Predicting change in symptoms of depression during the transition to university: The roles of BDNF and working memory capacity

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Abstract Studies on depression risk emphasize the importance of both cognitive and genetic vulnerability factors. The present study has provided the first examination of whether working memory capacity, the BDNF Val66Met polymorphism, and their interaction predict changes in symptoms of depression during the transition to university. Early in the semester, students completed a self-report measure of depressive symptoms and a modified version of the reading span task to assess working memory capacity in the presence of both neutral and negative distractors. Whole blood was genotyped for the BDNF Val66Met polymorphism. Students returned at the end of the semester to complete additional self-report questionnaires. Neither working memory capacity nor the BDNF Val66Met polymorphism predicted change in depressive symptoms either independently or in interaction with self-reported semester difficulty. The BDNF Val66Met polymorphism, however, moderated the association between working memory capacity and symptom change. Among met carriers, lower working memory capacity in the presence of negative—but not neutral—distractors was associated with increased symptoms of depression over the semester. For the val/val group, working memory capacity did not predict symptom change. These findings contribute directly to biological and cognitive models of depression and highlight the importance of examining Gene × Cognition interactions when investigating risk for depression.

Keywords Depression · Working memory · Cognitive control · Stress · Gene–environment

The transition to university is a time of high stress (Bouteyre, Maurel, & Bernaud, 2007; Stader & Hokanson, 1998). Adolescents are challenged to adjust to increased academic rigor, independence from prior support systems, and greater financial responsibility (Bouteyre et al., 2007; Dwyer & Cummings, 2001; Smyth, Hockemeyer, Heron, Wonderlich, & Pennebaker, 2008). Students vary considerably in how well they adapt to these stressors (Osinsky, Lösch, Hennig, Alexander, & MacLeod, 2012). Although more than 40% of students report increased symptoms of depression during the transition to university, other students report no change, or even improvement, in their mood (Bouteyre et al., 2007; Osinsky et al., 2012). Both cognitive and genetic factors may be associated with individual differences in response to stressors (Beck, 1967, 2008; Gibb, Beevers, & McGeeary, 2013; Monroe & Simons, 1991). In particular, individual differences in cognitive control and in the met allele of the brain-derived neurotropic factor (BDNF) gene have both been associated with difficulty regulating emotions in response to stress and with increased risk for depression. In the present...
Increasing evidence points to the importance of cognitive control as a risk factor for depression (Goeleven, De Raedt, Baert, & Koster, 2006; Joormann, 2010; Joormann & Gotlib, 2010). Given that working memory has a limited capacity, efficient functioning depends on maintaining task-relevant materials while ignoring, or inhibiting, task-irrelevant distractions. Engle and colleagues (e.g., Engle, 2002; Engle, Tuholski, Laughlin, & Conway, 1999) have defined working memory capacity as the ability to maintain task-relevant information in working memory in the presence of distraction (see Conway et al., 2005, for an overview). Working memory capacity (as defined by Engle, 2002) does not directly assess the number of items that can be stored in working memory, but rather, assesses the ability to maintain task-relevant information by minimizing interference from task-irrelevant information. Thus, low working memory capacity indicates greater interference from irrelevant material. Low working memory capacity has been associated with difficulty regulating emotional responses to stress (Schmeichel & Demaree, 2010) and has been linked with various types of psychopathology, including depression (Goeleven et al., 2006; see Joormann, 2010, for a review).

The majority of research on working memory capacity has focused on interference from neutral distractors. Although depression is associated with deficits in general cognitive processes, it is more often characterized by difficulties with mood-congruent content (see Mathews & MacLeod, 2005), as is predicted by cognitive models of depression (Beck, 1967, 1976). Depressed individuals show particular difficulty ignoring negative information and disengaging from it (Joormann & Gotlib, 2008). In fact, difficulty controlling the contents of working memory in the presence of negative, as compared to neutral, distractors was a better predictor of baseline negative affect and also a better predictor of increases in negative affect in response to stress (Compton, Arnstein, Freedman, Dainer-Best, & Liss, 2011). It might thus be important to examine working memory capacity in the presence of not only neutral, but also negative, distractors.

