Distracted and down: neural mechanisms of affective interference in subclinical depression

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Clinical and subclinical depression are associated with cognitive biases toward negative emotional information (Gotlib and Joormann, 2010; Joormann, 2010). Such biases purportedly arise from the match between negative information and the depressed individual’s mood (Dalgleish and Watts, 1990) and beliefs about the self and future (Beck, 1967, 2008). When negative material should be ignored, such as in executive function (EF) tasks that require control of attention (Banich, 2009; Miller and Cohen, 2001), performance impairments in depressed individuals have been attributed to the interfering effects of these biases (Gotlib and Cane, 1987; Joormann, 2004; Levin et al., 2007; Levens and Gotlib, 2010). In theory, the personal and emotional salience of negative information makes it especially distracting for depressed people and thus likely to tax cognitive control. Ultimately, the distracting nature of negative emotional information (affective interference) may launch ruminative self-referent thinking and interfere with goal-directed thinking and behavior.

In support of this theory, research aimed at identifying the neural mechanisms of affective interference has implicated brain regions involved in cognitive control, including lateral prefrontal cortex (LPFC), dorsal anterior cingulate cortex (dACC) and anterior insula (Mitterschiffthaler et al., 2008; Wang et al., 2008; Dichter et al., 2009; Engels et al., 2010; Herrington et al., 2010; Poland-Ross et al., 2013). Although the precise functions and network boundaries of these regions are unclear (see Seeley et al., 2007; Vincent et al., 2008; Power et al., 2011; Yeo et al., 2011), they can be broadly divided into a central executive or ‘frontoparietal’ network (FPN; Dosenbach et al., 2008; Vincent et al., 2008) and a ‘salience’ network (SN; Seeley et al., 2007). The SN and FPN are thought to work in concert to integrate information about stimuli salience with task goals, allocate attentional resources to other brain networks involved in external (i.e. ‘dorsal attention network’) or internal (i.e. ‘default network’, DN) attention and select and evaluate task-relevant behavior (Vincent et al., 2008; Menon and Uddin, 2010; Menon, 2011; Andrews-Hanna et al., 2014). The finding that affective interference alters activity of FPN and/or SN suggests that dysfunction in these systems may underlie the disruptive effects of affective biases on executive control.

However, ambiguity persists regarding the neural systems that are involved in affective interference in depression. First, despite brain systems involved in cognitive control being broadly implicated, evidence is mixed regarding which specific regions are involved, the function of such regions and the direction of effects (discussion in Murrough et al., 2011; Diener et al., 2012). Although several factors may contribute to such discrepancies, such as differences in the clinical or subclinical characteristics of the samples (Levin et al., 2007), these mixed findings are nevertheless difficult to interpret. In particular, it is not always easy to distinguish those aspects of abnormal neural response that represent dysfunctional or overtaxed cognitive control exerted to achieve task goals, vs task-irrelevant processing of salient negative material. One strategy to address these issues is to administer, in the same sample, a non-affective task that challenges cognitive control and has the same goals as the affective version of the task (Compton et al., 2003; Mohanty et al., 2007). This approach can localize brain regions that are recruited when cognitive control is taxed by task demands, regardless of the nature of distraction.
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Second, evidence is mixed regarding the neural systems involved in biased affective processing in depression. Several studies have detected increased activity in regions of the DN when depressed individuals attempt to disengage from (Johnson et al., 2009) or ignore (Mitterschiffthaler et al., 2008) emotional material. The DN has been shown to play an important role in internally directed thinking ranging from autobiographical memory to self-reflection (Buckner and Carroll, 2007; Andrews-Hanna et al., 2010, 2014); therefore, this pattern is consistent with the idea that affective biases stem from increased sensitivity to the personal salience of negative material and difficulty disengaging from self-focused rumination that may be evoked by negative material (Koster et al., 2011; Holtzheimer and Mayberg, 2011). However, other studies investigating affective bias have failed to implicate DN, instead linking depression to altered activity in brain systems implicated in basic emotional processing (Fales et al., 2008) or other regions (Diener et al., 2012). Together, these mixed results suggest that additional investigation of the neural mechanisms of affective bias is warranted.

