A Network of Amygdala Connections Predict Individual Differences in Trait Anxiety

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Abstract: In this study we demonstrate that the pattern of an amygdala-centric network contributes to individual differences in trait anxiety. Individual differences in trait anxiety were predicted using maximum likelihood estimates of amygdala structural connectivity to multiple brain targets derived from diffusion-tensor imaging (DTI) and probabilistic tractography on 72 participants. The prediction was performed using a stratified sixfold cross validation procedure using a regularized least square regression model. The analysis revealed a reliable network of regions predicting individual differences in trait anxiety. Higher trait anxiety was associated with stronger connections between the amygdala and dorsal anterior cingulate cortex, an area implicated in the generation of emotional reactions, and inferior temporal gyrus and paracentral lobule, areas associated with perceptual and sensory processing. In contrast, higher trait anxiety was associated with weaker connections between amygdala and regions implicated in extinction learning such as medial orbitofrontal cortex, and memory encoding and environmental context recognition, including posterior cingulate cortex and parahippocampal gyrus. Thus, trait anxiety is not only associated with reduced amygdala connectivity with prefrontal areas associated with emotion modulation, but also enhanced connectivity with sensory areas. This work provides novel anatomical insight into...
potentials mechanisms behind information processing biases observed in disorders of emotion. Hum Brain Mapp 00:000–000, 2015. © 2015 Wiley Periodicals, Inc.

Key words: amygdala; trait anxiety; diffusion tensor imaging; emotion regulation; probabilistic tractography; medial prefrontal cortex; extinction

INTRODUCTION

Anxiety-related disorders are the most prevalent mental illnesses [Kessler et al., 2005a, 2005b], and high trait anxiety is associated with increased risk for numerous mental disorders, including depression and bipolar disorder [Bruckl et al., 2007; Reinherz et al., 2000]. Neurocognitive models of anxiety highlight the importance of the amygdala [Davis, 1992; Rauch et al., 2003], and interactions with regions important for cognitive control, as well as emotion generation, regulation, and perception [Bishop, 2007; Milad and Quirk, 2012]. Despite its relevance to affective disorders, little is known about the relationship between trait anxiety and the integrity of structural connections between the amygdala and these systems.

Functionally, individual differences in trait anxiety are negatively correlated with aspects of the ventromedial prefrontal cortex, including mOFC activity, and positively correlated with amygdala activity during fear modulation [Indovina et al., 2011]. Furthermore, functional connectivity between the amygdala and mOFC is negatively related to temperamental anxiety [Kim et al., 2011; Pezawas et al., 2005]. Thus, robust amygdala-mOFC connectivity may be protective for anxiety. Conversely, activation of dACC appears involved in the acquisition of threat-related learning [Phelps et al., 2004] and anxiety [Kim et al., 2011; Milad et al., 2009]. In addition to connections between amygdala and prefrontal cortex, research has found that enhanced functional connectivity between amygdala and perceptual regions during fear generalization is positively correlated with trait anxiety [Dunsmoor et al., 2011]. One possibility is that high trait anxiety is associated with enhanced structural connectivity between amygdala and regions involved in perceptual and semantic processing.

The connective architecture of the brain plays a key role in determining a region’s functional attributes, and functional connectivity is often assumed to reflect, at least in part, the underlying anatomical connectivity. While there is evidence suggesting that a correspondence often exists between the strength of functional connectivity and structural pathways in a number of networks [Greicius et al., 2009; Hermundstad et al., 2013, 2014], evidence of functional connectivity can be observed in the absence of anatomical connectivity [Honey et al., 2009]. Although there are studies relating anxiety to functional connectivity, far less is known about the relationship between measures of structural white matter and individual differences in anxiety. Kim and Whalen [2009] demonstrated using DTI that fractional anisotropy (FA; a measure of white-matter microstructure) in a region of putative uncinate fasciculus is negatively correlated with trait anxiety. More recently, in a large multi-cohort study, Westlye et al. [2011] found that FA throughout much of the white-matter skeleton was negatively correlated with harm avoidance (a personality trait characterized by heightened worrying and anxiety). Despite these contributions, it remains unknown whether a more distributed pattern of structural connections between the amygdala in particular and other neural regions combine to contribute to individual differences in trait anxiety. Moreover, no studies to date have examined which amygdala structural connections make positive contributions to anxious traits.