Similar to the evidence on low working memory capacity, the met allele of the BDNF Val66Met polymorphism has been associated with impaired responses to stress and heightened vulnerability to depression (e.g., Alexander et al., 2010; Classen, Wells, Knopik, McGeary, & Beeevers, 2011; Colzato et al., 2011; Gatt et al., 2009; Pandey & Dwivedi, 2009; Pei et al., 2012; Shalev et al., 2009; Vinberg et al., 2009; see also the meta-analysis by Hosang, Shiles, Tansay, McGu Finn, & Uher, 2014). The genetic variation in exon 11 of the BDNF gene results in an amino-acid substitution from valine to methionine at codon 66 (Val66Met) and influences BDNF protein production. The met allele is typically associated with less BDNF secretion. The BDNF protein is central to neuronal growth, facilitates neuronal plasticity, and promotes adaptive responses to stress (McAllister, 2002).

Considerable evidence supports the BDNF hypothesis of depression, which posits that low levels of the BDNF protein play a central role in depression onset (Dwivedi et al., 2003; Karege et al., 2002; Karege, Vaudan, Schwald, Perroud, & La Harpe, 2005; Shimizu et al., 2003). However, the evidence for a link between the BDNF polymorphism and depression has been inconsistent. Whereas some studies have shown the met allele to be associated with major depressive disorder, particularly in interaction with stress (Carver, Johnson, Joormann, LeMoult, & Cuccaro, 2011; Gatt et al., 2009; Schumacher et al., 2005), other studies have shown the opposite (Chen, Li, & McGue, 2012), and still others—including a recent meta-analysis—have reported no association between the BDNF Val66Met polymorphism and depression (Oswald et al., 2005; Verhagen et al., 2010). Researchers, therefore, have begun to examine the associations between BDNF and cognitive risk factors, and particularly strong associations have been found between BDNF and constructs related to working memory capacity. For example, the BDNF met allele has been associated with deficits in working memory and executive functioning (e.g., Egan et al., 2003; Gatt et al., 2009; Rybakowski, Borkowska, Czerski, Skibińska, & Hauser, 2003), reduced gray-matter volume in corresponding brain regions, such as the hippocampus and prefrontal cortex (e.g., Bueller et al., 2006; Pezawas et al., 2004), and abnormal hippocampal activation during the N-back working memory task (e.g., Egan et al., 2003). Thus, research suggests that BDNF and working memory capacity are important and related risk factors for depression (see the reviews by Egan et al., 2003, and Joormann, 2010).

BDNF and working memory capacity may influence depression during times of stress in several ways. On the one hand, we might anticipate a mediation model, whereby working memory capacity mediates the relation between BDNF, stress, and depression. In this context, we might expect the BDNF met allele to be associated with lower working memory capacity, which in turn would be associated with increased symptoms of depression during increased stress. On the other hand, other models of depression (e.g., Gibb et al., 2013) posit a moderation model, whereby BDNF, working memory capacity, and stress interact to influence the pathophysiology of depression. According to this view, during times of stress, one might expect the influence of working memory capacity to depend on BDNF genotype: Whereas individuals with the val/val genotype might be protected against other risk factors for depression, given the associations between the BDNF Val66Met polymorphism and activity-dependent BDNF release in the brain (Egan et al., 2003), met carriers might be susceptible to risk factors that are related to BDNF, such as working memory capacity. Increasing evidence has been
found for moderation models of risk (Gibb, Benas, Grassia, & McGearry, 2009; Gibb, Uhrlass, Grassia, & Benas, 2009; Lau, Rijskijk, & Eley, 2006; Osinsky et al. 2012). Moreover, studies that have examined both mediation and moderation models have exclusively found support for the latter (Gibb, Benas, et al., 2009; Osinsky et al. 2012).

Our goal in the present study was to examine the roles of working memory capacity, BDNF, and stress in vulnerability to depression. We focused on BDNF and working memory capacity because of evidence that they are important and related risk factors for depression (for reviews, see Egan et al., 2013; Joormann, 2010). We predicted change in depressive symptoms over the course of a semester using self-reported semester difficulty, BDNF, working memory capacity in the presence of neutral distractors, and working memory capacity in the presence of negative distractors. In line with the BDNF hypothesis of depression, we expected BDNF to interact with stress experienced during the semester, such that greater stress would predict greater increases in depressive symptoms for met carriers. On the basis of the cognitive model of depression (Beck, 1967, 1976), we expected working memory capacity with negative distractors—but not with neutral ones—to interact with stress experienced during the semester, such that greater stress would predict greater increases in depressive symptoms for individuals with low working memory capacity for negative distractors. Moreover, building on Gene × Cognition × Environment models of depression (Gibb et al., 2013) and increasing empirical evidence (Osinsky et al. 2012; Gibb, Benas, et al., 2009; Gibb, Uhrlass, et al., 2009), we predicted that working memory capacity for negative distractors would be particularly predictive of increases in depressive symptoms for participants who were met carriers.