Third, the dynamic nature by which brain systems involved in top-down control or in affective bias relate to one another has largely been inferred from patterns of co-activation rather than directly tested. Increasingly, researchers propose that depression is related to dysfunctions within and between large-scale brain networks (Hamilton et al., 2011, 2013; Marchetti et al., 2012; Whitfield-Gabrieli and Ford, 2012), suggesting the importance of examining interactions between neural systems in addition to magnitude of activation. One method for investigating relationships between brain regions is through psychophysiological interaction (PPI) analysis, in which correlations between activity in a seed region and activity in other regions of the brain are compared between task conditions. If affective interference were related to the interaction between regions involved in top-down control, and regions that process emotional or personally salient information, one would predict altered functional connectivity between such regions in individuals with elevated depression.

The present study aimed to address these issues by investigating the neural systems involved in affective interference in individuals with varying levels of depressive symptoms, from minimal (non-depressed) to severe (current depression). Participants were scanned with functional magnetic resonance imaging (fMRI) while performing an emotion-word Stroop task (Williams et al., 1996), instructing individuals to direct their attention to the color of ink in which an emotional word is printed while ignoring the meaning of the word (e.g. ‘lonely’ written in blue ink). Brain regions in which response to negative words is correlated with depression may be mechanisms of affective interference.

To localize brain regions recruited for top-down control in this sample, a color-word Stroop (Stroop, 1935) was also administered. The color-word Stroop has the same task goal as the emotion-word Stroop, i.e. attend to ink color, but incongruent words are distracting owing to their conflicting color meaning (e.g. ‘red’ written in blue ink) rather than affective content. These tasks differ in several aspects (e.g. semantic conflict), but their shared task goal makes them suitable for comparison to identify regions involved in top-down control in the service of that goal (Compton et al., 2003; Mohanty et al., 2007). Hence, the comparison of brain regions that are active for incongruent words across people, and those that are active for negative words at higher levels of depression, can help to isolate those top-down control systems that are challenged by negatively valenced affective interference. In contrast, brain regions in which response to negative words is associated with depression, but that fail to overlap with those that are responsive to incongruent words, may represent biased processing related to affective content. Finally, patterns of connectivity between regions involved in cognitive control and those involved in affective bias may provide unique insight regarding the network dynamics that underlie affective interference.

**METHOD**

**Recruitment and sample characteristics**

The sample consisted of 92 right-handed native English-speaking participants (ages 18–25 years, mean 19.03 years; 58% female; 80% European American), recruited from introductory psychology classes at the University of Illinois at Urbana-Champaign. Analyses performed with an overlapping sample, but which address research questions distinct from those of the present study, are reported elsewhere (Engels et al., 2010; Silton et al., 2011). Participants were pre-screened and selected for high variance in severity of current depressive symptoms, as assessed with the 8-item anhedonic depression subscale (MASQ-AD8) of the Mood and Anxiety Symptom Questionnaire (MASQ; Watson et al., 1995a, b; Nitschke et al., 2001). Participants were also screened and selected for low covariance between symptoms of depression and symptoms of anxiety, to minimize potential confounding effects of anxiety (see Supplementary Material).

Participants were screened to exclude: (i) use of psychoactive medications, (ii) abnormal color vision, (iii) previous loss of consciousness that exceeded 10 min, (iv) claustrophobia, (v) recent drug or alcohol use, (vi) excessive caffeine intake or (vii) recent lack of sleep. In addition, of 106 participants enrolled in the study, 14 were excluded from the present analyses for excessive motion in the scanner (N=6), equipment malfunction (N=7) or missing questionnaire data (N=1). Hence, the final sample for the present analyses consisted of 92 participants.