To address these unknowns, the present study combined a multiple regression analysis using regularized least square regression model, ridge regression [Hoerl and Kennard, 1970], with maximum likelihood estimates of structural connectivity of the amygdala using seed-based probabilistic tractography [Behrens et al., 2003, 2007]. This powerful approach allows for the identification of structural connections between the amygdala and multiple brain regions in a whole-brain manner [Saygin et al., 2011, 2012]. After first ensuring that the pattern of amygdala structural connectivity could be used to predict individual differences in trait anxiety significantly above chance, we sought to determine which connections were most reliably included in the prediction model. Thus, we tested the intriguing possibility that trait anxiety would not only be associated with reduced connectivity between the amygdala and emotion control regions like mOFC, but that it would also be associated with both enhanced structural connectivity between amygdala and dACC, as well as between the amygdala and sensory cortical pathways responsible for driving perceptual and semantic representations of emotional reactivity and sensation.

METHODS

Participants were seventy-two healthy right-handed participants [mean age = 25.5 ± 6.5 (SD), 41 females and 31 males; no significant difference in age between genders (P > 0.5)]. Trait anxiety scores were obtained for each participant via the State-Trait Anxiety Inventory (STAI) [Spielberger, 1983] [mean = 30.6 ± 7.7 (SD), range 21–63; Mwomen = 29.8 ± 6.4 (SD), Mmen = 31.6 ± 9.2 (SD); no significant difference in trait anxiety between genders (P > 0.3)]. Each subject completed a diffusion-tensor imaging
 segment covered the whole brain (repetition time-and echo time 75 ms. The high-resolution T1-weighted
tricles for each amygdala for each subject [Saygin et al., plus the brainstem while avoiding a mask of the ven-
tribution of connecting tracts to each ipsilateral target
amygdala seed region, which produced a frequency dis-
25,000 sample tracts were drawn from each voxel in the
space using FSL-FDT [Behrens et al., 2007] in which
was carried out in each participant’s native diffusion
control. Prior to probabilistic tractography, the DTI data
were eddy current corrected, skull stripped, and the
subcortical regions and one bilateral region (brain-
ind images of each participant. This produced an indi-
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plus the brainstem while avoiding a mask of the ven-
ticles for each amygdala for each subject [Saygin et al.,
2011, 2012]. Each resulting amygdala image was trans-
moved to MNI space to determine the group-wise prob-
ability density for each amygdala-target pair. In this
manner we identified the group-wise peak voxel for
each amygdala-target pair, and placed a 6 mm radius
spherical mask centered on the peak voxel (see Suppor-
ting Information Table I for MNI coordinates of the peak
voxel for each amygdala-target pairing). Once each
spherical group mask was obtained they were reverse
transformed to each individual’s native space using
nearest-neighbor registration. In this manner we could
determine the maximum likelihood of amygdala-target
connectivity by taking the value of the amygdala voxel
with the largest number of sample tracts connecting to
the respective target and dividing by 25,000 (producing
a scaled value between 0 and 1). Thus, the maximum
likelihood estimate was determined for each of the 86
amygdala-target combinations, 43 pathways per hemi-
sphere, for each of the 72 participants (observations).
These values from both right and left amygdala were
combined, which produced a matrix of 72 observations
by 86 features for performing multiple regression. We
elected to focus exclusively on ipsilateral connections
in order to reduce the number of features included in a
single full model and because tracer studies using non-
human primates find that first order amygdala connec-
tion are largely ipsilateral [Ghashghaei and Barbas, 2002;
Ghashghaei et al., 2007]. Gender was included as an
additional feature in the model given evidence from
meta-analyses of functional imaging data suggesting
possible gender differences in amygdala activity to emo-
tional stimuli [Sergerie et al., 2008; Wager et al., 2003].
Notably, the exclusion of gender in the analysis
described below had no effect on the prediction accuracy
of the model, nor did the feature corresponding to gen-
der make a reliable contribution to a full model.

