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Blood-based gene-expression biomarkers of post-traumatic stress disorder among deployed marines: A pilot study

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KEYWORDS

Alternative splicing; mRNA; Peripheral blood mononuclear cells; Microarray; Transcriptome; Trauma; Diagnosis; Biomarker; Antioxidant; Oxidative stress

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Summary: The etiology of post-traumatic stress disorder (PTSD) likely involves the interaction of numerous genes and environmental factors. Similarly, gene-expression levels in peripheral blood are influenced by both genes and environment, and expression levels of many genes show good correspondence between peripheral blood and brain tissues. In that context, this pilot study sought to test the following hypotheses: (1) post-trauma expression levels of a gene subset in peripheral blood would differ between Marines with and without PTSD; (2) a diagnostic biomarker panel of PTSD among high-risk individuals could be developed based on gene-expression in readily assessable peripheral blood cells; and (3) a diagnostic panel based on expression of individual exons would surpass the accuracy of a model based on expression of full-length gene transcripts. Gene-expression levels in peripheral blood samples from 50 U.S. Marines (25 PTSD cases and 25 non-PTSD comparison subjects) were determined by microarray following their return from deployment to war-zones in Iraq or Afghanistan. The original sample was carved into training and test subsets for construction of support vector machine classifiers. The panel of peripheral blood biomarkers achieved 80% prediction accuracy in the test subset based on the expression of just two full-length transcripts (GSTM1 and GSTM2). A biomarker panel based on 20 exons attained an improved 90% accuracy in the test subset. Though further refinement and replication of these biomarker profiles are required, these preliminary results provide proof-of-principle for the diagnostic utility of blood-based mRNA-expression in PTSD among trauma-exposed individuals.

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1. Introduction

Post-traumatic stress disorder (PTSD) is a severe anxiety syndrome that is currently diagnosed based on the emergence and persistence of core clinical symptoms including hyperarousal, re-experiencing, avoidance, or emotional numbing for a period greater than one month. Early psychosocial intervention after stress exposure may help reduce some of the symptoms and prevent the development of chronic PTSD (Litz et al., 2002). However, many individuals initially presenting with PTSD-like symptoms recover spontaneously and do not develop chronic PTSD (McFarlane, 1997). Thus, identifying which individuals will benefit most from early intervention can be challenging. Despite possible benefits of early intervention and a growing knowledge of the pathophysiology accompanying PTSD, a readily assessable diagnostic biomarker for PTSD has yet to be validated.

Classical descriptions of PTSD pathophysiology have included dysregulation of the hypothalamic-pituitaryadrenal (HPA) axis, but a specific pattern of dysregulation is not consistently observed across studies. Similarly, heightened inflammatory signaling has been reported in some (but not all) contexts (Baker et al., 2012b). Some have proposed a model of insufficient regulation of inflammatory signaling (Heinzelmann and Gill, 2013). Yet, there is an apparent paradox; *i.e.*, the observation that peripheral blood mononuclear cells (PBMCs) from PTSD patients show increased sensitivity to glucocorticoid-mediated suppression of an *insitu* inflammatory response (van Zuiden et al., 2012b).

There is considerable evidence that genetic effects, environmental influences, and their interaction play a role in the development of PTSD (Afifi et al., 2010; Koenen et al., 2009; True et al., 1993). There is a well-established body of clinical literature supporting a link between early life events, previous exposure to traumatic stress, and other psychosocial factors with the development of PTSD (Brewin et al., 2000; Ozer et al., 2003; DiGangi et al., 2013). Many biological investigations of PTSD have focused on the HPA axis and glucocorticoid receptor (GR) signaling pathways. In a cross-sectional study, Binder et al. (2008) identified an interaction between child abuse history and genetic polymorphisms in FKBP5 (a negative regulator of GR sensitivity) in predicting adult PTSD symptomology among a sample of non-psychiatric medical clinic patients. Mehta et al. (2011) described the association between genetic polymorphisms in FKBP5 and dysregulated neuroendocrine profiles described in PTSD. van Zuiden et al. (2012a) provided evidence that increased GR density is a pre-trauma risk factor for the development of PTSD and that dysregulation of GR density may be associated with an interaction between polymorphisms in the GR gene and concomitant early life stress. Another line of research suggests that genetic variants in corticotropin-releasing hormone type 1 receptor (CRHR1) are a risk factor for PTSD in children who were abused at an early age (Gillespie et al., 2009). PTSD is thus thought to be a disorder whose development is influenced by multiple genetic and environmental effects that establish a susceptible biological state; this vulnerability may be reflected in gene-expression signatures.

In light of the less-than-absolute heritability of PTSD and the prominent role of environmental factors, the pursuit of static genetic markers alone (*e.g.*, single nucleotide polymorphisms and copy-number variations) likely will not yield a suitable diagnostic biomarker. Gene-expression (*i.e.*, mRNA) levels, which potentially reflect the effects of both heredity and environment, may be better indicators of the aberrant biology underlying PTSD. PTSD clearly is a brain disorder, but assaying gene-expression levels – either acutely or longitudinally – in the brains of living human subjects at risk for PTSD is impossible. Yet, peripheral blood expression levels of many genes are moderately correlated with

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the expression levels of those genes in other tissues, including *postmortem* brain (Tylee et al., 2013) suggesting the possibility that peripheral blood gene-expression can be harnessed to construct useful profiles of brain disorders. Previous work by our group and by others has demonstrated that peripheral blood gene-expression provides a useful biomarker signal for a number of neuropsychiatric disorders, including schizophrenia, bipolar disorder, and autism spectrum disorders (Glatt et al., 2009, 2005; Tsuang et al., 2005).

In the context of PTSD, several prior studies identified differences in peripheral blood gene-expression levels between individuals with PTSD and similarly exposed comparison subjects without PTSD (Neylan et al., 2011; Segman et al., 2005; Yehuda et al., 2009; Zieker et al., 2007) (see Glatt et al., 2013 for a brief review of these studies). Taken together, these studies suggest that PTSD is associated with alterations in peripheral blood gene transcripts thought to play a role in HPA axis function, glucocorticoid signaling, immune and inflammatory signaling, and metabolism of reactive oxygen species. Consolidating this evidence with the results from a large body of epidemiologic, genomic, and neurobiological studies of the disorder (e.g., Uddin et al., 2010) led us to recently propose a theory of PTSD predicated on dysregulation of immune and inflammatory processes in general, and cellular immunity in particular (Baker et al., 2012b). We maintain that a variety of specific genetic factors and environmental influences may play a role in producing this dysregulated immune and inflammatory phenotype within different individuals. For this reason, we propose that a blood-based diagnostic biomarker calibrated to detect commonly-dysregulated immune and inflammatory transcripts may ultimately provide the best sensitivity for detecting PTSD within a clinical sample. Our previous work in this domain supports this hypothesis and further proposes that pre-existing dysregulation of immune and inflammatory processes may dispose individuals to develop PTSD at some future time, following exposure to traumatic stress (Glatt et al., 2013). Another recent publication, examining a large group of Marine Resiliency Study subjects across multiple cohorts, provided strong evidence that predeployment inflammation levels, assessed via measurement of plasma C-reactive protein level, were a strong positive predictor for the development of post-deployment PTSD after controlling for other risk factors (Eraly et al., 2014).

In the context of this prior work, we report here the results of a pilot study examining transcriptome-wide expression-profiling of pre- and post-exposure peripheral blood samples from individuals with uniquely elevated rates of trauma exposure and PTSD development: participants in the Marine Resiliency Study (MRS) following return from active war zones in Iraq or Afghanistan, as part of an ongoing longitudinal investigation (Baker et al., 2012a). The objectives of this pilot study were to evaluate the following hypotheses: (1) post-trauma expression levels of some genes in peripheral blood cells would differ between Marines with PTSD and matched comparison subjects; (2) a readily assessable, predictive biomarker panel of the PTSD diagnostic status could be developed based on gene-expression levels in peripheral blood cells; and (3) a diagnostic panel based on the expression of individual exons would surpass the accuracy of a model based on the expression of full-length transcripts of genes. We interpret the results of these analyses in two contexts: (1) as a means of identifying biological functions, processes, pathways, and protein domains whose genomic dysregulation may indicate or influence the development of the disorder; and (2) as an approach to the construction of classifiers that might ultimately assist in the clinical diagnosis of PTSD in such populations.

2. Methods

2.1. Ascertainment and clinical characterization of subjects

The MRS is a prospective study of factors predictive of PTSD among approximately 2,600 Marines in four battalions deployed to Iraq or Afghanistan. The research team conducted structured clinical interviews on Marine bases and collected blood samples and data at four time points: pre-deployment, and 1-week, 3-months, and 6-months postdeployment. Measures collected, including those used in this study, have been described in detail previously (Baker et al., 2012a).

The principal exclusion criteria were identical to those used for the pre-deployment gene-expression studies (Glatt et al., 2013). Subjects were excluded if they showed clinically significant PTSD prior to deployment, manifesting in: 1) a pre-deployment PTSD Checklist (PCL) score >44; and/or 2) a pre-deployment diagnosis of PTSD based on the Clinician-Administered PTSD Scale (CAPS). PTSD cases were identified as those subjects who were issued a CAPS-based PTSD diagnosis at 3- and/or 6-months post-deployment. Unaffected comparison subjects were identified as those subjects who, at no time, attained a PCL score >44 and who were not issued a CAPS-based PTSD diagnosis at any postdeployment interview. Among subjects who were included in the full MRS sample and assigned to case or comparison groups based on these criteria, we then selected for analysis 25 male PTSD cases and 25 male comparison subjects based on similar demographics, pre-deployment clinical characteristics, deployment history, and levels of trauma exposure as determined from the combat and post-battle experiences subscales of the deployment risk and resilience inventory (DRRI). The group of subjects selected for this study largely overlapped with those featured in our previous study of pre-deployment gene-expression (Glatt et al., 2013); 24 of the 25 PTSD cases and 23 of the 25 comparison subjects within the present study were also featured in the pre-deployment study. The demographic, clinical, and combat-experiential characteristics of the subjects are shown in Table 1. The two groups were comparable on all demographic and combat-experiential variables. Within both the case and comparison groups, 50% of the subjects had previously been deployed on at least one occasion and the average number of previous deployments was not significantly different between the two groups. Although no subject met diagnostic threshold for PTSD at pre-deployment as determined by either clinician ratings on the CAPS or self-ratings on the PCL, the eventual PTSD cases did have significantly higher clinician ratings on the