**MR data acquisition**

A 3T Siemens Allegra scanner with a quadrature headcoil was used for data acquisition. For functional scans, 370 functional images were acquired with the following echoplanar image (EPI) parameters: 2000 ms TR, 25 ms TE, flip angle 80°, FOV=22 cm. Thirty-eight oblique axial slices (3.4375 mm × 3.4375 mm in-plane resolution, 3 mm slice thickness, 0.3 mm gap between slices) were acquired parallel to the anterior and posterior commissures. These parameters were identical across functional runs. After the functional scans, a high-resolution T1-weighted image with the same slice prescription was acquired to provide anatomical data to register each participant’s functional data to standard space. For anatomical scans, a T1-weighted 160-slice MPRAGE sequence was acquired (1 × 1 mm in-plane resolution, 1 mm slice thickness, sequence parameters of 1700 ms TR, 3.5 ms TE, 900 ms inversion time). In addition, a multi-echo gradient-echo field map scan (TEs of 10 and 12.46 ms) was acquired before the EPI scans with a slice prescription identical to the functional slices for correction of geometric distortions.

**Procedures**

The present study consisted of two research sessions. During session 1, participants were informed of study procedures, provided written consent and completed psychosocial measures. During session 2, IMRI data were collected during two Stroop tasks, with task order counterbalanced across participants. Participants also completed an electroencephalography (EEG) session (see Silton et al., 2011).

**Assessment of depression**

At session 1, participants completed the MASQ (see Recruitment and sample characteristics section) (Table 1). Session 1 MASQ-AD8 (Nitschke et al., 2001) scores provided the primary measure of
depressive symptom severity used for all subsequent analyses. Using a continuous measure of depression is consistent with the view that depression is often better treated as a dimensional phenomenon (Hankin et al., 2005; Widiger and Samuel, 2005) and provides improved statistical power (Irwin and McClelland, 2003).

In addition, a graduate student in clinical psychology with extensive diagnostic training with the Structured Clinical Interview for Axis I Disorders, Non-Patient Edition (SCID-NP; First et al., 2007) administered the SCID-NP. A second experienced interviewer and a clinical faculty supervisor reviewed written case summaries detailing each criterion symptom, and assessed lifetime DSM-IV-TR diagnoses of depressive disorders (major depressive disorder, dysthymia or depressive disorder not otherwise specified) on the following scale: 1 = absent, 2 = features (≥2 symptoms), 3 = provisional (one symptom short of full criteria) and 4 = definite.

### Table 1 Sample characteristics

<table>
<thead>
<tr>
<th>Demographics</th>
<th>M (s.d.)</th>
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<tbody>
<tr>
<td>Age</td>
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<tr>
<td>MASQ-AD8</td>
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<tr>
<td>MASQ-AA</td>
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<td>PSWQ</td>
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<table>
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<tr>
<th>Clinical diagnoses</th>
<th>Percentage with Dx (%)</th>
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</thead>
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<tr>
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<tr>
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<tr>
<td>Substance/alcohol dependence</td>
<td>6.70</td>
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<tr>
<td>Eating disorder</td>
<td>2.17</td>
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</table>

Note. MASQ = Mood and Anxiety Symptom Questionnaire; -AD8 = Anhedonic Depression Subscale; -AA = Anxious Arousal Subscale; PSWQ = Penn State Worry Questionnaire; MDD = major depressive disorder; diagnoses = meets criteria for definite (full symptom criteria) or provisional (one symptom short of full criteria) disorder as revealed by a Structured Clinical Interview for Axis I Disorders (see ‘Methods’ section).

### Emotion-word Stroop task (e-Stroop)

The e-Stroop consisted of blocks of positive or negative emotion words alternating with blocks of neutral (non-emotional, e.g., ‘carpet’) words (Williams et al., 1996). Word stimuli, each presented one time per session, were selected from the set of Affective Norms for English Words (Bradley and Lang, 1999). The e-Stroop was designed to maintain arousal and frequency of usage. Positive words were selected and matched for arousal and word length; neutral words were selected for low arousal and neutral valence.