**Segmentation and Probabilistic Tractography**

Relevant to the current study and our focus on the
amygdala and individual differences, the validity and
utility of probabilistic tractography has recently been
demonstrated in two studies of amygdala segmentation
[Bach et al., 2011; Saygin et al., 2011], and another study
predicting individual differences in functional activation
in fusiform gyrus [Saygin et al., 2012]. The primary anal-
ysis was performed after combining the ipsilateral con-
nectivity estimates for both the right and left amygdala.
The method for determining the connectivity estimates
is described below and summarized in Figure 1. Seed
and target masks were defined using the cortical and
subcortical automated segmentation tools in FreeSurfer
[Fischl et al., 2002, 2004] using the T1-weighted anatom-
ical images of each participant. This produced an indi-
vidually derived right and left amygdala seed mask and
85 target masks, which included 84 unilateral cortical
and subcortical regions and one bilateral region (brain-
stem), all of which were visually inspected for quality
control. Prior to probabilistic tractography, the DTI data
were eddy current corrected, skull stripped, and the
principal diffusion directions of each voxel were deter-
mined [Behrens et al., 2007]. Probabilistic tractography
was carried out in each participant’s native diffusion
space using FSL-FDT [Behrens et al., 2007] in which
25,000 sample tracts were drawn from each voxel in the
amygdala seed region, which produced a frequency dis-
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Notably, the exclusion of gender in the analysis
described below had no effect on the prediction accuracy
of the model, nor did the feature corresponding to gen-
der make a reliable contribution to a full model.

**Data Acquisition**

Diffusion-tensor and T1-weighted imaging was per-
formed on a 3-Tesla Siemens MRI scanner with a 32-
channel head coil at Robarts Research Institute, University
of Western Ontario. DTI images were acquired in the axial
plane with echo-planar imaging consisting of 55 slices,
2.1 × 2.1 × 2 mm voxels, 200 × 200 mm field of view, 96
× 96 mm base resolution, 65 isotropically weighted diffu-
sion directions, b-value = 700 s/mm², repetition time 6 s,
and echo time 75 ms. The high-resolution T1-weighted
anatomical scan covered the whole brain (repetition time-
= 2,300 ms, echo time = 4.25 ms; field of view = 25.6 cm;
192 slices; 1 mm³ isovoxels; 256 × 256 matrix).

**Multiple Regression analysis**

Multiple linear regression was implemented using the ridge regression model from the scikit-learn toolbox in python [Pedregosa et al., 2011]. Ridge regression is a regularized regression model and was selected for use in the current study because it is robust to instances in which the number of features is greater than the number of observations and can be suitable for dealing with instances of collinearity between features [Hoerl and Kennard, 1970]. Thus, regularization identifies a subset of features important to the model and enhances the magnitude of their respective weights while lowering the weights of unimportant features. While other sparse regression techniques like LASSO are suitable for dealing with high dimensional datasets, it does not perform as well with collinearity between features [Zou and Hastie, 2005]. The regularization parameter, alpha, was determined using nested cross-validation with an auto-
mated internal cross-validation function, ridgeCV, which performs leave-one-out cross validation using only the training set for determining alpha. We performed the

**Amygdala Connectivity and Trait Anxiety**

(DTI) scan. All participants were in good health, and had
no history of psychiatric illness, neurological disease, or
head injury as determined by screening and interviews
using the Structured Clinical Interview of the DSM-IV
[First et al., 2002]. All participants provided informed
written consent, and the study was approved by the
research ethics board of the University of Western
Ontario.
regularization parameter estimate by considering a vector of alpha values from 0.01 to 10 in steps of 0.01. The resulting mean alpha was 6.22, which was also used in the permutation testing. In order to assess whether the regression model could significantly predict trait anxiety scores we used a stratified six-fold cross-validation approach with nested-cross-validation for regularization parameter estimation (see Fig. 1 for a schematic). Similar to previous application of sparse regression in neuroimaging [Wager et al., 2011], a stratified sixfold approach was selected as it tends to have a prediction accuracy biased towards zero (i.e., is more conservative), and produces more consistent (i.e., less variable) results relative to leave-one-out cross-validation [Hastie et al., 2009; Kohavi, 1995]. We performed six iterations in which we split the data, trained the regression model on 60 participants and tested on the remaining 12 participants (see Fig. 1). The data was split such that no participant was ever included in the training and testing set simultaneously, no participant was in more than one of the six test sets, and there was a similarly distributed range of trait anxiety values at each fold. Concatenating the 12 predicted anxiety scores from each fold produced a vector of 72 predicted anxiety scores, one predicted score for each participant. The accuracy of the model was derived by computing the MSE between the 72 predicted trait-anxiety scores and actual trait-anxiety scores provided by self-report. [Color figure can be viewed in the online issue, which is available at wileyonlinelibrary.com.]

Figure 1.