4

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Table 1	Demographic, clinical,	and experiential	characteristics of PTSD	cases and non-PTSD of	comparison subjects.
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PTSD cases	Comparison subjects	р
25	25	
22.4±3.1	21.9 ± 3.1	0.576
13 (52.0)	13 (52.0)	1.000
17 (68.0)	19 (76.0)	0.754
3 (12.0)	5 (20.0)	0.721
8 (32.2)	8 (32.2)	
14 (56.0)	12 (48.0)	
18.5 ± 13.0	19.3 ± 14.8	0.846
7.25 ± 4.5	8.0 ± 4.5	0.518
22.4±118	14.0 ± 8.7	0.006*
$\textbf{62.8} \pm \textbf{19.0}$	11.8 ± 10.8	<0.001*
24.3 ± 6.5	22.8 ± 3.4	0.330
$\textbf{42.9} \pm \textbf{17.2}$	$\textbf{23.0} \pm \textbf{5.2}$	<0.001*
49.0 ± 12.4	21.6±6.1	<0.001*
$\textbf{39.3} \pm \textbf{15.0}$	19.8 ± 2.4	<0.001*
	PTSD cases 25 22.4 \pm 3.1 13 (52.0) 17 (68.0) 3 (12.0) 8 (32.2) 14 (56.0) 18.5 \pm 13.0 7.25 \pm 4.5 22.4 \pm 118 62.8 \pm 19.0 24.3 \pm 6.5 42.9 \pm 17.2 49.0 \pm 12.4 39.3 \pm 15.0	PTSD casesComparison subjects252522.4 \pm 3.121.9 \pm 3.113 (52.0)13 (52.0)17 (68.0)19 (76.0)3 (12.0)5 (20.0)8 (32.2)8 (32.2)14 (56.0)12 (48.0)18.5 \pm 13.019.3 \pm 14.87.25 \pm 4.58.0 \pm 4.522.4 \pm 11814.0 \pm 8.762.8 \pm 19.011.8 \pm 10.824.3 \pm 6.522.8 \pm 3.442.9 \pm 17.223.0 \pm 5.249.0 \pm 12.421.6 \pm 6.139.3 \pm 15.019.8 \pm 2.4

Notes: (1) demographic characteristics of each sample are reported as mean \pm s.d. unless otherwise noted. (2) Sample means and proportions were compared using independent samples *t*-tests and chi-square tests, respectively. (3) *p*-Values < .05 are indicated with^{*}.

CAPS at pre-deployment, whereas no significant difference in pre-deployment self-ratings on the PCL were observed

2.2. mRNA sample acquisition, stabilization, isolation, and storage

Close collaboration with the Marine Corps and the Navy, which provides health support for the Marine Corps, enabled comprehensive on-site data collection. Clinical interviews and sample blood draw (10 ml) were both performed within 4h of each other on the same day, 3 months after return from deployment. Specific methods for stabilization, isolation and storage of mRNA samples were described previously (Glatt et al., 2013).

2.3. mRNA quantitation, quality control, and hybridization

Specific methods employed for mRNA sample quantitation and quality control assessment were also described previously (Glatt et al., 2013). The quantity and purity of mRNA in each of the 50 samples were sufficient for microarray hybridization assay. Two batches of 25 samples each (balanced with PTSD cases and controls) were then assayed on GeneChip[®] Human Exon 1.0 ST Arrays (Affymetrix, Inc.; Santa Clara, CA, USA) per the ''Whole Transcript Sense Target Labeling Assay'' protocol (Affymetrix, 2006) using 1 μ g of total RNA from each sample.

2.4. Microarray data import, normalization, transformation, summarization, and quality control

Partek[®] Genomics Suite software, version 6.6[©] 2012 (Partek Incorporated; St. Louis, MO), was utilized for all analytic procedures performed on microarray scan data. Interrogating probes were imported, and corrections for background signal were applied using the robust multi-array average

(RMA) method (Irizarry et al., 2003), with additional corrections applied for the GC-content of probes. The set of GeneChips was standardized using guantile normalization and expression levels of each probe underwent log-2 transformation to yield distributions of data that more closely approximated normality. As most genes were measured by multiple probe sets (typically one probe set per exon, but sometimes more), summarization of probes took place at two levels: first, probes tagging the same exon were summarized by median polish to arrive at one expression value per exon; second, exons tagging the same gene were summarized by median polish to arrive at one expression value per gene. All probesets were expressed with a signal:noise ratio \geq 3; thus, no probesets were excluded from analyses of differential expression. A total of 257,106 probesets were analyzed, mapping to 28,536 whole transcripts and 253,002 exons. Unsupervised clustering of subjects revealed no evidence of batch effects based on scan date. Principal Components Analysis (PCA) of the 50 post-deployment data points identified no outliers; all 50 subjects' data were well within the four-SD ellipsoid on each of the first three PCA dimensions, and deviation among redundant probes located within the same chip was low.

2.5. Microarray data analyses

Four independent sets of analyses were performed on the microarray data, as described below. For analyses of covariance (ANVOCAs), nominal uncorrected *p*-value thresholds were selected in order to generate reasonably sized lists of differentially expressed genes and exons for biological annotation analysis and machine learning classifier construction.

2.5.1. Identification of differentially expressed genes and their associated biological terms

We utilized ANVOCAs to determine which full-length genetic transcripts were differentially expressed at postdeployment in peripheral blood cells between PTSD cases

(n=25) and comparison subjects (n=25). We performed 28,536 ANCOVAs to assess each gene's expression level as a function of PTSD status (case or control), deployment cohort (three levels corresponding to three battalions deployed at different times), age, ancestry (dichotomized as Caucasian or not, as most subjects were Caucasian), and prior deployment status (first or subsequent deployment).

Family-wise Bonferroni-correction was applied to the ANCOVA *p*-values to determine whether any genes reached a genome-wide level of significance. To generate a relatively large candidate-gene list for functional profiling and construction of classifiers, we utilized an uncorrected typel-error rate for diagnosis in these analyses at 0.01. We then reduced the dimensionality of the resulting list of candidate biomarkers through analysis of annotation-enrichment using the DAVID algorithm (Dennis et al., 2003) to determine if the gene list disproportionately represented any biological ''terms''. Details of the enrichment analysis are described previously (Glatt et al., 2013). Bonferroni-correction was applied to the *p*-values obtained in the enrichment analyses of these annotation terms.

Pearson correlations were examined between each gene and the summed score from both DRRI subscales (Combat Experience Scale and Post-Battle Experience Scale), first within the PTSD group, and separately within the comparison group, in order to identify genes whose expression level varied with the degree of combat stress exposure. Family-wise Bonferonni and FDR *q*-values were used to correct for multiple observations. Among PTSD cases, the genes associated with the 200 most significant correlations were analyzed for biological annotations enrichment using DAVID.

2.5.2. Discovery and replication of gene-based diagnostic predictors

We utilized a machine-learning technique (support vector machine, SVM) to construct, evaluate, optimize, and cross-validate classification algorithms predicting PTSD status based on gene-expression levels at post-deployment for a training subset of our full sample. Training (n = 40)and validation (n = 10) subsets (distinct from those utilized in Glatt et al. (2013)) were carved from the full sample using pseudo-random selection in order to preserve a similar distributions of diagnostic status, demographic features (age, ancestry), and covariate values (deployment cohort, prior deployments) for both subsets. All analyses for classifier construction were carried out in the training subset and completely independent from the test subset. Using the same panel of factors and covariates described above, 28,536 ANCOVAs were performed; we generated a large list of candidate genes based on a nominally-selected uncorrected p < 0.01. The probes on this list were then supplied as potential predictors in an SVM, as various model parameters and predictor combinations were evaluated to identify the model with the highest accuracy in identifying cases and comparison subjects based solely on the expression levels of a minimal gene set identified by shrinking centroids after two-level nested 10-fold cross-validation. The top-performing model was then deployed on the test subset (5 cases and 5 comparison subjects) to determine its generalizability in accurately predicting case status based on gene-expression levels. (The 10 subjects used for model validation were not significantly different from those in the training set in terms of demographic, gene-expression QC, experiential, or clinical factors; data not shown.)

2.5.3. Identification of differentially expressed exons and their associated biological terms

Within the full sample (n=50), we examined exonexpression levels utilizing 22,204 ANCOVAs to identify putative alternative splicing differences between individuals with PTSD and comparison subjects. The same factors evaluated in gene-based analyses (PTSD status, cohort, age, ancestry and prior deployment status) were assessed for their main effects and their interaction with exonID as predictors of exon-expression levels; the PTSD status × exonID interaction term was examined as an indicator of putative alternative splicing, cf. (Glatt et al., 2009). Family-wise Bonferroni-correction was applied to the ANCOVA p-values to determine whether any interaction term reached a genome-wide level of significance. We utilized an uncorrected type-I-error rate of 0.0005 to obtain a candidate gene-list for functional profiling and classifier construction. Enrichment analyses were performed using the DAVID algorithm and were evaluated against a Bonferroni-corrected p-value accounting for the number of terms evaluated.

2.5.4. Discovery and replication of exon-based diagnostic predictors

As outlined above for full-length transcripts under Section 2, we used SVMs to construct, evaluate, optimize, and crossvalidate classification algorithms predicting eventual PTSD status based on exon-expression levels at pre-deployment for the same training subset of our full sample. Using identical subject allocations to training and validation subsets; we first generated a large candidate list of putatively alternatively spliced genes within the training subset (nominal uncorrected p < 0.0005 for the interaction of PTSD status and exonID), using 22,204 ANCOVAs and the same panel of factors, covariates, and interaction terms described above. For each gene on the list, the most significantly dysregulated exon was identified and supplied as a potential predictor in the SVM classifiers. Various model parameters and predictor combinations were then evaluated to identify the model with the highest accuracy in identifying cases and comparison subjects based solely on the expression levels of a minimal exon set identified by shrinking centroids after twolevel nested 10-fold cross-validation. The top-performing model was then deployed on the test subset (5 cases and 5 comparison subjects) to determine its generalizability in accurately predicting case status based on exon-expression levels.