The e-Stroop included 16 blocks (4 positive, 4 negative and 8 neutral blocks; 32 s each), each consisting of 16 trials, for a total of one run of 256 trials. Also included were four fixation blocks (32 s each) in which a fixation cross was presented, and five periods of rest (6–34 s) in which participants viewed written instructions to relax with their eyes open. Each participant was randomly assigned to one of eight orders of stimulus presentation, optimized to control for stimulus order effects. Each trial consisted of a word presented in red, yellow, green or blue ink for 1500 ms followed by a fixation cross presented for 275–725 ms (onset-to-onset intertrial interval of 2000 ± 225 ms). Word presentation and recording of behavioral responses were controlled by STIM software (James Long Company, Caroga Lake, NY). Words were presented in capital letters via back projection onto a screen outside the scanner bore and a mirror fixed to the head coil. Participants responded using both hands (middle and index fingers), with a specific and unchanging response mapping of color to finger; 32 practice trials presented before the first Stroop task allowed the participant to acquire the stimulus-response mapping.

### Color-word Stroop task (c-Stroop)

The c-Stroop consisted of blocks of congruent or incongruent words alternating with blocks of neutral (non–color-related, e.g. ‘divide’) words, for 256 trials presented in 16 blocks (4 congruent, 4 incongruent, 8 neutral) (Stroop, 1935). Within congruent and incongruent blocks, 50% of trials were neutral to prevent reliance on word-reading strategies. Block counterbalancing, stimulus presentation parameters and color-response mapping were identical to that described above.

### Neuroimaging data analysis

#### Preprocessing

Image processing and analyses used tools from the FMRIB Software Library analysis package ([http://www.fmrib.ox.ac.uk/fsl](http://www.fmrib.ox.ac.uk/fsl)), AFNI ([http://afni.nimh.nih.gov/afni](http://afni.nimh.nih.gov/afni)) and Matlab (Mathworks, Inc., Natick, MA). First, each fMRI time series was motion-corrected with FMRIB’s Linear Image Registration Tool (jenkinson et al., 2002). Participants demonstrating <3.3 mm absolute motion or 2 mm relative motion were included in the analysis. Second, spikes (artifactual sudden intensity shifts) were corrected with the AFNI tool 3dDespike. Third, each time series was corrected for geometric distortions caused by inhomogeneity in the magnetic field. These distortions were corrected using Fugue in FSL with a field map collected before the EPI sequence, with the same slice prescription as the functional scans. The remaining preprocessing steps were conducted using FMRIB’s Expert Analysis Toolbox. The first three volumes of each data set were discarded, retaining volumes collected when the magnetic resonance signal was at a steady state, yielding 367 images per task. Each time series was temporally filtered with a high-pass filter (212 s) to remove drift in signal intensity, and spatially smoothed with a three-dimensional Gaussian kernel (full-width half maximum = 8 mm). Functional and structural data were registered into Montreal Neurological Institute 152 stereotaxic space with FMRIB’s Linear Image Registration Tool.

#### Lower-level single-subject analysis

FMRIB’s Improved Linear Model was used to perform regression analyses on each participant’s time series for each Stroop task separately, and statistical maps were created with a regression analysis performed at each intra-cerebral voxel (Woolrich et al., 2001). Regressors were created for each condition, with fixation blocks left as the unmodeled baseline. Each regressor was convolved with a double-gamma response function, yielding a per-voxel effect-size parameter estimate (β) map representing the magnitude of activity associated with that condition compared with baseline.