Schematic of the multiple regression approach: After deriving a group-wise peak voxel for each amygdala-target pair (TOP), participants were split using a stratified sixfold cross validation approach (BOTTOM), producing six independent training (n = 60) and testing (n = 12) sets. The regression model was trained on the scaled values of the maximum likelihood of amygdala-target connectivity derived from probabilistic tractography (between 0 and 1, represented as the gray-scaled boxes). Testing the model at each fold produced predicted trait-anxiety scores for 12 participants. The accuracy of the model was derived by computing the MSE between the 72 predicted trait-anxiety scores and actual trait-anxiety scores provided by self-report. [Color figure can be viewed in the online issue, which is available at wileyonlinelibrary.com.]
randomly pairing observations with a trait anxiety score. The \( P \) value was determined as the proportion of iterations in which a model generated on the randomized data outperformed or was equal to the model generated on the real data. We also assess how well the predicted anxiety scores correlated with participant’s self-report trait anxiety using Pearson’s correlation \((r)\), and the proportion of variance accounted for by the regression model \((r^2)\). Indeed, estimation of correlation effect sizes in this manner is known to produce more robust estimates [Hastie et al., 2009].

Full-Model Estimation: In order to determine which features (i.e., which amygdala-target anatomical connections) made a reliable contribution to a full-model for predicting individual differences in anxiety we adapted the bootstrap procedure used by Wager et al. [2011]. We performed 1,000 bootstrapped samples with replacement, which produced training sets of 60 observations. In this fashion, we derived 1,000 independent models and their respective weights. To assess the reliability of each amygdala-target connection to predicting trait anxiety we determine which features had 95% confidence intervals which were either entirely above or below zero. The median coefficient and confidence intervals for each reliable amygdala-target pathway and the confidence intervals are reported in Table I. An estimation of which amygdala-target connections made the most reliable contribution to the full-model is extremely relevant to the neuroscience of anxiety, though it must be noted that predictions are necessarily made from the combination of all weights and are therefore not strictly independent [Wager et al., 2011].

**RESULTS**

**Model Testing**

The ridge model (mean \( z = 6.22 \)) trained on the maximum likelihood of amygdala-target connectivity for both right and left amygdala, as well as the gender of participants, performed significantly better than chance at predicting trait anxiety (\( P < 0.01 \), one-tailed, Fig. 2). While the MSE of trait anxiety estimates was 55.01, the mean MSE from the randomized permutation samples was 60.88. The Pearson’s correlation between the predicted trait anxiety and participants self-reported trait anxiety was also significant (\( r = 0.246, P < 0.05 \)), indicating that our model of amygdala connectivity to a network of regions accounted for 6% of the variance in trait anxiety (see Fig. 2).

**Full-Model Estimation**

Given that the model could predict trait anxiety significantly better than chance from probabilistic connectivity profiles of amygdala-target pathways of right and left amygdala, a critical question was which pathways make the most reliable contribution to a full-model (see Table I for full results and Fig. 3). Whereas those connections resulting in reliably positive weights suggest that stronger anatomical connectivity between the amygdala and those regions is associated with elevated trait anxiety, reliably negative weights indicates that stronger anatomical connectivity is associated with reduced trait anxiety. In line with our hypotheses, our full-model estimation revealed that the reliable amygdala pathways with positive weights included neural regions associated with the expression of anxious behavior, right dACC, as well as those associated with perception, semantic representation, and sensation, including right ITG, right temporal pole, right entorhinal cortex, and left paracentral lobule. Those reliable amygdala pathways with negative weights included regions associated with the extinction learning, right mOFC, as well as those regions associated with memory encoding and environmental context recognition, left isthmus of the cingulate cortex and left parahippocampal gyrus.

### TABLE I. Targets of structural connectivity with the amygdala that make a reliable contribution to the full-model for predicting trait-anxiety