2.6. Validation of gene-expression with quantitative real time polymerase chain reaction

A subset of nine transcripts featured in SVM classifiers were selected for validation with quantitative real time polymerase chain reaction (QRTPCR). First, total RNA was quantitatively converted with High-Capacity cDNA Reverse Transcription Kits (Applied Biosystems, San Diego City, CA) to generate single-stranded cDNA (for a $20 \,\mu$ l reaction). Aliquots of 20 ng of cDNA were analyzed via QRTPCR using

the Prism 7900 HT Fast Real-Time PCR system (Applied Biosystems). Statistical analysis was performed using the comparative \triangle CT method. All reactions were run in duplicate and normalized against gyceraldehyde-3-phosphate dehydrogenase (*GAPDH*) and hypoxanthine phosphoribosyl-transferase 1 (*HPRT1*). For one transcript (*GSTM1*), QRTPCR analysis was repeated with 75 ng cDNA in order to compensate for low signal. The fold change values were compared using independent samples *t*-tests (*p* < 0.05).

3. Results

6

3.1. Identification of differentially expressed genes and their associated biological terms

No gene's expression level was related to PTSD status at a Bonferroni-corrected level of significance, which is not surprising given the relatively small sample size and large number of transcripts tested. We did, however, identify 64 probes dysregulated with a nominally significant p < 0.01in Marines diagnosed with PTSD (Table 2). Thirty-three of these 64 probes were down-regulated, whereas 31 were up-regulated. Log2 fold-change (FC) of these probes in eventual PTSD cases ranged from 2.00-fold down-regulation to 1.66-fold up-regulation. Among the 64 probes, 59 were recognized pathway participants within the DAVID database: however, no significantly enriched annotations were identified. Exploratory pathway analysis of the differentially expressed genes in Table 2 using the Reactome database also revealed no significant enrichments. When examining geneexpression levels significantly correlated with summed DRRI score, no correlation p-values survived genome-wide correction among comparison subjects. Among PTSD cases, 13,336 correlation p-values survived an FDR correction threshold of 5%. The probesets featured in the 200 most significant correlations were submitted to DAVID for annotation enrichment analysis. The following terms were significantly enriched, with corresponding Bonferroni-corrected p-values: regulation of actin cytoskeleton (5 of 8 probesets down-regulated in PTSD, p=0.03), nucleotide-binding (17 of 30 probesets down-regulated, p = 0.04), host-virus interaction (6 of 10 probesets down-regulated, p = 0.07), and long-term depression (4 of 5 up-regulated in PTSD, p = 0.08).

3.2. Discovery and replication of a gene-based diagnostic predictor

To construct a gene-based classifier and assess its generalizability, we first derived a list of potential classifier transcripts as those probes with a difference in expression between PTSD case and comparison subjects attaining p < 0.01 in a training subsample of 20 cases and 20 comparison subjects while controlling for the same factors and covariates as in analysis 1. This analysis and filtering left 66 probes (Table 3) that were then used to build and optimize SVM classifiers. The optimal SVM (identified through two-level nested 10-fold cross-validation with shrinking centroids, cost = 401, tolerance = 0.001, kernel = radial basis function, and gamma = 0.001) comprised just 2 of the 66 starting probes (Table 3, probes in bold font) and attained

78% accuracy in classifying those individuals with PTSD in the training sample. We then tested the identical 2-gene SVM (with the same parameters, but with no shrinkage or cross-validation) in the remaining test subset (5 cases and 5 comparison subjects), where it yielded 80% accuracy. Among cases, four of five were correctly classified, while four of five comparison subjects were also classified correctly. These values correspond to a sensitivity, specificity, positive predictive value and negative predictive value in the test sample of 80%, 80%, 80%, and 80%, respectively. Expression levels for *GSTM1* and *GSTM2* are shown for PTSD cases and comparison subjects in Fig. 1. QRTPCR analysis demonstrated that *GSTM2* expression was reduced among PTSD cases, but results were less consistent for *GSTM1* (Table 4).

3.3. Identification of differentially expressed exons and their associated biological terms

The interaction of diagnosis and exonID identified putative isoform-expression differences (p < 0.0005) in 63 genes, 11 of which attained Bonferroni-corrected significance (Table 5). An example of between-group differences in exon expression for one of these eleven genes (DYNC1LI1) is illustrated in Fig. 2, where the PTSD cases have significantly lower levels of expression of a single probe corresponding to the fifth exon; this region corresponds to a retained intron, which could account for this pattern of expression difference. The list of 63 genes was analyzed by the DAVID algorithm and Reactome database (Table 6). DAVID analysis revealed five significantly enriched annotations (armadillo-like helical domain, macromolecule catabolic process, acetylation, modification-dependent protein catabolic process, modification-dependent macromolecule catabolic process). Analysis using the Reactome database revealed a single enriched pathway (class 1 MHC mediated antigen processing and presentation).

3.4. Discovery and replication of an exon-based diagnostic predictor

To construct an exon-based classifier and assess its generalizability we first identified potentially differentially spliced exons within our training subsample of 20 cases and 20 comparison subjects based on the diagnosis-x-exonID interaction term, using a nominal threshold of p < 0.00005, while controlling for the same factors and covariates as in the analyses above. For genes displaying more than one dysregulated probe between diagnostic groups, we selected the probe with the most significant between-group difference in expression level based on the p-values from planned comparisons. This analysis and filtering yielded 56 exons with expression differences between PTSD cases and comparison subjects (Table 7) that were then used to build and optimize SVM classifiers. The optimal SVM (identified through two-level nested ten-fold cross-validation with shrinking centroids, cost = 401, tolerance = 0.001, kernel = radial basis function, and gamma = 0.001) comprised 20 of the 56 starting probes (Table 7, probes in bold font) and attained 100% accuracy in classifying those individuals in the training sample with PTSD. We then tested the identical 20-exon SVM (with the same parameters, but with no

Blood-based gene-expression biomarkers of post-traumatic stress disorder among deployed marines: A pilot study

Transcript	Gene symbol	Gene product	Diagnostic group main effect			
cluster ID			Fold-change in cases	F	p	
7971296	EPSTI1	Epithelial stromal interaction 1 (breast)	1.66	7.6	8.6E – 0	
7921434	AIM2	Absent in melanoma 2	1.47	10.4	2.4E-0	
8056408	GALNT3	UDP-N-acetyl-alpha-D-galactosamine:polypeptide N-acetylgalactosaminyltransferase 3 (GalNAc-T3)	1.32	17.7	1.3E-0	
7970096	ING1	Inhibitor of growth family, member 1	1.28	12.3	1.0E – 0	
8046861	ITGAV	Integrin, alpha V (vitronectin receptor, alpha polypeptide, antigen CD51)	1.23	10.3	2.5E – 0	
8102817	ELF2	E74-like factor 2 (ets domain transcription factor)	1.23	8.7	5.2E – 0	
8124022	DTNBP1	Dystrobrevin binding protein 1	1.23	13.3	7.0E – 0	
8145175	PDLIM2	PDZ and LIM domain 2 (mystique)	1.21	9.3	3.9E – 0	
7953032	LRTM2	Leucine-rich repeats and transmembrane domains 2	1.20	8.3	6.1E-0	
7905131	CA14	Carbonic anhydrase XIV	1.20	11.4	1.5E – 0	
8044613	CBWD1	COBW domain containing 1	1.20	10.7	2.1E-0	
8067680	PRIC285	Peroxisomal proliferator-activated receptor A interacting complex 285	1.18	7.9	7.5E – 0	
8161537	CBWD3	COBW domain containing 3	1.18	8.3	6.2E – 0	
8155636	CBWD3	COBW domain containing 3	1.16	8.4	5.7E – 0	
8077099	SCO2	SCO cytochrome oxidase deficient homolog 2 (yeast)	1.15	7.5	9.0E - 0	
8012953	TRIM16	Tripartite motif-containing 16	1.15	8.2	6.5E – 0	
8037355	ZNF428	Zinc finger protein 428	1.15	8.4	5.8E – 0	
8161587	CBWD3	COBW domain containing 3	1.15	7.4	9.3E - 0	
7963157	RACGAP1	Rac GTPase activating protein 1	1.15	9.2	4.0E – 0	
7982868	CHAC1	ChaC, cation transport regulator homolog 1 (E. coli)	1.15	7.3	9.6E - 0	
7969638	ENST00000459449	Ncrna:snoRNA chromosome:GRCh37:13:95862598:95862702:1	1.14	9.1	4.3E – 0	
8053248	C2orf65	Chromosome 2 open reading frame 65	1.14	9.9	3.0E – 0	
8067773	ZNF512B	Zinc finger protein 512B	1.13	10.4	2.4E-0	
8074916	C22orf43	Chromosome 22 open reading frame 43	1.12	10.7	2.1E – 0	
8139921	CALN1	Calneuron 1	1.11	11.1	1.7E – 0	
7921492	IGSF9	Immunoglobulin superfamily, member 9	1.10	7.3	9.6E – 0	
7975562	PAPLN	Papilin, proteoglycan-like sulfated glycoprotein	1.10	7.8	7.9E – 0	
7996219	NDRG4	NDRG family member 4	1.09	8.7	5.0E – 0	
7919267	AK125616	cDNA FLJ43628 fis, clone SPLEN2027268.	1.09	8.4	5.8E – 0	
7919347	AK125616	cDNA FLJ43628 fis, clone SPLEN2027268	1.09	8.4	5.8E – 0	
8064242	NCRNA00176	Non-protein coding RNA 176	1.06	12.1	1.1E – 0	
8161133	SPAG8	Sperm associated antigen 8	-1.07	11.1	1.7E – 0	
7960434	GENSCAN00000019809	cDNA:Genscan chromosome:GRCh37:12:5141715:5142095:-1	-1.07	8.6	5.4E – 0	