Conditions of interest for the present study were negative or incongruent words. For consistency, neutral words were selected as the baseline condition in each task, yielding two contrasts of interest: negative–neutral and incongruent–neutral. Because the present study focused on distractors that were expected to elicit affective interference, and previous studies have shown that depressed individuals fail to exhibit interference with positive distractors (Gotlib and McCann, 1984; Gotlib and Cane, 1987; Elliott et al., 2002; Clasen et al., 2013), positive words were not considered a condition of interest. However,
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because depression has been associated with deficits in other aspects of cognitive control with positive information (e.g., selecting or approaching positive material, Levens and Gotlib, 2010), individual difference analyses were also performed to investigate whether depressive symptoms predicted altered response to positive distractors (see Supplementary Material).

**Group-average and Individual difference analyses**

Higher-level statistical analyses were carried out with FMRIB’s Local Analysis of Mixed Effects. Monte Carlo simulations were performed with AFNI’s AlphaSim program (Ward, 2000) to correct for multiple comparisons at an overall familywise error rate of $P < 0.05$. Intrinsic smoothing of the functional data was calculated using AFNI’s 3DFWHMx. Voxels that survived a voxelwise threshold of $z > 2.58$ ($P < 0.01$) and an accompanying cluster-extent threshold of $P < 0.05$ were considered significant.

Four independent group-level analyses were performed. First, two group averages were computed to identify brain regions commonly recruited across the full sample in response to each contrast of interest. Second, two whole-brain correlation analyses were conducted to identify regions in which individual differences in depressive symptoms predicted activation to each contrast of interest.

**Conjunction analyses**

A conjunction analysis was performed to isolate regions in which independent significant effects were detected both in the depressive (correlation) response to the negative distractor contrast and in the group (average) response to the incongruent distractor contrast. This conjunction was expected to identify regions involved in cognitive control when such control is challenged. To be complete, a second conjunction was performed between the former and the group (average) response to the negative distractor contrast.

The conjunction procedure entailed comparing the $P$ value of each voxel between thresholded group maps (Nichols et al., 2005). If both $P$ values exceeded the threshold of $< 0.01$ and represented the same direction of effect (i.e., positive values signifying greater activity, or negative values signifying reduced activity), the voxel was set to 1; otherwise, the voxel was set to 0. Conjunction maps were cluster-corrected based on the more conservative threshold of the two component maps.

**PPI analyses**

PPI analyses (Friston et al., 1997; O’Reilly et al., 2012) were conducted to examine patterns of functional connectivity during affective interference. Specifically, PPI tested whether individual differences in depressive symptoms predicted connectivity between neural regions generally involved in cognitive control (i.e., regions implicated in conjunction analysis by both the depressive response to the negative distractor contrast, and the group-average response to the incongruent distractor contrast) and regions that may be involved in affective bias (i.e., regions implicated only in the depressive response to the negative distractor contrast, but not in the group response to the incongruent distractor contrast).

To accomplish the PPI analysis, at the peak of any cortical cluster identified by the conjunction of interest, a 5-voxel spherical seed region was created in MNI space. The seed region was transformed to native space, and the timecourse extracted for the e-Stroop. Lower-level models were rerun (using pre-processed data, O’Reilly et al., 2012), including all task regressors and three new regressors: the seed timecourse, the interactions between the seed timecourse and the negative distractor condition and the interaction between the seed timecourse and the neutral distractor conditions. The contrast of the latter two regressors provided a measure of the difference in functional connectivity of the seed region for negative vs neutral distractors. Group-level models were rerun as described above. To investigate patterns of connectivity within the set of brain regions implicated in affective interference, the PPI analysis was restricted to regions that were previously identified in the whole-brain correlation between depressive symptoms and neural response to negative distractors. To be complete, whole-brain PPI results are also included.

A second PPI analysis was performed using the same seed region(s), but examining average connectivity across the group for incongruent as compared with neutral distractors in the e-Stroop. This PPI investigated the normative patterns of functional connectivity of brain systems involved in cognitive control during a non-affective EF task.