<table>
<thead>
<tr>
<th>Target region</th>
<th>Coefficient</th>
<th>C.I.</th>
<th>Peak sphere coordinate</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Targets making a reliably positive contribution to predicting anxiety</strong></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>L Paracentral lobule</td>
<td>577.40</td>
<td>141.75</td>
<td>1550.36</td>
</tr>
<tr>
<td>R Dorsal anterior cingulate cortex</td>
<td>42.58</td>
<td>4.57</td>
<td>146.12</td>
</tr>
<tr>
<td>R Caudate nucleus</td>
<td>3.38</td>
<td>0.43</td>
<td>7.45</td>
</tr>
<tr>
<td>R Inferior temporal gyrus</td>
<td>0.94</td>
<td>0.07</td>
<td>2.46</td>
</tr>
<tr>
<td>R Entorhinal cortex</td>
<td>0.74</td>
<td>0.14</td>
<td>1.24</td>
</tr>
<tr>
<td>R Temporal pole</td>
<td>1.00</td>
<td>0.35</td>
<td>1.87</td>
</tr>
<tr>
<td><strong>Targets making a reliably negative contribution to predicting anxiety</strong></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>L Posterior cingulate cortex</td>
<td>-2.64</td>
<td>-6.79</td>
<td>-0.86</td>
</tr>
<tr>
<td>R Medial orbital frontal cortex</td>
<td>-1.11</td>
<td>-2.25</td>
<td>-0.06</td>
</tr>
<tr>
<td>L Parahippocampal gyrus</td>
<td>-0.70</td>
<td>-1.40</td>
<td>-0.003</td>
</tr>
</tbody>
</table>
DISCUSSION

Using maximum likelihood structural connectivity estimates derived from probabilistic tractography, the current study is the first to demonstrate that the pattern of amygdala structural connectivity is predictive of a significant portion of the variance associated with individual differences in trait anxiety. Furthermore, consistent with our prediction and relevant to the understanding of neurocognitive models of anxiety and emotions more generally, the current results revealed that a reliable network of multiple structural pathways connected with the amygdala contributes to individual differences in trait anxiety. Higher trait anxiety was related to stronger connections between the amygdala and target regions implicated in affect generation, dACC [Milad and Quirk, 2012], and target regions involved in perceptual and semantic processing, including right ITG, right temporal pole, right entorhinal cortex, and left paracentral gyrus [Murray, 2007; Murray et al., 2007]. Conversely, higher trait anxiety was associated with weaker connections between the amygdala and target regions implicated in extinction learning, such as right mOFC [Milad and Quirk, 2012], and those involved in memory and visuospatial processing, including right posterior cingulate cortex [Parvizi et al., 2006; Vogt et al., 1992, 2006] and parahippocampal gyrus [Epstein and Kanwisher, 1998; Epstein and Higgins, 2007]. Although prior studies have identified pathways that show a negative relationship with trait anxiety, our study is the first to identify pathways wherein stronger connectivity is predictive of higher individual differences in trait anxiety.

Connections Making a Positive Contribution to Trait Anxiety

The present study is the first to demonstrate that stronger structural connections between amygdala and dACC are related to higher levels of trait anxiety. This supports findings from both animal and human research implicating dACC and its direct projections to the amygdala in the generation and persistence of fear-related learning [Milad and Quirk, 2012; Quirk et al., 2006]. Using human fMRI, Phelps et al. [2004] found greater dACC during the acquisition of fear conditioning, while Milad et al. [2009] found greater activation of dACC and impaired extinction recall in individuals with post-traumatic stress disorder. Additionally, Milad et al. [2007] found that the cortical thickness of dACC was positively correlated with emotional reactivity, as measured using skin conductance activity to fear-conditioned versus safe stimuli. Indeed, a pathway involving the dACC and the amygdala has been referred to as an “aversive amplification circuit,” that drives harm-avoidant behaviors [Robinson et al., 2014]. Thus, together with previous functional and anatomical research, the present findings indicate that stronger connections between dACC and amygdala contribute to increased trait anxiety.

We also observed that the strength of amygdala connectivity to regions associated with perception were positively associated with trait anxiety. The target regions for these pathways included ITG, temporal pole, entorhinal cortex, and paracentral lobule. Human lesion studies have demonstrated that the temporal pole and parts of the anterior
temporal lobe are necessary for facets of emotion perception, such as emotional prosody recognition [Adolphs et al., 2002]. Along with such findings, the current results suggest that enhanced connectivity between the amygdala and perceptual and sensory regions is related to increased sensitivity to threat-related information [Murray, 2007]. This is also consistent with functional imaging studies demonstrating that emotional face processing is associated with enhanced functional connectivity between amygdala and regions of the occipitotemporal cortex [Morris et al., 1999; Pessoa et al., 2002]. Furthermore, our findings are also in accordance with clinical studies that find that interactions between the amygdala and perceptual cortices are positively associated with anxiety disorders [Ahs et al., 2009; Gilboa et al., 2004]. In addition to visual perception regions, expression of feelings during an emotion induction paradigm recruits activity in somatosensory regions, including aspects of the paracentral lobule [Saxbe et al., 2013]. Stronger connections between the amygdala and these regions were also associated with increased trait anxiety in the current study.

