0.026 (0.879)	0.028 (0.917)	-0.028 (0.917)
Central	0.141 (0.401)	0.211 (0.208)	-0.135 (0.590)	0.045 (0.857)
Occipital	0.199 (0.234)	0.258 (0.123)	0.141 (0.600)	0.423 (0.116)
24 months				
Frontal	$0.334 (0.013)^{a}$	0.420 (0.002) ^a	0.159 (0.330)	0.307 (0.060)
Central	$0.426 \left(< 0.001\right)^{a}$	0.459 (0.001) ^a	0.259 (0.112)	0.301 (0.064)
Occipital	0.236 (0.079)	0.314 (0.020) ^{<i>a</i>}	0.242 (0.136)	0.389 (0.016) ^a

¹Represents associations that remained significant after FDR correction.