**Behavioral analyses**

Average reaction time (RT) and percent accuracy were computed for each trial type, and comparison between conditions was performed by repeated-measures analysis of variance (ANOVA). Percent interference scores were calculated for negative ([negative RT-neutral RT]/neutral RT) and incongruent distractors ([incongruent RT-neutral RT]/neutral RT), a method that controls for scaling effects (Lansbergen et al., 2007). To examine the influence of depressive symptoms, mean-deviated MASQ-AD8 scores were added to ANOVAs as a covariate (Miller and Chapman, 2001; Verona and Miller, in press), and for any significant interactions follow-up correlations were performed.

Neuroimaging analyses were repeated including performance measures as covariates. Because the inclusion of such covariates failed to affect the pattern or significance of effects, simple analyses are reported here.

**RESULTS**

**Behavioral results**

**Performance across the group**

Significant differences in percent accuracy, $F(3,91) = 60.68, P < 0.01$, response speed, $F(1,91) = 68.63, P = 0.01$, and RT interference, $F(1,91) = 134.99, P < 0.01$, were observed between distractor types. Follow-up t-tests revealed that on average across the group, participants were slower and less accurate in responding to incongruent distractors than any other distractor type, $(Ps < 0.05$; Figure 1 and Supplementary Figure S1).

**Effects of depression on performance**

Trends emerged in which depressive symptoms moderated differences in accuracy, $F(1,91) = 2.96, P = 0.09$, response speed, $F(1,91) = 3.81, P = 0.05$, and RT interference, $F(1,91) = 2.85, P = 0.09$, between distractors. Follow-up correlations indicated that higher levels of depression predicted poorer accuracy, $r(91) = -0.30, P < 0.01$, and slower RT, $r(91) = 0.24, P = 0.02$, for negative distractors. However, depression also predicted somewhat poorer accuracy, $r(91) = -0.18, P = 0.08$, and slower RT, $r(91) = 0.23, P = 0.03$, for neutral distractors in the e-Stroop, and depression failed to predict RT interference to negative compared with neutral distractors, $r(91) = 0.08, P = 0.46$. Together, these results indicate that participants high in depression showed impaired performance on the e-Stroop for negative, but also to some extent for neutral, distractors (Figure 1). Depression was not significantly correlated with accuracy, RT or RT interference in the c-Stroop $(Ps > 0.11$; Supplementary Figure S1).
Neuroimaging results

Group-average response to negative distractors

Figure 2A shows regions in which, across the full sample, activity differed for negative compared with neutral distractors. Although this analysis was not directly relevant to the hypotheses tested in the present study, it shows the normative response to the task of ignoring negative information. Regions implicated included activation in bilateral areas of lPFC, and deactivation in midline regions including posterior cingulate cortex (PCC), medial prefrontal cortex (mPFC) and bilateral parahippocampal gyrus (PHG; Table 2).

Relationships between depressive symptoms and response to negative distractors

Figure 2B shows brain regions in which depressive symptoms predicted response to the negative distractor contrast. Critical to the present study goals, this analysis was conducted to identify neural systems involved in affective interference. Higher depression predicted increased activity in a region of dACC positioned at the intersection of the FPN and SN (Yeo et al., 2011), which has been suggested to play a role in cognitive control (Dosenbach et al., 2006). Elevated depression also predicted increased activity in regions of cuneus and DN (Figure 2D and Supplementary Materials).

Comparing brain systems involved in affective interference with brain systems involved in top-down control

The first key goal of the present study was to isolate top-down control systems that are challenged by affective interference, by comparing the depression-related neural response to negative distractors to the group (average) response to incongruent distractors. Conjunction analyses revealed a single prefrontal region in dACC that was active both when participants high in depression responded to negative (-neutral) distractors and when participants as a group responded to incongruent (-neutral) distractors. Additional subcortical regions were also
activated by both contrasts, e.g. caudate, brain stem and left thalamus (subcortical cluster extending to insula).