Method

Participants and procedure

Undergraduate students participated in exchange for partial credit toward a course requirement. Interested students replied to a posting on the department website early in the semester. They came to the laboratory in groups of about 20 to complete Session 1. After providing informed consent, participants completed a measure of working memory capacity and a self-report measure of depressive symptoms (described below). Blood also was drawn for genotyping by a trained phlebotomist. Approximately 2.5 months later, participants returned to the laboratory to complete Session 2, during which they completed additional self-report questionnaires. Of the 246 students (160 females, 86 males) who provided valid information at Session 1, 169 returned at the end of the semester, and of those, 167 completed all of the Session 2 measures. The sample size was determined according to power calculations, with the aim of detecting a moderate effect size that would be clinically significant (cf. Cohen’s $f^2 = 0.15$) with 80% power, assessed with a two-tailed alpha of .05. Power calculations were conducted using G*Power 3.1.7 on the basis of guidelines provided by Faul, Erdfelder, Buchner, and Lang (2009) for conducting power analyses for interaction terms. Participants’ reasons for attrition included students dropping the course or participating in alternative experiments (returning for the second session had not been specified as a precondition for participating in the first). The mean age of the final sample was 18.49 years ($SD = 1.69$). Participants self-identified as having the following ethnicities: 99 non-Hispanic White, 40 Hispanic, 10 Asians, 7 African Americans, 4 Caribbean islanders, and 7 “others.”

Working memory capacity

Working memory span tasks are widely used measures of working memory capacity (see Conway et al., 2005, for an overview). We created an affective version of the reading span task (RSpan; Engle et al., 1999; Kane et al., 2004) to assess individual differences in working memory capacity in the presence of negative versus neutral distraction. During each trial of the RSpan task, letters were presented one at a time for 1,000 ms each. Participants were asked to memorize these letters for a later check of memory. After the presentation of each letter, a sentence appeared, and participants were asked to determine whether the sentence was logical. If a sentence was logical (e.g., “I like to run in the park”), participants selected TRUE; if the sentence was illogical (e.g., “I like to run in the sky”), participants selected FALSE. The computer progressed to the next letter if participants made a response or if their response time was 1,000 ms longer than the person’s average reading time during practice trials. Cumulative sentence accuracy was recorded and displayed to the participants. Unanswered sentences were recorded as inaccurate, and participants were informed that their average sentence accuracy must be above 85% for the data to be valid (as recommended by Conway et al., 2005). Each trial consisted of between three and seven letter–sentence sets. At the end of each trial, 12 letters appeared in a $4 \times 3$ matrix on the screen, and participants indicated which of those letters they had been shown in the trial.

The present RSpan task differed from past versions in that half of the 30 trials presented sentences with negative content (e.g., “When I saw the man get shot I felt terrified and helpless.” or “Liz couldn’t stop crying when she found out that she had failed her class.”). The other half of the trials presented neutral sentences (e.g., “We like to eat eggs and bacon for breakfast in the morning.” or “The seventh graders had to build a volcano for their science class.”). The outcome measures of interest for this task were the sums of correctly
recalled letters on negative-sentence trials (RSpan-Negative) and neutral-sentence trials (RSpan-Neutral), as proportions of the total letters presented (see the guidelines by Conway et al., 2005). Lower RSpan scores reflected lower working memory capacity when performing in the presence of negative or neutral distractors, respectively. Stated differently, lower RSpan scores indicated greater interference from irrelevant material, reflecting less cognitive control.