It is interesting to consider the above results in the context of recent models of emotion perception, and cognitive biases.
in affective disorders. It is thought that excitatory interactions between the amygdala and sensory regions enhance representations of emotional stimuli [Anderson and Phelps, 2001; Mitchell and Greening, 2012; Vuilleumier and Driver, 2007]. In affective disorders, however, emotional thoughts or stimuli disproportionately affect cognition, producing “efficient but maladaptive” [Beck, 2008; p 971] information processing biases [e.g., Greening et al., 2013, 2014]. Thus, one possibility is that in highly anxious individuals persistent functional connectivity strengthens anatomical connections between the amygdala and brain regions involved in both affect production and emotion perception, possibly via Hebbian mechanisms. In this manner, emotions exert an even stronger influence on sensory processes, cognition, and behavior. Alternatively, the strength of such connections may represent an inherited trait, predisposing affected individuals to increased risk of psychopathology. These strengthened connections may pose a particular challenge for attempts to regulate unwanted emotional reactions. For example, it was recently demonstrated that, despite attempts to down-regulate negative effect, depressed patients had persistently elevated activity in the amygdala and sensory areas, despite activating regions associated with emotional control [Greening et al., 2014]. Thus, our results, and recent experimental evidence, provide novel insight into the importance of enhanced amygdala connectivity to parts of the prefrontal cortex and the temporal and somatosensory regions, the latter of which are pathways that have been overlooked in current neurocognitive models of psychopathology.

The current findings also have potentially important implications for theories related to emotion regulation and the management of affective disorders. For example, previous research has demonstrated that affective encoding of emotional stimuli can be down-regulated indirectly through attentional mechanisms [Mitchell et al., 2007; Pessoa et al., 2002]. It is thought that this occurs because attention augments the representation of non-emotional task-relevant information in occipitotemporal cortices, thereby leading, through competitive interactions, to the suppression of unwanted, possibly pathological, emotional representations [Blair and Mitchell, 2009; Mitchell, 2011; Ochsner et al., 2012]. In such models, top-down attentional mechanisms are thought to compete with pathways associated with emotional attention (amygdala-sensory cortex connections) to determine the relative impact of affective stimuli and representations. In line with this idea, our finding show that increased anatomical connectivity in this latter pathway is associated with higher levels of trait anxiety.

Connections Making Negative Contribution to Trait Anxiety

We observed that the strength of connectivity between the amygdala and mOFC was negatively related to trait anxiety. This is consistent with findings in both rodents and humans suggesting that the mOFC regulates amygdala output during extinction learning [Milad and Quirk, 2012]. Recent neuroimaging studies suggest an inverse relationship between activity in the amygdala and medial prefrontal cortex, including mOFC, during the modulation of fear-related stimuli [Amting et al., 2010; Linnman et al., 2012]. This pattern is impaired in patients with anxiety disorders [Etkin et al., 2010; Shin et al., 2005]. Intriguingly, a recent resting state connectivity study demonstrated that individuals with low (but not high) trait anxiety displayed positive resting state functional connectivity between the amygdala and mOFC [Kim et al., 2011]. Positive resting state activity between regions is interpreted here as reflecting an increased capacity for communication and mutual influence. Interestingly, however, negative resting state mPFC-amygdala resting state activity has been shown in rats, which is suggested to specifically reflect an inhibitory interaction [Liang et al., 2012]. Although the specific functional significance and direction of the functional connectivity may require additional consideration, the existing functional data also highlights the importance of cross-talk between mPFC and amygdala in anxiety. Complementing these functional effects, both Westlye et al. [2011] and Kim and Whalen [2009] similarly identify that white-matter in mOFC areas was negatively related to anxious traits. In sum, the present results together with other related findings support the conclusion that strong functional and structural connectivity between amygdala and mOFC is protective against anxiety, possibly by way of mechanisms associated with extinction learning.