The second key goal of the present study was to identify neural regions implicated by affective interference that are outside the set of areas involved in top-down control, and in which activity may relate to biased processing of the emotional content of negative distractors. The above conjunction analysis indicated that PCC regions, in which depression-related activation to the negative distractor contrast was detected, failed to show activation to the incongruent distractor contrast across the group (indeed, overlapping regions of PCC were deactivated for the latter contrast). Together, these results suggest that PCC regions implicated in affective interference may be involved in biased emotional processing.

The results of the second conjunction analysis (conducted to be complete, but not the subject of the present hypotheses), between the depression-related and the group (average) responses to negative (-neutral) distractors, failed to reveal any regions of overlapping effects.

### Task-related changes in functional connectivity

The third key goal of the present study was to examine patterns of connectivity between regions involved in cognitive control and those involved in affective bias, either during affective interference (as predicted by depression) or during non-affective cognitive control (across...
The second PPI analysis tested whether, across the group, functional connectivity between the seed dACC and the affective-interference ROI differed for incongruent and neutral distractors. Within this ROI, activity in dACC was more positively correlated with activity in regions of caudate for incongruent than for neutral distractors. In addition, whole-brain PPI showed that dACC was more positively correlated with activity in other regions of FPN, e.g., dlPFC, and regions involved in attention to the external environment, e.g., dorsal parietal cortex, for incongruent than for neutral distractors (Figure 3C).

DISCUSSION
This study provides evidence that severity of depressive symptoms predicts affective interference to negative words, and predicts activity in brain regions involved in cognitive control or in internally directed thinking during affective interference. High functional connectivity between these systems may represent dynamic interactions between overtaxed or dysfunctional cognitive control and internal thoughts evoked by negative content. Together, these results suggest that affective interference stems from the increased salience of negative emotional information, coupled with impairments in allocating resources away from internal thought and toward the external environment.

Overtaxed or dysfunctional cognitive control in depression
Neural correlates of affective interference that represent dysfunctional or overtaxed cognitive control were identified in the present study by comparing the depression-related response to negative distractors with the normative response to incongruent distractors. Higher severity of depression predicted increased activity to negative distractors in a region of dACC that was also active in the group response to incongruent distractors, and which has been shown to be involved in cognitive control in previous research (Dosenbach et al., 2006). This pattern of hyperactivity converges with other studies examining affective interference (Mitterschiffthaler et al., 2008; Wang et al., 2008) and is consistent with the idea that the task of ignoring negative task-irrelevant information is especially taxing at higher levels of depression.

It may be that, because highly depressed individuals experience greater interference by negative material, they must recruit dACC to ‘pick up the slack’ in top-down control (e.g., Banich, 2009; Silton et al., 2011). Another interpretation is that dACC hyperactivity represents abnormalities in how cognitive control is exerted in the face of negative information, e.g., not only overtaxed top-down regulation of other brain systems but also qualitatively different regulatory effects (Johnstone et al., 2007). As an SN ‘hub’ that interacts with FPN and DN, dACC has been proposed to play an active role in allocating resources to external- or internal-attention systems (Sridharan et al., 2007). Such network interactions are believed to enable dynamic adjustments to changing environmental demands and support adaptive behavior (Corbetta et al., 2008).

Hence, in the case of affective interference, hyperactivity in dACC may represent abnormalities in top-down control for appropriate allocation of resources: attention should be allocated toward perceptual features of stimuli, but instead may be allocated to internal thoughts. Note that this interpretation does not require that dACC is responsible for determining how resources should be allocated (a process that may be supported by other regions, e.g., IPFC), but suggests that dACC may mediate such allocation.