Genotyping

Extraction and genotyping were completed at the Hussman Institute of Human Genomics, University of Miami Miller School of Medicine. Three nanograms of genomic DNA were extracted from whole blood according to established protocols. Genotyping was done for the G<sup>A</sup> (valinermethionine) variation at Position 758 of the BDNF coding sequence (rs6265), using Taqman allelic discrimination assays from Applied Biosystems (ABI). Cycling was performed on GeneAmp PCR Systems 9700 thermocyclers using conditions specified by ABI. After endpoint fluorescence was measured on the ABI 7900 HT system, genotype discrimination of the results was conducted using ABI’s HT Sequence Detection Systems version 2.3 analyses. As a check of genotyping accuracy, 32 quality control samples were included. The sample call rates were >99.7%. The BDNF genotype frequencies were as follows: 110 val/val, 47 val/met, and 10 met/met. The genotype frequencies were in Hardy–Weinberg equilibrium, \( \chi^2(N = 167) = 2.51, p > .05 \). In line with previous studies, the carriers of one or more met alleles were combined in one group (met carriers) and compared against the val/met group.

Questionnaires

**Depression** To assess depression severity within the past two weeks, participants completed the Beck Depression Inventory–II (BDI; Beck, Steer, & Brown, 1996). The BDI is a 21-item, self-report measure assessing the severity of depressive symptoms. It has high test–retest reliability (\( r = .93 \)) and good internal consistency (\( \alpha = .91; \) Beck, Steer, Ball, & Ranieri, 1996), including in student samples (Beck, Steer, & Carbin, 1988). This measure was completed at the first session early in the semester, and again at the second session later in the semester.

**Semester difficulties** To assess difficulties experienced during the semester, participants completed a ten-item questionnaire in the second session. The questions included, “I am happy overall with my academic performance this semester,” “All in all, my semester has been quite good,” and “My social life has been pretty nonexistent this semester” (reverse coded). Responses were made on a 5-point scale that ranged from 1 (I agree a lot) to 5 (I disagree a lot). Items were coded so that higher scores indicated a more difficult semester, \( M = 23.56, SD = 7.32, \alpha = .80 \).

**Results**

Sample characteristics

The participants who completed Session 2 did not differ from those who did not by ethnicity, Time 1 BDI score, working memory capacity score, or BDNF genotype, \( ps > .05 \). However, the nonreturners were older (\( M = 19.26, SD = 2.62 \)) than the returners (\( M = 18.49, SD = 1.69 \)), \( t(242) = 2.75, p = .01 \). In addition, a slightly lower percentage of females were in the nonreturner than in the returner group (55.84% and 68.86%, respectively), \( \chi^2(1, N = 244) = 3.91, p = .05 \). We therefore tested the effects of age and gender on change in depressive symptoms.

Table 1 shows the characteristics of the final sample, stratified by BDNF genotype groups. Genotype groups did not differ significantly in age, \( t(165) = 0.58, p = .57 \). The groups also did not differ in percentage of females, \( \chi^2(1, N = 167) = 0.01, p = .93 \), or ethnic distribution, \( \chi^2(5, N = 167) = 10.47, p = .06 \). Nonetheless, given that BDNF Val66Met allele frequencies have been found to differ across ethnic group, ethnicity was also entered in the model when testing our main hypotheses. Semester difficulty did not differ significantly across genotype groups, \( t(165) = 0.18, p = .86 \). Moreover, the genotype groups also did not differ significantly in RSpan-Negative or RSpan-Neutral, \( t(165) < 1, ps > .05 \).

**BDI and BDNF**

The genotype groups did not differ significantly in their BDI scores at the start of the semester (Time 1), \( t(165) = 1.78, p = .08 \). At the end of the semester (Time 2), however, met carriers reported higher BDI scores than did the val/met group, \( t(165) = 2.26, p = .03 \) (see Table 1).

**Cross-semester changes in BDI scores**

Across all participants, BDI scores did not change significantly across the semester, \( t(166) = 1.55, p = .12 \); however, we did observe high variability in cross-semester change in BDI scores (mean difference BDI-Time2 – BDI-Time1 = –0.62, SD = 5.13). A hierarchical multiple regression analysis was conducted to examine the effects of BDNF genotype and working memory capacity in the presence of negative and neutral distractors on cross-semester changes in BDI scores. Following recommendations by Jacobson and Truax (1991), and in line with Osinsky et al. (2012), we used the reliable change index proposed by Jacobson, Follette, and Revenstorf.
Table 1  Participant characteristics