We also made the novel observation that weaker amygdala connectivity with PHG and the posterior cingulate cortex was related to higher levels of trait anxiety. The functional significance of this finding is less clear. Nevertheless, both the PHG [Epstein and Kanwisher, 1998; Epstein and Higgins, 2007] and posterior cingulate cortex [Burianova and Grady, 2007; Maddock et al., 2001] have been implicated in memory as well as in spatial and contextual encoding. In addition, functional studies have revealed that both PHG and posterior cingulate cortex are also sensitive to emotional information [Maddock et al., 2001; Robinson et al., 2013]. Importantly, connections between the PHG and amygdala have been implicated in the context-dependent regulation of fear memory [see Maren et al., 2013]. Given the complimentary function of these regions, one possibility is that reduced amygdala connectivity with the PHG and PCC may be associated with a reduced capacity to restrict threat-related associations to specific contexts, resulting in greater susceptibility to generalized (i.e., trait) anxiety, as was observed in the current study. This remains speculative, however, and further work clarifying the role of these pathways in emotional learning and anxiety is warranted.

Implications for Clinical Studies of Anxiety

Although the current work involved an examination of anxiety in a nonclinical sample, our findings appear
consistent with previous studies of patients with anxiety disorders. Consistent with our finding that amygdala-mOFC connectivity was negatively related to anxiety, the most frequently reported finding in anxiety disorders is reduced FA in regions of the uncinate fasciculus [Baur et al., 2011; Hettema et al., 2012; Phan et al., 2009; Tromp et al., 2012]. On the other hand, whereas we found that greater connectivity between the amygdala and dACC was associated with higher anxiety, studies of patients with post-traumatic stress disorder find reductions in FA around the anterior cingulate cortex [Kim et al., 2005; Schuff et al., 2011]. However, the whole-brain approach used in those studies does not allow for inferences regarding amygdala-dACC connectivity per se. The current approach provides important information about the specificity of structural connections to the amygdala and their role in anxiety and affective disorders more specifically.

**Limitations and Future Directions**

It is important to note that probabilistic tractography is agnostic to whether connections are first-order or higher. It is therefore possible that the connections described between the amygdala and the given target regions are indirect. Recent models of emotion control do indeed implicate such indirect pathways in the modulation of amygdala function [Blair and Mitchell, 2009; Delgado et al., 2008; Mitchell, 2011]. It is also likely that multiple anatomical risk factors contribute to individual differences in anxiety and emotional reactivity. The current study demonstrated that an amygdala-centric network may represent one such risk factor, accounting for a small but significant proportion of the variance in trait anxiety in our sample of participants. However, it is possible that the use of alternative seed regions could yield other important pathways, as suggested by previous whole-brain DTI studies [Baur et al., 2011; Hettema et al., 2012]. The approach used in the present study could be adapted in the future to examine the pattern of connectivity of other structures implicated in anxiety [e.g., middle frontal gyrus, Bishop, 2009], which did not factor into the current model. In addition, the current evidence regarding laterality is equivocal and requires future research, as our findings differentially implicated both right and left amygdala, while previous research has emphasized either left [Kim and Whalen, 2009], or bilateral [Westlye et al., 2011] amygdala structural connections in trait anxiety. Furthermore, it is noteworthy that the amygdala, rather than acting as a homogenous unit, is made up of multiple functionally heterogeneous nuclei that can make different or even opposing functional contributions [Amaral, 2002; Janak and Tye, 2015]. In the present study, we do not speculate on which nuclei within the amygdala might be driving the particular observed effects due to concerns about reliably disting

**CONCLUSION**

The current findings demonstrate that an amygdala-centric network of structural connections accounts for a significant proportion of individual differences in trait anxiety. For the first time, we show that trait anxiety is positively related to the strength of structural connectivity between the amygdala and a distributed set of brain regions implicated in affect generation and perception, dACC, ITG, and temporal pole. These findings provide evidence that individual differences in structural connections may contribute to the information processing biases observed in dysregulated affect. Critically, consistent with previous functional and structural literature, we found that whereas amygdala-dACC connectivity was positively related to anxiety, amygdala-mOFC connectivity was negatively related to trait anxiety, emphasizing the importance of interactions between these structures in modulating anxiety and fear. We also found that greater connections between amygdala and regions implicated in memory and environmental context representation were protective against anxiety. Future work is needed to determine whether individual differences in the identified network arise during development (e.g., via Hebbian mechanisms), represent a
congenital risk factor, or both. The present study also provides a novel approach for estimating individual differences in personality traits from patterns of structural connectivity, which can be applied in a multitude of domains within social-affective neuroscience.

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REFERENCES