Negative affective biases and autobiographical thought in depression
Important differences emerged in the set of regions that co-activated with (and were functionally related to) dACC in the case of affective interference vs non-affective cognitive control. For example, when
individuals across the group responded to incongruent distractors, activation was observed in dorsal parietal regions involved in attending to the external environment (Vincent et al., 2008), and increased functional connectivity was detected between dACC and such external-attention systems. In contrast, elevated depression was related to increased activity in DN regions in response to negative distractors, and greater functional connectivity was observed between dACC and PCC, an important 'hub' of the DN whose subdivisions 'echo' neural signals from other brain networks, including the SN (Buckner et al., 2009; Leech et al., 2012; Leech and Sharp, 2014). How might such DN hyperactivity relate to affective bias?

Putative functions of the DN include introspective, self-referent and autobiographical thinking (Buckner and Carroll, 2007; Spreng et al., 2008; Andrews-Hanna et al., 2014). Activity in the DN tends to decrease during externally focused tasks, when resources should be allocated away from introspection and toward task-relevant processing (Buckner et al., 2008). Indeed, in the present study, DN regions were broadly deactivated across the group in response to both tasks. Critically, this pattern was reversed for individuals high in depression, yet only when depressed individuals were confronted by negative distracting information. In such cases, increased severity of depression predicted greater activity in key DN regions including PCC and PHG, consistent with prior research (Northoff et al., 2011; Anticevic et al., 2012; Marchetti et al., 2012; Whitfield-Gabrieli and Ford, 2012). Hyperactivity of key DN regions involved in autobiographical thought supports the theory that depressed individuals have a tendency to engage in ruminative styles of thinking (Holtzheimer and Mayberg, 2011; Koster et al., 2011), particularly pertaining to negative and self-referential material (Andrews-Hanna et al., 2013). Together, this pattern suggests that DN hyperactivity may represent task-independent thinking prompted by the personal salience of negative cues.

**Dynamic relationships between default and control/SNs in depression**

The finding that affective interference is related not only to increased activity in but also amplified connectivity between dACC and PCC is consistent with the idea that depressed individuals exhibit abnormal network dynamics (Hamilton et al., 2013). Depression is conceptualized as a disorder characterized by difficulty shifting out of internally...
directed and ruminative thinking (Holtzheimer and Mayberg, 2011); the present results suggest that altered dACC function may contribute to such cognitive tendencies. Specifically, increased functional connectivity between dACC and regions of DMN may signify deficits in the capacity to switch between networks: when emotionally salient but task-irrelevant information is detected, individuals high in depression may be unable to shift away from internal thoughts and toward the external world. Together, these findings are consistent with evolving theories about the nature of pathological cognition in depression and suggest that brain systems involved in switching between large-scale networks may play a critical role in such pathology.

**Limitations and future directions**

The present study focused primarily on subclinical depression. This approach is designed to complement affective interference research conducted in clinical populations (see Supplementary Material), but the extent to which these findings would generalize to those with more severe symptoms is unknown.

The present study did not explicitly test the potential influence of anxiety in moderating neural mechanisms of affective interference. Symptoms of depression and anxiety are highly comorbid (Kessler et al., 2003), and anxiety has been linked to abnormal activity in brain regions that are also implicated in depression (Heller and Nitschke, 1998; Nitschke et al., 2001). Controlling for anxiety post hoc did not alter present results (see Supplementary Material), and the present sample was selected for low covariance between symptoms and depression. Hence, it is unlikely that comorbid anxiety was responsible for the effects observed here.

**CONCLUSIONS**

In conclusion, the present study suggests that affective interference in subclinical depression is related to overtaxed or dysfunctional cognitive control and disrupted allocation of cognitive resources to internal- or external-attention systems. This finding is consistent with previous research, and critically extends prior studies by directly testing functional correlates.

CONCLUSIONS was responsible for the effects observed here.

A meta-analysis of neurofunctional imaging studies of emotion and cognition in major depression. NeuroImage, 61, 677–83.


Gotlib, I.H., Hare, J. (2005). The neural basis of difficulties disengaging from negative irrelevant material in major depression. Psychological Science, 24, 334-44.


Affective interference in depression