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Transcript	Gene symbol	Gene product	Diagnostic group main e	effect	
cluster ID			Fold-change in cases	F	р
7901967	ENST00000489463	Ncrna_pseudogene:scRNA_pseudogene	-1.08	8.5	5.7E – 0
		chromosome:GRCh37:1:64121426:64121718:1			
7926670	ENST00000430957	cDNA:known chromosome:GRCh37:10:23425901:23426107:1	-1.08	8.6	5.2E – 0
8098167	C4orf39	Chromosome 4 open reading frame 39	-1.08	8.6	5.4E – 0
7937474	NS3BP	NS3BP	-1.08	8.2	6.3E – 0
7935690	ENST00000471360	Ncrna_pseudogene:Mt_tRNA_pseudogene chromosome:GRCh37:10:101817589:101817658:-1	-1.08	7.6	8.5E – 03
8090366	UROC1	Urocanase domain containing 1	-1.08	7.5	8.8E - 0
8103753	MORF4	Mortality factor 4	-1.09	8.0	7.2E – 0
8130939	DLL1	Delta-like 1 (Drosophila)	-1.09	7.7	8.0E - 0
8019437	CCDC57	Coiled-coil domain containing 57	-1.09	8.0	7.0E – 0
7966596	IQCD	IQ motif containing D	-1.10	13.5	6.3E – 0
7974695	ENST00000480540	cDNA:pseudogene chromosome:GRCh37:14:59261372:59261747:1	-1.10	8.3	6.0E – 0
8175815	PNCK	Pregnancy up-regulated non-ubiquitously expressed CaM kinase	-1.10	9.6	3.3E - 0
8030292	DKKL1	Dickkopf-like 1	-1.10	8.5	5.6E – 0
7997533	OSGIN1	Oxidative stress induced growth inhibitor 1	-1.11	7.7	8.0E - 0
8067671	SRMS	Crc-related kinase lacking C-terminal regulatory tyrosine and N-terminal myristylation sites	-1.11	7.6	8.3E-0
7998929	ENST00000470337	Ncrna_pseudogene:tRNA_pseudogene chromosome:GRCh37:16:3220961:3221031:-1	-1.12	7.3	9.7E – 0.
8017361	ENST00000460492	cDNA:pseudogene chromosome:GRCh37:17:60593682:60594128:-1	-1.12	15.2	3.2E – 04
7973900	C14orf19	Immunoglobulin (CD79A) binding protein 1 pseudogene	-1.13	12.8	8.4E – 0 [,]
7950321	UCP3	Uncoupling protein 3 (mitochondrial, proton carrier)	-1.13	11.0	1.8E – 0
7965838	ENST00000411000	Ncrna:snRNA chromosome:GRCh37:12:102190188:102190280:-1	-1.15	9.1	4.2E – 0
8113413	NUDT12	Nudix (nucleoside diphosphate linked moiety X)-type motif 12	-1.16	9.2	4.0E – 0
8175775	MAGEA1	Melanoma antigen family A, 1 (directs expression of antigen MZ2-E)	-1.16	7.5	8.8E - 0
7948667	AHNAK	AHNAK nucleoprotein	-1.18	7.9	7.3E – 0
7913252	PINK1	PTEN induced putative kinase 1	-1.18	7.9	7.5E - 0.02
8118974	RPL10A	Ribosomal protein L10a	-1.19	7.8	7.7E – 0
7979551	PPP2R5E	Protein phosphatase 2, regulatory subunit B', epsilon isoform	-1.20	7.3	9.6E - 0
7972021	TBC1D4	TBC1 domain family, member 4	-1.21	8.2	6.4E – 0
8106393	F2R	Coagulation factor II (thrombin) receptor	-1.23	9.1	4.3E - 0
7900597	C1orf50	Chromosome 1 open reading frame 50	-1.24	10.5	2.3E - 0
7903753	GSTM2	Glutathione S-transferase mu 2 (muscle)	-1.58	16.5	2.0E - 04
7903765	GSTM1	Glutathione S-transferase mu 1	-2.00	22.0	2.6F - 0'

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Transcript	Gene symbol	Gene product	Diagnostic group main	effect	
cluster ID"			Fold-change in cases	F	р
7904853	GPR89A	G protein-coupled receptor 89A (GPR89A), transcript variant 1, mRNA.	1.46	9.7	3.8E – 0
8095139	SRD5A3	Steroid 5 alpha-reductase 3 (SRD5A3), mRNA.	1.44	7.7	8.9E – 0
8056408	GALNT3	UDP-N-acetyl-alpha-D-galactosamine:polypeptide N-acetylgalactosaminyltransferase 3 (GalNAc-T3) (GALNT3), mRNA.	1.40	21.9	4.4E – 0
8102817	ELF2	E74-like factor 2 (ets domain transcription factor) (ELF2), transcript variant 1, mRNA.	1.30	10.0	3.3E – C
7970096	ING1	Inhibitor of growth family, member 1 (ING1), transcript variant 4, mRNA.	1.26	7.9	8.2E – 0
7938208	RBMXL2	RNA binding motif protein, X-linked-like 2 (RBMXL2), mRNA.	1.26	8.8	5.6E – C
8047784	ZDBF2	Zinc finger, DBF-type containing 2 (ZDBF2), mRNA.	1.23	8.3	6.8E-C
8124022	DTNBP1	Dystrobrevin binding protein 1 (DTNBP1), transcript variant 2, mRNA.	1.20	7.9	8.0E – C
7961418	ENST0000364606	Ncrna:rRNA chromosome:GRCh37:12:13593818:13593935:-1	1.18	11.6	1.7E – 0
8162562	C9orf130	Chromosome 9 open reading frame 130 (C9orf130), transcript variant 2, non-coding RNA.	1.16	9.3	4.5E – C
7969638	ENST00000459449	Ncrna:snoRNA chromosome:GRCh37:13:95862598:95862702:1	1.16	8.6	5.9E – C
7930561	HABP2	Hyaluronan binding protein 2 (HABP2), transcript variant 1, mRNA.	1.15	10.0	3.3E – 0
8067773	ZNF512B	Zinc finger protein 512B (ZNF512B), mRNA.	1.14	8.9	5.3E – C
7999356	AF090898	Clone HQ0149 PRO0149 mRNA, complete cds.	1.13	7.6	9.3E – 0
8139921	CALN1	Calneuron 1 (CALN1), transcript variant 2, mRNA.	1.10	9.0	5.0E – C
7949668	CCDC87	Coiled-coil domain containing 87 (CCDC87), mRNA.	1.10	7.9	8.2E – C
8091239	ENST00000516936	Ncrna:rRNA chromosome:GRCh37:3:142310519:142310633:-1	-1.03	7.5	9.8E – C
7910188	ENST00000365394	Ncrna:rRNA chromosome:GRCh37:1:227748882:227749001:1	-1.04	9.7	3.7E – C
8161133	SPAG8	Sperm associated antigen 8 (SPAG8), transcript variant 2, mRNA.	-1.07	9.3	4.4E – C
7925250	GNG4	Guanine nucleotide binding protein (G protein), gamma 4 (GNG4), transcript variant 2, mRNA.	-1.08	8.7	5.8E-C
8141423	MIR106B	MicroRNA 106b (MIR106B), microRNA.	-1.08	9.3	4.3E – 0
7926670	ENST00000430957	Cdna:known chromosome:GRCh37:10:23425901:23426107:1	-1.08	7.7	9.0E – 0
7966596	IQCD	IQ motif containing D (IQCD), mRNA.	-1.09	10.1	3.2E – 0
7946977	SAA4	Serum amyloid A4, constitutive (SAA4), mRNA.	-1.09	8.7	5.8E – C
8017361	ENST00000460492	Cdna:pseudogene chromosome:GRCh37:17:60593682:60594128:-1	-1.10	9.1	4.8E-C
8090366	UROC1	Urocanase domain containing 1 (UROC1), transcript variant 1, mRNA.	-1.10	8.5	6.2E – 0
8018673	QRICH2	Glutamine rich 2 (QRICH2), mRNA.	-1.11	8.2	7.1E – 0
7973900	C14orf19	Chromosome 14 open reading frame 19 (C14orf19), non-coding RNA.	-1.11	7.5	9.9E – C
7974695	ENST00000480540	Cdna:pseudogene chromosome:GRCh37:14:59261372:59261747:1	-1.11	9.7	3.7E – 0
8019437	CCDC57	Coiled-coil domain containing 57 (CCDC57), mRNA.	-1.11	8.5	6.1E – 0
8130939	DLL1	Delta-like 1 (Drosophila) (DLL1), mRNA.	-1.12	9.0	5.0E – 0
8146334	ENST00000343867	Cdna:pseudogene chromosome:GRCh37:8:48068735:48069425:1	-1.12	9.2	4.7E – 0
7988283	LOC645212	Hypothetical LOC645212 (LOC645212), transcript variant 1, non-coding RNA.	-1.12	8.6	5.9E – 0
8029693	FOSB	FBJ murine osteosarcoma viral oncogene homolog B (FOSB), transcript variant 1, mRNA.	-1.12	8.4	6.6E – 0
8034276	ZNF653	Zinc finger protein 653 (ZNF653), mRNA.	-1.13	12.8	1.1E – 0
8142997	PLXNA4	Plexin A4 (PLXNA4), transcript variant 1, mRNA.	-1.13	8.3	6.7E – 0
8012126	CLDNZ	Claudin 7 (CLDN7) transcript variant 1 mRNA	_1 13	77	9 OF _ C