<table>
<thead>
<tr>
<th>Variable</th>
<th>Val/Val (N=110)</th>
<th>Met Carrier (N=57)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Age, M (SD)</td>
<td>18.55 (1.62)</td>
<td>18.39 (1.83)</td>
</tr>
<tr>
<td>Female, %</td>
<td>69.09</td>
<td>68.42</td>
</tr>
<tr>
<td>Non-Hispanic White, %</td>
<td>54.55</td>
<td>68.42</td>
</tr>
<tr>
<td>Hispanic, %</td>
<td>28.18</td>
<td>15.79</td>
</tr>
<tr>
<td>African American, %</td>
<td>6.36</td>
<td>0.00</td>
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<tr>
<td>Caribbean, %</td>
<td>2.73</td>
<td>1.75</td>
</tr>
<tr>
<td>Asian, %</td>
<td>3.64</td>
<td>10.53</td>
</tr>
<tr>
<td>Other, %</td>
<td>4.55</td>
<td>3.51</td>
</tr>
<tr>
<td>Semester Difficulty</td>
<td>23.64 (7.23)</td>
<td>23.42 (7.56)</td>
</tr>
<tr>
<td>RSpan-Negative</td>
<td>0.78 (0.12)</td>
<td>0.79 (0.14)</td>
</tr>
<tr>
<td>RSpan-Neutral</td>
<td>0.78 (0.13)</td>
<td>0.79 (0.13)</td>
</tr>
<tr>
<td>BDI-Time1, M (SD)</td>
<td>7.08 (6.49)</td>
<td>9.05 (7.31)</td>
</tr>
<tr>
<td>BDI-Time2, M (SD)</td>
<td>6.22 (6.76)</td>
<td>8.91 (8.26)</td>
</tr>
</tbody>
</table>

BDI = Beck Depression Inventory; Time 1 was the beginning of the semester, Time 2 the end of the semester

(Summary text)

which accounts for measurement fluctuations over time. Thus, cross-semester change in BDI was quantified using the following formula:

$$\text{BDI change} = \frac{\text{BDI-Time2} - \text{BDI-Time1}}{S_{\text{diff}}}$$

where $S_{\text{diff}}$ is the standard error of the difference,

$$S_{\text{diff}} = \sqrt{2SE^2}, \text{and } SE^2 = S_1 \sqrt{1-r_{12}}.$$

Age, gender, and ethnicity were entered in Block 1, and BDNF, RSpan-Negative, RSpan-Neutral, and semester difficulty were entered in Block 2. The interactions of BDNF × Semester Difficulty, RSpan-Negative × Semester Difficulty, and RSpan-Neutral × Semester Difficulty were entered in Block 3. The interactions of BDNF × RSpan-Neg and BDNF × RSpan-Neut were entered in Block 4. Finally, the three-way BDNF × RSpan-Negative × Semester Difficulty and BDNF × RSpan-Neutral × Semester Difficulty interactions were entered in Block 5. Continuous variables were centered, and categorical variables were dummy coded.

Block 1 did not yield a significant effect, $R^2 = .04, F(7, 159) < 1, p = .56, \hat{f}^2 = .04$, but Block 2 accounted for a significant portion of variance, $\Delta R^2 = .07, \Delta F(4, 155) = 2.84, p = .03, \hat{f}^2 = .07$. Of the Block 2 variables, semester difficulty was the only one to significantly predict BDI change, $t(155) = 2.43, p = .02, \beta = .19, r_{\text{par}}(155) = .19$: Higher semester difficulty predicted greater increase in BDI scores. Block 3 did not account for a significant portion of variance, $\Delta R^2 = .02, \Delta F(3, 152) = 1.02, p = .39, \hat{f}^2 = .02$, but Block 4 significantly improved the prediction of BDI change, $\Delta R^2 = .04, \Delta F(2, 150) = 3.20, p = .04, \hat{f}^2 = .04$. The interaction between BDNF and RSpan-Neutral did not significantly contribute to BDI scores, $t(150) = 1.06, p = .29, \beta = .16, r_{\text{par}}(150) = .08$. The interaction between BDNF and RSpan-Negative, however, significantly predicted change in BDI, $t(150) = 2.31, p = .02, \beta = -.36, r_{\text{par}}(150) = -.17$. Block 5 did not significantly improve the prediction of BDI change, $R^2 = .02, \Delta F(2, 148) = 1.68, p = .19, \hat{f}^2 = .02$. We therefore examined the interaction between BDNF and RSpan-Negative without the higher-order interaction being included in the model.