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Transcript	Gene symbol	Gene product	Diagnostic group main	effect	
cluster ID#			Fold-change in cases	F	p
8071382	ZNF74	Zinc finger protein 74 (ZNF74), transcript variant 1, mRNA.	-1.13	8.5	6.3E – 03
8148607	GLI4	GLI family zinc finger 4 (GLI4), mRNA.	-1.14	8.2	7.2E – 03
8099279	ABLIM2	Actin binding LIM protein family, member 2 (ABLIM2), transcript variant 1, mRNA.	-1.14	11.7	1.7E – 03
8070744	C21orf2	Chromosome 21 open reading frame 2 (C21orf2), mRNA.	-1.14	8.3	6.8E – 03
8125149	SLC44A4	Solute carrier family 44, member 4 (SLC44A4), transcript variant 1, mRNA.	-1.15	7.5	9.7E – 03
8178653	SLC44A4	Solute carrier family 44, member 4 (SLC44A4), transcript variant 1, mRNA.	-1.15	7.5	9.7E – 03
8179861	SLC44A4	Solute carrier family 44, member 4 (SLC44A4), transcript variant 1, mRNA.	-1.15	7.5	9.7E – 03
7928306	ENST00000363300	Ncrna:misc_RNA chromosome:GRCh37:10:73980510:73980610:1	-1.15	9.4	4.3E – 03
8141795	POLR2J3	Polymerase (RNA) II (DNA directed) polypeptide J3 (POLR2J3), mRNA.	-1.16	8.0	7.9E – 03
8010629	CCDC137	Coiled-coil domain containing 137 (CCDC137), mRNA.	-1.16	14.6	5.3E – 04
8112159	ANKRD55	Ankyrin repeat domain 55 (ANKRD55), mRNA.	-1.16	9.4	4.2E – 03
8064014	SLC17A9	Solute carrier family 17, member 9 (SLC17A9), mRNA.	-1.16	12.1	1.4E – 03
8164665	RAPGEF1	Rap guanine nucleotide exchange factor (GEF) 1 (RAPGEF1), transcript variant 1, mRNA.	-1.16	8.6	6.0E – 03
8113413	NUDT12	Nudix (nucleoside diphosphate linked moiety X)-type motif 12 (NUDT12), mRNA.	-1.17	7.9	8.3E – 03
7948995	ATL3	Atlastin-3 gene:ENSG00000184743	-1.17	8.0	7.9E – 03
8130867	THBS2	Thrombospondin 2 (THBS2), mRNA.	-1.17	8.3	6.9E – 03
8010848	TBCD	Tubulin folding cofactor D (TBCD), mRNA.	-1.18	7.4	1.0E – 02
7965838	ENST00000363300	Ncrna:snRNA chromosome:GRCh37:12:102190188:102190280:-1	-1.18	10.9	2.3E – 03
8157804	OLFML2A	Olfactomedin-like 2A (OLFML2A), mRNA.	-1.19	8.3	6.8E – 03
8118974	RPL10A	Ribosomal protein L10a (RPL10A), mRNA.	-1.20	8.8	5.4E – 03
7972021	TBC1D4	TBC1 domain family, member 4 (TBC1D4), mRNA.	-1.26	8.1	7.4E – 03
7898679	NBPF3	Neuroblastoma breakpoint family, member 3 (NBPF3), mRNA.	-1.27	12.4	1.3E – 03
8088458	FHIT	Fragile histidine triad gene (FHIT), transcript variant 1, mRNA.	-1.27	7.7	8.8E - 03
8135268	EIF4B	Eukaryotic translation initiation factor 4B (EIF4B), mRNA.	-1.30	8.0	7.7E – 03
7900597	C1orf50	Chromosome 1 open reading frame 50, mRNA (cDNA clone MGC:2448 IMAGE:2959109), complete cds.	-1.31	14.8	4.9E – 04
8138950	RP9	Retinitis pigmentosa 9 (autosomal dominant) (RP9), mRNA.	-1.33	9.1	4.8E-03
8137464	PSPH	Phosphoserine phosphatase (PSPH), mRNA.	-1.35	7.8	8.4E – 03
7903753	GSTM2	Glutathione S-transferase mu 2 (muscle) (GSTM2), transcript variant 1, mRNA.	-1.51	12.9	1.0E – 03
7903765	GSTM1	Glutathione S-transferase mu 1 (GSTM1), transcript variant 1, mRNA.	-2.14	19.7	9.0E - 05

* Rows are sorted by decreasing fold-change in PTSD cases relative to non-PTSD comparison subjects.

[#] Transcripts in bold comprised the optimal 2-probe SVM classifier of PTSD status identified by training and testing in independent samples.

Blood-based gene-expression biomarkers of post-traumatic stress disorder among deployed marines: A pilot study



Fig. 1 Microarray-derived expression levels (ordinate) of summarized exon probesets reflecting whole-transcript expression levels (abscissa) of glutathione S-transferase mu 1 (*GSTM1*) and glutathione S-transferase mu 2 (*GSTM2*) in peripheral blood mononuclear cells from PTSD cases (n = 25) and comparison subjects (n = 25). These transcripts were notably down-regulated among PTSD cases within the full sample (fold changes -1.58 and -2.00, respectively) and were identified as the sole components of the optimal performing SVM classifier of diagnostic status, which achieved 80% accuracy in the test subset (n = 10; 4 of 5 cases correctly identified).

shrinkage or cross-validation) in the remaining test subset (n = 10; 5 cases and 5 comparison subjects), where it yielded a diminished but reasonable 90% accuracy (higher than the accuracy observed in gene-based analyses). All PTSD cases were correctly classified, while four of five comparison subjects were classified correctly. These values correspond to sensitivity, specificity, positive predictive and negative predictive values of 100%, 80%, 83% and 100%, respectively. QRTPCR analysis of seven exons in the classifier failed to detect significant differences in expression levels between PTSD cases and comparison subjects (Table 4).

4. Discussion

There is emerging support for the hypothesis that peripheral blood transcriptomic signatures associated with PTSD involve dysregulation of genes that function in immune and inflammatory processes or their regulation To this picture we add new and compelling pilot data suggesting that dysregulation of genes whose proteins function in the management of cellular oxidative stress may also be clinically useful biomarkers for distinguishing PTSD cases from trauma-exposed subjects who are resilient to PTSD.



Fig. 2 Microarray-derived expression levels (ordinate) of individual exon-probes (abscissa) of dynein, cytoplasmic 1, light intermediate chain 1 (*DYNC1L11*) in peripheral blood mononuclear cells from PTSD cases (n = 25, squares) and comparison subjects (n = 25, triangles). The interaction of diagnosis and exon ID was highly significant (p = 6.7E - 07, Bonferroni-corrected p = 1.4E - 02) owing to the selective down-regulated of an exon (probeset ID 8086013; p = 0.019) in the context of comparable expression levels of all other exons.

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D.S. Tylee et al.

Table 4 QRT-PCR	/alidation of gen	e and exon expression	on in full sample of PTS	SD cases and comparison su	ubjects.		
Assay ID	Gene	<i>GADPH</i> -normalize	ed average 2 ^{−∆Ct}	GADPH-normalized	HPRT1-normalized	average $2^{-\Delta Ct}$	HPRT 1-normalized
		PTSD	Control	p-values	PTSD	Control	p-values
Hs01683722_gH	GSTM1	0.367 ± 0.741	1.138 ± 1.529	0.031*	0.685 ± 1.367	1.251 ± 1.139	0.128
Hs03044640_gH	GSTM2	0.732 ± 0.596	1.114 ± 0.655	0.036*	0.300 ± 0.184	0.531 ± 0.262	0.001*
Hs00180203_m1	CUL2	1.000 ± 0.323	1.223 ± 0.612	0.114	0.768 ± 0.250	0.872 ± 0.350	0.236
Hs00211676_m1	DYNC1LI1	0.277 ± 0.261	0.333 ± 0.204	0.401	0.441 ± 0.434	0.466 ± 0.243	0.803
Hs00948075_m1	HUWE1	0.333 ± 0.204	0.586 ± 0.570	0.192	0.144 ± 0.105	0.208 ± 0.215	0.184
Hs00277883_m1	LARP7	0.674 ± 0.400	0.711 ± 0.272	0.709	0.689 ± 0.381	0.752 ± 0.331	0.537
Hs00382272_m1	PNPLA8	1.115 ± 0.240	1.123 ± 0.276	0.910	1.520 ± 0.694	1.254 ± 0.478	0.122
Hs01554570_m1	RBM5	0.462 ± 0.317	0.617 ± 0.621	0.271	0.268 ± 0.183	0.329 ± 0.303	0.390
Hs00208869_m1	TRAPPC8	0.788 ± 0.536	0.787 ± 0.375	0.990	$\textbf{0.548}\pm\textbf{0.372}$	$\textbf{0.612}\pm\textbf{0.313}$	0.512
Values are reported a	s mean±standar	d deviation. <i>p</i> -Values	< 0.05 are indicated with	*			

Yet, dysregulation of genes with immune-, inflammatoryand antioxidant-activity is probably only a small piece of the biological puzzle of PTSD pathophysiology, as many of the differentially expressed genes, as well as the exons comprising the best-performing PTSD-diagnostic classifier, were apparently unrelated to these functions; these other genes had highly disparate functions. Collectively, profiles of dysregulated genes in immune, inflammatory and other pathways may serve as potent biological indicators upon which diagnosis and early intervention may ultimately be based. The present study demonstrates proof-of-principle for the construction of blood-based PTSD diagnostic biomarkers that ranged in accuracy from 80 to 90% in a small subset that was held completely independent from classifier construction.

It is important to note that these classifiers employed decision-rules based solely on mRNA expression levels. To our knowledge, our group is among the first to employ datadriven (SVM) modeling on a list of differentially expressed transcripts in order to identify a subset of transcripts that were most predictive of PTSD status. These two strategies may be useful for identifying exons, genes, and pathways that play a role in the etiology of PTSD, but that may have been overlooked by other approaches focusing on well-established candidate genes. If these profiles of mRNAexpression differences in PTSD cases can be further refined and replicated, and if SVM-based models are found to perform reliably in larger or more diverse populations, then this study proposes an avenue for early diagnosis among trauma-exposed individuals, potentially fostering earlier intervention. However, it is likely that a more accurate classification model can be constructed in the future by taking into account additional known risk factors for PTSD, such as family history, personality traits, pre-existing mental disorders (Koenen et al., 2003a,b), peri-traumatic dissociation and post-trauma social support (Brewin et al., 2000; Ozer et al., 2003), non-genomic biological markers available in the MRS dataset (Baker et al., 2012b; Eraly et al., 2014), and other factors not necessarily related to gene-expression.

The present study did not account for many of these preand peri-traumatic risk factors, but future efforts to construct diagnostic models should seek to incorporate such data. Nevertheless, a single diagnostic classifier of PTSD (no matter how precisely constructed) may never perform with 100% accuracy, which is why it will be essential to pursue (in larger samples) the characteristics of subjects for whom such a classifier does not work. Of equal interest is the possibility that there are two or more unique biomarker profiles that are diagnostic of the same phenotypic outcome. In fact, etiologic heterogeneity may be a hallmark of complex disorders including PTSD, so it may not be possible to identify a single ''one-size-fits-all'' biomarker profile. In the future, methodologies that facilitate the identification of distinct biomarker profiles associated with the same phenotype may be required in order to account for etiologic heterogeneity in PTSD and other complex disorders. Another distinct possibility is that for some cases of PTSD there is no blood-based biomarker signature to be found. We are currently investigating each of these possibilities further. It is also important to acknowledge that the present study did not account for possible effects of pharmacological therapy (e.g., anti-depressants, anxiolytics, and antipsychotics) or

Blood-based gene-expression biomarkers of post-traumatic stress disorder among deployed marines:

Transcript cluster ID	Gene symbol	Gene product	Accession number	F	p	Adjusted p	q	Probesets (n)	Dysregulated probesets (<i>n</i>)	Dysregulatec probeset ID*
7903765	GSTM1	Glutathione S-transferase mu 1	NM_000561	7.8	1.1E – 09	2.3E – 05	2.3E – 05	9	9	7903767
7954810	LRRK2	Leucine-rich repeat kinase 2	NM_198578	2.6	2.8E-09	5.8E – 05	2.9E – 05	53	10	7954831
8158597	GPR107	G protein-coupled receptor 107	NM_001136557	3.1	2.2E – 07	4.4E-03	1.5E – 03	27	5	8158617
8052443	USP34	Ubiguitin specific peptidase 34	NM_014709	2.0	3.9E – 07	7.9E – 03	2.0E – 03	80	4	8052505
8022767	TRAPPC8	Trafficking protein particle complex 8	NM_014939	2.8	5.9E – 07	1.2E – 02	2.3E-03	31	2	8022775
8086008	DYNC1LI1	Dynein, cytoplasmic 1, light intermediate chain 1	NM_016141	4.5	6.7E – 07	1.4E-02	2.3E-03	13	1	8086013
8136662	MGAM	Maltase-glucoamylase (alpha-glucosidase)	NM_004668	2.4	9.3E – 07	1.9E – 02	2.3E-03	46	1	8136700
8070467	TMPRSS2	Transmembrane protease, serine 2	NM_001135099	4.2	9.6E-07	1.9E – 02	2.3E-03	14	1	8070472
8059596	TRIP12	Thyroid hormone receptor interactor 12	NM_004238	2.4	1.0E-06	2.0E - 02	2.3E-03	43	2	8059600
8142307	PNPLA8	Patatin-like phospholipase domain containing 8	NM_015723	3.9	1.1E-06	2.3E-02	2.3E-03	16	1	8142322
7978285	ADCY4	Adenylate cyclase 4	NM_001198592	2.9	2.1E-06	4.3E – 02	4.0E – 03	26	2	7978294
8149986	ZNF395	Zinc finger protein 395	NM_018660	3.9	2.6E – 06	5.2E – 02	4.3E-03	15	1	8149998
8058118	KCTD18	Potassium channel tetramerisation domain containing 18	NM_152387	5.5	5.0E – 06	0.10	7.8E – 03	8	1	8058121
8010454	RNF213	Ring finger protein 213	NM_020914	2.2	1.0E – 05	0.21	1.5E – 02	44	1	8010469
7946610	EIF4G2	Eukaryotic translation initiation factor 4 gamma, 2	NM_001418	2.8	1.5E-05	0.31	2.1E-02	24	1	7946612
7982904	RTF1	Rtf1, Paf1	NM_015138	2.8	2.6E – 05	0.52	3.3E – 02	22	3	7982911
8079869	RBM5	RNA binding motif protein 5	NM_005778	2.6	2.9E – 05	0.58	3.4E – 02	26	1	8079878
8076077	DDX17	DEAD (Asp-Glu-Ala-Asp) box polypeptide 17	NM_006386	3.2	4.1E-05	0.84	4.5E-02	16	3	8076093
7968128	PABPC3	Poly(A) binding protein, cytoplasmic 3	NM_030979	11.3	4.2E-05	0.85	4.5E-02	3	1	7968129
8172914	HUWE1	HECT, UBA and WWE domain containing 1	NM_031407	1.7	5.0E – 05	1.00	5.0E - 02	90	5	8172940
7903777	GSTM5	Glutathione S-transferase mu 5	NM_000851	4.7	5.2E – 05	1.00	5.0E - 02	8	3	7903782
8139896	PMS2P4	Postmeiotic segregation	NR_022007	8.0	6.3E-05	1.00	5.8E-02	4	1	8139900

Table 5 Exons significantly dysregulated in peripheral blood mononuclear cells from the full sample of eventual PTSD cases at post-deployment

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Ple	Table 5 (Co	ontinued)									
ase ci st-traun	Transcript cluster ID	Gene symbol	Gene product	Accession number	F	p	Adjusted p	q	Probesets (n)	Dysregulated probesets (<i>n</i>)	Dysregulated probeset ID⁺
ite ti natic	8076455	RRP7A	Ribosomal RNA processing 7 homolog A (S. cerevisiae)	NM_015703	6.3	9.0E - 05	1.00	7.5E – 02	5	1	8076458
his	7996677	NUTF2	Nuclear transport factor 2	NM_005796	4.9	1.0E – 04	1.00	7.5E – 02	7	2	7996683
res	8015642	PSMC3IP	PSMC3 interacting protein	NM_016556	3.5	1.0E – 04	1.00	7.5E – 02	12	1	8015646
irticle s dis	8058182	FAM126B	Family with sequence similarity 126, member B	NM_173822	3.5	1.0E-04	1.00	7.5E – 02	12	0	-
in sorder	8029884	SAE1	SUMO1 activating enzyme subunit 1	NR_027280	4.1	1.1E – 04	1.00	7.5E – 02	9	1	8029892
press am	8096938	LARP7	La ribonucleoprotein domain family, member 7	NM_016648	3.0	1.1E – 04	1.00	7.5E – 02	16	2	8096944
s as: ong	7965359	ATP2B1	ATPase, Ca ⁺⁺ transporting, plasma membrane 1	NM_001001323	2.5	1.1E – 04	1.00	7.5E – 02	24	1	7965379
Tyl deplo	8159984	C9orf46	Chromosome 9 open reading frame 46	NM_018465	4.8	1.1E – 04	1.00	7.5E – 02	7	1	8159991
ee	7965652	CDK17	Cyclin-dependent kinase 17	NM_002595	2.8	1.2E - 04	1.00	7.6E – 02	18	1	7965654
	8077858	ATG7	ATG7 autophagy related 7	NM_006395	2.8	1.2E – 04	1.00	7.6E – 02	18	1	8077874
.S., et narines:	7971602	RCBTB1	Regulator of chromosome condensation (RCC1) and BTB (POZ) domain containing	NM_018191	3.4	1.2E – 04	1.00	7.6E – 02	12	1	7971613
P al.			protein 1								
, Blı pilot	7971620	KPNA3	Karyopherin alpha 3 (importin alpha 4)	NM_002267	2.7	1.3E – 04	1.00	7.8E – 02	19	1	7971637
bod-b stu	7956910	CAND1	Cullin-associated and neddylation-dissociated 1	NM_018448	2.7	1.4E – 04	1.00	7.8E – 02	19	3	7956914
ased dy. I	8002778	MLKL	Mixed lineage kinase domain-like	NM_152649	3.3	1.4E – 04	1.00	7.8E – 02	13	1	8002781
gene	8160213	ТТС39В	Tetratricopeptide repeat domain 39B	NM_152574	2.6	1.4E – 04	1.00	7.8E – 02	22	2	8160226
e-expression oneuroendocr	7962112 7981068	CAPRIN2 SERPINA1	Caprin family member 2 Serpin peptidase inhibitor, clade A (alpha-1 antiproteinase, antitrypsin), member 1	NM_001002259 NM_001002236	2.5 4.0	1.5E — 04 1.6E — 04	1.00 1.00	7.8E — 02 8.1E — 02	24 9	1 2	7962123 7981074
bior	8103951	ACSL1	Acyl-CoA synthetase long-chain family member 1	NM_001995	2.5	1.8E-04	1.00	9.3E – 02	23	1	8103965
narkers of gy (2014),	7999841	SMG1	SMG1 homolog, phosphatidylinositol 3-kinase-related kinase (<i>C</i> . <i>elegans</i>)	NM_015092	2.0	2.0E – 04	1.00	1.0E – 01	42	4	7999869

post-traumatic stress disorder among deployed marines: http://dx.doi.org/10.1016/j.psyneuen.2014.09.024

Table 5 (Co	ntinued)									
Transcript cluster ID	Gene symbol	Gene product	Accession number	F	p	Adjusted p	q	Probesets (n)	Dysregulated probesets (<i>n</i>)	Dysregulated probeset ID⁺
7933047	CUL2	Cullin 2	NM_003591	2.5	2.1E-04	1.00	0.10	23	2	7933056
8017162	RNFT1	Ring finger protein, transmembrane 1	NM_016125	3.6	2.2E – 04	1.00	0.10	10	3	8017163
7967881	MPHOSPH8	M-phase phosphoprotein 8	NM_017520	3.0	2.4E – 04	1.00	0.11	14	1	7967895
7900426	SMAP2	Small ArfGAP2	NM_022733	3.4	2.5E – 04	1.00	0.11	11	1	7900432
7903893	CD53	CD53 molecule	NM_000560	3.3	2.5E – 04	1.00	0.11	12	2	7903894
8078569	GOLGA4	Golgin A4	NM_002078	2.2	2.6E – 04	1.00	0.11	30	2	8078594
7967117	OASL	2'—5'-oligoadenylate synthetase-like	NM_003733	3.8	2.6E – 04	1.00	0.11	9	3	7967123
7975416	PCNX	Pecanex homolog (Drosophila)	NM_014982	2.1	2.7E – 04	1.00	0.11	36	1	7975429
8179298	CSNK2B	Casein kinase 2, beta polypeptide	NM_001320	3.6	2.8E - 04	1.00	0.11	10	1	8179308
8143327	PARP12	Poly (ADP-ribose) polymerase family, member 12	NM_022750	3.1	2.9E – 04	1.00	0.11	13	2	8143336
8021496	KIAA1468	KIAA1468	NM_020854	2.2	2.9E – 04	1.00	0.11	31	4	8021504
8141846	FBXL13	F-box and leucine-rich repeat protein 13	NM_145032	2.3	3.0E - 04	1.00	0.12	25	4	8141865
7929677	PI4K2A	Phosphatidylinositol 4-kinase type 2 alpha	NM_018425	3.3	3.2E – 04	1.00	0.12	11	0	-
8088348	FAM116A	Family with sequence similarity 116, member A	NM_152678	2.5	3.3E – 04	1.00	0.12	20	2	8088366
7937363	PKP3	Plakophilin 3	NM_007183	3.0	3.5E – 04	1.00	0.13	14	2	7937370
8089785	POPDC2	Popeye domain containing 2	NM_022135	3.7	3.6E – 04	1.00	0.13	9	2	8089794
7967563	UBC	Ubiquitin C	NM_021009	2.9	3.6E – 04	1.00	0.13	14	2	7967584
8077171	RABL2B	RAB, member of RAS oncogene family-like 2B	NM_001130921	3.0	3.7E – 04	1.00	0.13	13	1	8077180
8112772	AP3B1	Adaptor-related protein complex 3, beta 1 subunit	NM_003664	2.2	3.7E – 04	1.00	0.13	28	2	8112794
7997626	KLHL36	Kelch-like 36 (Drosophila)	NM_024731	4.7	3.9E – 04	1.00	0.13	6	2	7997629
8043251	РТСД3	Pentatricopeptide repeat domain 3	NM_017952	2.3	5.0E – 04	1.00	0.16	24	0	-
8072170	KREMEN1	Kringle containing transmembrane protein	NM_032045	2.9	5.0E – 04	1.00	0.16	14	4	8072173

 * Rows are sorted by increasing *p*-value for the interaction of diagnosis and exonID.