Simple slope analyses determined that lower RSpan-Negative scores were associated with greater increases in BDI scores across the semester among met carriers, $t(45) = 2.70, p = .01, \beta = -.50, r_{\text{par}}(45) = -.30$, but did not have a significant effect among those homozygous for the val allele, $t(97) = 1.03, p = .31, \beta = .15, r_{\text{par}}(97) = .10$ (see Fig. 1).

Thus, for participants in the met carrier group, we found a negative association between working memory capacity in the presence of distraction from negative information and change in depressive symptoms over the course of the semester. Moreover, as was recommended by Roisman et al. (2012), and using tools developed by Preacher, Curran, and Bauer (2006), we conducted region-of-significance (RoS) analyses to examine the values of RSpan-Negative for which there was a significant difference between BDNF met carriers and those homozygous for the val allele at $\alpha = .05$. For RSpan-Negative (centered) scores less than $-0.4$, met carriers reported greater increases in depressive scores across the semester than did those homozygous for the val allele, $t(150) = 1.98$. In contrast, for RSpan-Negative (centered) scores greater than $0.36$, met carriers reported greater decreases in depressive scores across the semester than did those homozygous for the val allele, $t(150) = 1.98$.

**Discussion**

In this study, we examined whether self-reported semester difficulty, BDNF, working memory capacity, and their interaction predicted change in symptoms of depression during the transition to university, a period of heightened stress (Fisher &
In line with past research (Clarke, MacLeod, & Shirazee, 2008; Osinsky et al., 2012), we found substantial individual differences in cross-semester change in symptoms of depression. Neither BDNF nor working memory capacity independently predicted more than a nominal change in depressive symptoms. In addition, neither BDNF nor working memory capacity interacted with self-reported semester difficulty to predict more than a nominal change in depressive symptoms (less than 2%). BDNF genotype, however, moderated the association between working memory capacity and change in symptoms of depression. Specifically, for met carriers (i.e., individuals with what has been labeled the at-risk BDNF genotype), lower working memory capacity in the presence of negative— but not neutral—distractors was associated with increased symptoms of depression over time. In fact, working memory capacity in the presence of negative distractors predicted approximately 9% of the variance in depressive symptom change in this group. For those with the val/val genotype, however, working memory capacity predicted 1% of symptom change.

The findings from this study highlight the importance of including the BDNF Val66Met polymorphism in etiological models of depression. Although our findings do not indicate a direct link between BDNF and depressive symptoms, BDNF was important via its interaction with working memory capacity in the presence of negative distractors. This was the first study to examine the interaction between BDNF and working memory capacity as predictors of change in depressive symptoms; however, our findings are in line with those of other studies demonstrating that the interaction between genetic and cognitive factors predicts increases in depressive symptoms during times of stress in both children (Gibb, Benas, et al., 2009; Gibb, Uhrlass, et al., 2009) and young adults (Osinsky et al., 2012). Osinsky and colleagues, for example, examined whether the interaction between the serotonin transporter gene (5-HTTLPR) and attentional biases for negative information predicted students’ emotional changes across their first university semester. As in the present study, a Gene × Cognition interaction emerged: Attention biases to negative information predicted increased depressive symptoms across the semester only for those with the at-risk genotype.

Interestingly, neither BDNF nor working memory capacity interacted with semester difficulty to predict a substantial portion of change in BDI scores, with the effect size of $f^2 = .02$ being in the “small” range (Cohen, 1988). Thus, although the interaction of the two vulnerability factors predicted change in depressive symptoms during the transition to university, the vulnerability factors did not interact with our measure of stress to predict a significant amount of change in depressive symptoms. It is possible that the presence of stressful experiences is less important than are individual differences in the ways that stressful periods are processed, indexed via information-processing measures such as the ability to inhibit negative irrelevant information. This possibility is consistent with the observed interaction between BDNF and working memory capacity in the presence of negative distractors. It is important to interpret nonsignificant interactions with caution, however, given that our measure of...
stress was ad hoc, capturing only perceived stress from academic and social domains, without assessing important domains such as family or financial stress, and given that it did not include objective ratings of severity that have been shown to be of importance in Gene × Environment interactions (Karg, Burmeister, Shedden, & Sen, 2011). Our ability to consider stress effects may also have been limited by the relatively restricted variability in the amounts of stress that the participants reported. The present study differed from past research that had found Gene × Stress, or even Gene × Cognition × Stress, interactions (e.g., Chen et al., 2012; Gibb, Uhrlass, et al., 2009) in that all of our participants were experiencing an objectively stressful transition period (Bouteyre et al., 2007; Stader & Hokanson, 1998). It is also important to note that only a small change in depressive symptoms occurred across the semester. Although we observed significant variability in depressive symptom change, the overall magnitude of change may have been insufficient to have been influenced by stress. This may also explain why we found that the interaction between BDNF, working memory capacity, and semester difficulty exerted only a small effect on symptom change. However, this study was underpowered to demonstrate such an interaction, and the null results should be interpreted with caution. Future research should consider using stronger measures of stress to test these possible explanations.