* Exon probesets listed were the most significantly differentially expressed (per gene) between diagnostic groups. Information on the identities of all dysregulated exons for each gene can be obtained from the authors upon request.

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DAVID category	Term	Count (%)	Fold-enrichment	p	Bonferroni- corrected p	Gene corresponding to dysregulated exon
INTERPRO	IPR011989:armadillo-like helical	7 (11.1%)	14.9	6.5E-06	1.0E-03	KIAA1468, AP3B1, PKP3, CAND1, TRIP12, KPNA3, LRRK2
GOTERM_BP_FAT	GO:0009057 \sim macromolecule catabolic process	12 (19.0%)	4.5	4.2E – 05	1.8E – 02	SAE1, SMG1, USP34, FBXL13, HUWE1, CAND1, ATG7, TRIP12, MGAM, CUL2, KLHL36, UBC
SP_PIR_KEYWORDS	Acetylation	22 (34.9%)	2.5	5.8E – 05	7.9E – 03	EIF4G2, AP3B1, NUTF2, SMG1, HUWE1, PTCD3, CUL2, LARP7, UBC, ACSL1, KIAA1468, CSNK2B, SAE1, PABPC3, RTF1, ATG7, CAND1, KPNA3, RNF213, RBM5, MPHOSPH8, GSTM5
GOTERM_BP_FAT	GO:0019941 ~ modification-dependent protein catabolic process	10 (15.9%)	5.1	1.0E-04	4.3E – 02	SAE1, USP34, FBXL13, HUWE1, CAND1, ATG7, TRIP12, CUL2, KLHL36, UBC
GOTERM_BP_FAT	GO:0043632 ~ modification-dependent macromolecule catabolic process	10 (15.9%)	5.1	1.0E – 04	4.3E – 02	SAE1, USP34, FBXL13, HUWE1, CAND1, ATG7, TRIP12, CUL2, KLHL36, UBC
Reactome	Term	Count (%)		р	FDR-corrected p	Gene corresponding to dysregulated exon
	Class 1 MHC mediated antigen processing and presentation	6 (9.9%)		1.0E – 04	1.5E – 02	SAE1, CUL2, ATG7, TRIP12, HUWE1, UBC

Table 6 Annotations enriched at corrected significance levels among differentially-expressed exons (*p* < 0.0005) in peripheral blood mononuclear cells from the full sample of PTSD cases at post-deployment^{*}.

^{*} Rows are sorted by increasing *p*-value for the enrichment of annotations.

Table 7Exons Significantly Dysregulated in Peripheral Blood Mononuclear Cells from a Subset of PTSD Cases at Post-Deploymentand Used in Predictive SVM Classifiers*.

Transcript Gene Symb Cluster ID ^{#,}		Gene Product	Interaction p	Exon ID	Fold-change	F	p
8096938	LARP7	La ribonucleoprotein domain family, member 7	ein domain 7.2E – 09 809		3.52	8.6	5.9E – 03
8097148	RNF213	Ring finger protein 213	7.1E – 11	8010469	3.06	9.6	3.8E – 03
8086008	DYNC1LI1	Dynein, cytoplasmic 1, light intermediate chain 1	4.0E – 10	8086013	3.06	7.9	8.3E-03
8134122	PTCD3	Pentatricopeptide repeat domain 3	1.6E – 07	8043256	3.05	7.7	9.0E - 03
8142307	PNPLA8	Patatin-like phospholipase domain containing 8	5.3E-09 8142322 2		2.87	9.3	4.4E-03
8172914	DIDO1	Death inducer-obliterator 1	3.7E – 06	8067576	2.77	5.1	3.0E - 02
8076455	RRP7A	Ribosomal RNA processing 7 homolog A (S.cerevisiae)	1.7E – 05	8076458	2.74	12.8	1.1E-03
8158597	GPR107	G protein-coupled receptor 107	2.3E – 10	8158617	2.74	11.5	1.8E – 03
8054092	CUL2	Cullin 2	3.4E – 06	7933056	2.71	11.9	1.5E – 03
8105191	KIAA1468	KIAA1468	1.0E – '07	8021504	2.71	11.1	2.1E-03
8045090	ZC3H11A	Zinc finger CCCH-type containing 11A	1.1E – 05	7908985	2.67	6.5	1.6E – 02
8056837	TTC17	Tetratricopeptide repeat domain 17	1.1E – 06	7939453	2.63	6.3	1.7E – 02
8079869	NEK9	NIMA (never in mitosis gene a)- related kinase 9	6.7E – 06	7980282	2.62	8.4	6.4E-03
8079869	RBM5	RNA binding motif protein 5	1.5E – 08	8079878	2.62	8.9	5.3E – 03
8105191	PARP8	Poly (ADP-ribose) polymerase family, member 8	1.6E – 05	8105199	2.57	4.6	4.0E - 02
8083523	AQR	Aquarius homolog (mouse)	2.6E – 05	7987328	2.57	6.3	1.7E – 02
8136662	MGAM	Maltase-glucoamylase (alpha-glucosidase)	5.1E-08	8136700	2.57	4.4	4.3E – 02
8107375	TRAPPC8	Trafficking protein particle complex 8	6.8E-08	8022775	2.56	7.1	1.2E – 02
8136662	UGGT1	UDP-glucose glycoprotein glucosyltransferase 1	2.1E – 06	8045104	2.53	10.7	2.5E – 03
8169541	XRN2	5′—3′ exoribonuclease 2	2.7E – 05	8061333	2.46	6.4	1.6E – 02
8103951	ACSL1	Acyl-CoA synthetase long-chain family member 1	1.2E – 05	8103961	2.45	5.2	3.0E - 02
8172914	HUWE1	HECT, UBA and WWE domain containing 1	1.3E – 07	8172982	2.42	7.7	9.0E – 03
8061324	CAND1	Cullin-associated and neddylation-dissociated 1	5.1E – 06	7956914	2.42	8.1	7.4E – 03
8160213	ТТС39В	Tetratricopeptide repeat domain 39B	3.3E – 06	8160226	2.37	10.2	3.0E – 03
8107375	YTHDC2	YTH domain containing 2	9.5E – 06	8107388	2.35	12.7	1.1E – 03
8149986	ZNF395	Zinc finger protein 395	4.6E-08	8149998	2.32	8.1	7.4E – 03
8076455	CDK17	Cyclin-dependent kinase 17	5.8E – 06	7965654	2.25	8.2	7.2E – 03
8149986	USP34	Ubiquitin specific peptidase 34	1.5E – 10	8052505	2.20	7.7	9.0E - 03
8092933	SMG1	SMG1 homolog, phosphatidylinositol 3-kinase-related kinase (C. <i>elegans</i>)	5.0E – 07	7994025	2.18	5.0	3.2E – 02
8156321	TMEM131	Transmembrane protein 131	2.2E – 05	8054124	2.18	5.6	2.3E – 02
8158597	GPR155	G protein-coupled receptor 155	3.6E – 06	8056852	2.15	6.7	1.4E – 02
8086008	ADAM10	ADAM metallopeptidase domain 10	7.6E – 06	7989240	2.10	5.1	3.0E - 02
8092933	ACAP2	ArfGAP with coiled-coil, ankyrin repeat and PH domains 2	2.4E – 05	8092954	2.10	5.0	3.1E – 02

Table 7 (Continued)							
Transcript Gene Symbol Cluster ID ^{#,}		Gene Product	Interaction p	Exon ID	Fold-change	F	p
8051882	DENND2D	DENN	3.4E – 06	7918493	2.07	5.7	2.3E – 02
8156321	SYK	Spleen tyrosine kinase	2.0E – 06	8156330	1.93	13.3	8.7E – 04
8083523	GMPS	Guanine monphosphate synthetase	8.9E – 06	8083535	1.83	5.9	2.1E – 02
8134122	ΑΚΑΡ9	A kinase (PRKA) anchor protein (yotiao) 9	7.8E – 07	8134144	1.81	8.3	6.9E-03
8160213	TRIP12	Thyroid hormone receptor interactor 12	1.2E – 06	8059619	1.79	6.5	1.6E-02
8059596	EIF4G2	Eukaryotic translation initiation factor 4 gamma, 2	7.3E – 07	7946612	1.76	5.9	2.1E-02
8130151	APC2	Adenomatosis polyposis coli 2	3.6E – 05	8024308	1.75	7.0	1.2E – 02
8130151	RAET1E	Retinoic acid early transcript 1E	3.7E – 05	8130160	1.74	8.1	7.4E-03
8076077	CAPRIN2	Caprin family member 2	6.6E – 06	7962123	1.69	6.4	1.6E – 02
8103951	NUP88	Nucleoporin 88 kDa	4.9E – 05	8011838	1.68	5.6	2.3E – 02
8078569	MPHOSPH8	M-phase phosphoprotein 8	1.9E – 05	7967895	1.63	18.3	1.5E – 04
8052443	FAM175B	Family with sequence similarity 175, member B	2.7E – 05	7931218	1.61	6.8	1.3E-02
8097148	KIAA1109	KIAA1109	1.7E – 05	8097151	1.59	7.7	9.1E-03
8169541	DOCK11	Dedicator of cytokinesis 11	1.4E – 05	8169559	1.41	5.5	2.5E – 02
8022767	SMAP2	Small ArfGAP2	3.0E – 05	7900432	-1.12	6.1	1.9E – 02
8179298	DDX17	DEAD (Asp-Glu-Ala-Asp) box polypeptide 17	1.1E – 06	8076093	-1.12	5.6	2.4E – 02
8067563	MDM2	Mdm2 p53 binding protein homolog (mouse)	6.3E – 07	7957003	- 1.16	5.5	2.5E – 02
8096938	SMG1	SMG1 homolog, phosphatidylinositol 3-kinase-related kinase (C. elegans)	2.7E – 08	7999869	-1.20	14.1	6.6E – 04
8043251	GSTM5	Glutathione S-transferase mu 5	4.2E – 05	7903782	-1.57	18.5	1.4E – 04
8078569	GOLGA4	Golgin A4	5.0E - 06	8078597	-1.62	5.2	2.9E-02
8142307	LRPPRC	Leucine-rich PPR-motif containing	5.6E – 07	8051896	-1.77	10.4	2.8E-03
8179298	CSNK2B	Casein kinase 2, beta polypeptide	2.3E – 05	8179308	-1.83	7.8	8.4E-03
8024306	GSTM1	Glutathione S-transferase mu 1	4.5E-09	7903772	-2.52	21.4	5.3E-05

* Rows are sorted by decreasing fold-change in eventual PTSD cases relative to non-PTSD comparison subjects.