The present study offers initial insight into how BDNF might increase risk for depression. One possibility is that the BDNF met allele increases depression risk not simply on the basis of whether people experience stressors, but through how they process them. In the present study, we found that BDNF interacted not with stress, but with individual differences in the way that negative material is processed. The same amount of stress might be quickly forgotten by some (those with high working memory capacity in the presence of negative distractors), yet might remain in the forefront of others’ memory (those with low working memory capacity in the presence of negative distractors). Thus, rather than focusing exclusively on whether participants experienced stress, future research might focus on individual differences in cognitive factors that influence how stress is processed (cf. Lazarus & Folkman, 1984).

It is also important to highlight that the met allele served as a vulnerability factor for participants who could not easily inhibit negative information, and yet—at the extreme—as a protective factor for those who could. This pattern of findings suggests a differential susceptibility pattern (Roisman et al., 2012), and it might explain the inconsistencies in past research examining BDNF, stress, and depression. Some past studies have shown a relation between the BDNF met allele and increased stress reactivity or increased risk for depression in interaction with stress (see the recent meta-analysis by Hosang et al., 2014). Others, however, have found the opposite (e.g., Alexander et al., 2010; Chen et al., 2012). As opposed to diathesis-stress models that only explain increased risk, differential susceptibility patterns offer insight into both risk and resilience (Roisman et al., 2012).

The present study was limited in its examination of changes in depressive symptoms versus the onset of a depressive episode. Future research should examine whether the interaction between BDNF and working memory capacity predicts who will go on to develop an episode of depression. A second limitation of the present study was that a sizable portion of the participants did not return for Session 2. One reason for the high attrition rates was that Session 1 was conducted in the early weeks of the semester, prior to the date when students were allowed to drop the course. In addition, students who completed Session 1 were also able to complete their course credit through participation in alternative experiments. An additional limitation was that the present study focused on undergraduate students, and it is possible that these findings would not generalize to more representative or impaired populations. Moreover, we should note that the negative sentences used in the present study were not depression-specific, but instead expressed negative affect in general. It will be important that future research examine the interaction between BDNF and cognitive control in the presence of depression-specific distractors.

Despite these limitations, the findings from the present study might provide insight into the inconsistent results linking BDNF and depression. Our findings also highlight the importance of examining Gene × Cognitive Bias interactions in etiological models of depression. Although the BDNF Val66Met polymorphism may not independently be a consistent predictor of changes in depressive symptoms, the importance of focusing on BDNF is highlighted by the fact that it confers risk in interaction with individual differences in cognitive control when processing negative information. Future research might also examine whether the interaction between BDNF and other executive functions or memory biases might also have similar effects, thereby enhancing our understanding of this debilitating disorder. Future research might also examine Gene × Gene × Cognitive Bias interactions. In particular, a sizable literature has indicated that the short allele of the serotonin transporter gene (5-HTTLPR) increases risk for depression in interaction with stressful life events (e.g., Karg et al., 2011). More recent work has also provided evidence that the interaction between 5-HTTLPR and BDNF increases risk for depression in combination with childhood adversity (e.g., Krishnan & Taylor, 2009). In addition, catechol-о-methyltransferase (COMT) has received increasing attention in recent years. In fact, Nagel and colleagues (Nagel et al., 2008) reported that BDNF met carriers performed significantly worse on executive functioning tasks if they were also COMT val carriers. Future work might consider incorporating both of these genes in order to test more complex etiological models of depression.
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References