[#] Exons of Transcript Cluster IDs in bold comprised the optimal 20-probe SVM classifier of eventual PTSD status identified by training and testing in independent samples.

other treatments on post-deployment gene-expression profiles. Five of the 25 PTSD subjects reported using at least one psychiatric medication at the time blood samples were obtained, while none of the comparison subjects reported psychiatric medication use. It is plausible that betweengroup differences in medication use could account for some of the gene-expression differences observed between these groups. In order to account for this possibility, we performed a separate ANCOVA comparing non-medicated PTSD subjects (n = 20) and comparison subjects (n = 25). The removal of medicated subjects from the PTSD group produced only minor changes in ANCOVA fold-change values for the genes of interest; the average difference in fold-change value was <2%. Previous genome-wide expression studies have addressed this issue by using samples from PTSD subjects who were not currently medicated (Zieker et al., 2007; Yehuda et al., 2009; Neylan et al., 2011). Other studies have not explicitly address medication status (Mehta et al., 2011; Segman et al., 2005). However, if the ultimate goal is to develop gene-expression-based diagnostic classifiers that are robust to real-world variability, then the inclusion of medicated subjects may be valuable. Future studies should attempt to account for medication status and statistically control for its effect on gene-expression in order to identify genes that are specific to PTSD pathophysiology.

Because of our relatively small sample size and the severe corrections for multiple-testing required when examining the entire transcriptome, we did not detect individual gene-expression differences in PTSD cases that surpassed stringent criteria for declaring statistical significance. As such, the focus of our efforts and interpretations has been on groups of genes, either in regard to their biological annotations or their collective ability to identify PTSD cases. Nevertheless, one gene identified here as dysregulated has

18

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been identified previously in studies seeking to identify blood-based diagnostic biomarkers for PTSD. Prior to rigorous correction for multiple observations, Neylan et al. (2011) reported up-regulation of GSTM1 in PTSD cases, whereas we observed down-regulation of GSTM1 in PTSD cases. It is plausible that differences in subject characteristics or study design could account for the discrepant findings. Neylan et al. (2011) found increased GSTM1 expression in PTSD subjects compared to a non-trauma exposed control group. Perhaps these discrepant findings could make sense in the context of a model where increased GSTM1 expression reflects an adaptive response to traumatic stress and the attenuation of this response disposes some trauma-exposed individuals to developing PTSD., These studies also differed with respect to the time-span between disease onset and blood sample collection. Remarkably, GSTM1 and GSTM2 were identified as the lone predictors within a diagnostic classifier that achieved 80% accuracy in the test subset, and the down-regulation of GSTM2 was confirmed by QRTPCR. In previous work, we observed down-regulation of GSTM1 among these same subjects in samples taken prior to their deployment and the development of clinically significant PTSD symptoms; GSTM1 expression levels were also part of a pre-deployment predictor of subsequent PTSD diagnosis (Glatt et al., 2013). Members of this enzyme class function in the detoxification of electrophilic compounds including carcinogens, therapeutic drugs, environmental toxins and products of oxidative stress – by conjugation with glutathione. Down-regulation of genes whose proteins are responsible for the metabolism of reactive oxygen species (ROS) was also observed in lifetime PTSD cases with current symptoms (Zieker et al., 2007). The apparent link between ROS metabolism and PTSD may make more sense in light of previous in vitro studies demonstrating redox regulation of intracellular GR signaling. Specifically, reduced expression of antioxidant protein or direct administration of ROS negatively modulated GR signaling and resulted in reduced expression of glucocorticoid-induced genes; this effect could be rescued by the administration of antioxidant compounds (Makino et al., 1996; Okamoto et al., 1999). It is also interesting to note that other groups have found associations between GSTM1 polymorphisms and other brain disorders, including schizophrenia (Gravina et al., 2011; Rodriguez-Santiago et al., 2010), bipolar disorder (Mohammadynejad et al., 2011), and alcohol withdrawal symptoms (Okubo et al., 2003).

Despite our relatively small sample size and the additional levels of correction for multiple-testing required for exon analyses, a number of differentially expressed exons surpassed stringent criteria for declaring statistical significance. Additionally, the exon-based predictive classifier appeared to perform better than the gene-based predictive classifier. Taken together, these data support our previous findings (Glatt et al., 2013), suggesting that exon expression (indexing the activity of individual splice variants) may be more reliable and biologically informative than the expression of full-length ''genes'' or aggregated transcript clusters. Yet, we could not successfully recapitulate these array-derived results by QRTPCR, so further validation attempts must be made. Two of the dysregulated exons we identified by array analysis are components of genes DDX17 and FAM175B, which have been identified as differentially expressed in previous PTSD biomarker studies. Yehuda et al. (2009) observed up-regulation of *DDX17* among PTSD cases. Sarapas et al. (2011) observed down-regulation of *FAM175B* among current PTSD cases, but not lifetime PTSD cases or trauma-exposed comparison subjects. It is also curious to note that the list of alternatively spliced transcripts was enriched for acetylation-dependent protein catabolism and acetylation-regulated proteins more generally. Emerging evidence indicates that the acetylation of amino acids within non-histone proteins plays a role in regulation of cell metabolism (Choudhary et al., 2009; Yang and Seto, 2008). If the differentially expressed exons in PTSD are found to contain acetylation-dependent regulatory domains, then it is plausible that the PTSD proteome may be abnormally (hypo- or hyper-) responsive to acetylation.

19

It is interesting to compare the results of the present pilot study with our own previous work examining pre-deployment gene-expression profiles associated with subsequent development of PTSD after return from deployment (Glatt et al., 2013), as many of the same subjects (24 of 25 cases, 23 of 25 comparison) were featured in both studies. A number of whole-gene transcripts appeared dysregulated both prior to deployment (in those who would later develop PTSD) and after deployment (in current PTSD cases). AIM2 and EPSTI1 were up-regulated in both analyses (i.e., at both pre- and post-deployment), while RPL10A and GSTM1 were down-regulated in both analyses; these may reflect stable markers for PTSD vulnerability. Alternatively, they could have been dysregulated at pre-deployment assessment because pathophysiogical changes had already begun among these subjects, many of whom had previously been deployed to war zones (or because of other unmeasured factors, such as early adversity). LRTM2, on the other hand, was down-regulated in pre-deployment samples, but upregulated in post-deployment samples; this may reflect a maladaptive change or a compensatory yet ultimately ineffective change. Additionally, one putatively alternatively spliced transcript was identified in both pre- and post-deployment analyses. For MGAM, a similar pattern of dysregulated exon expression was observed in both analyses, with PTSD cases showing increased expression of several probes and the largest expression difference observed in a probe (8136700) spanning the 23rd and 24th exons. This may also reflect a stable marker for PTSD vulnerability.

The present pilot study broadens the search for diagnostic biomarkers for PTSD beyond that of previous work. Several studies have compared genome-wide transcriptional profiles between PTSD cases and controls, but to our knowledge, very few transcripts have been identified as dysregulated across studies using independent samples. Two studies with overlapping sample pools reported reduced expression of FKBP5 among current PTSD cases (Sarapas et al., 2011; Yehuda et al., 2009). A third study by Mehta et al. (2011) reported that the interaction of PTSD status and FKBP5 genotype (rs9296158) was associated with dysregulation of FKBP5 expression. Pre-deployment expression levels of FKBP5 were also shown to independently predict post-deployment PTSD symptoms (van Zuiden et al., 2012a). In the present microarray study, diagnostic status was not associated with differences in FKBP5 expression, suggesting a need to further consider heterogeneity in PTSD etiology. As discussed above, GSTM1 was originally identified as up-regulated among PTSD

DAVID category	Term	Count (%)	Fold-enrichment	p	Bonferroni- corrected (p)	Gene corresponding to dysregulated exon
GOTERM_BP_FAT	GO:0001775 \sim cell activation	15 (8.1%)	5.1	1.2E – 06	2.1E-03	BLM, SWAP70, STAT5B, KLRK1, MINK1, PF4, IGF2, SLAMF1, TGFB1, AZU1, CCND3, C140RF19, TREML1, F2R, CD7
GOTERM_BP_FAT	GO:0009611 ~ response to wounding	19 (10.2%)	3.5	6.6E – 06	1.1E – 02	LIPA, CCR1, STAT5B, CXCR1, PF4, IGF2, SOD1, CCL5, DTNBP1, TGFB1, GP9, AZU1, ORM1, CX3CR1, CTSB, PLA2G4C, TREML1, KDM6B, F2R
GOTERM_CC_FAT	GO:0030141 \sim secretory granule	10 (5.4%)	5.7	5.9E – 05	1.3E – 02	AZU1, SPAG8, PPBP, CPA3, PF4, SPARC, TREML1, RACGAP1, TGFB1, GP9
GOTERM_BP_FAT	GO:0001817 \sim regulation of cytokine production	11 (5.9%)	6.0	1.5E – 05	2.5E – 02	AZU1, IL18, TIA1, STAT5B, KLRK1, IGF2, PF4, TRIM16, SOD1, TGFB1, F2R
GOTERM_CC_FAT	GO:0031091 \sim platelet alpha granule	6 (3.2%)	11.0	2.0E - 04	4.1E – 02	PPBP, PF4, SPARC, TREML1, TGFB1, GP9
GOTERM_CC_FAT	GO:0000323 \sim lytic vacuole	10 (5.4%)	4.9	2.0E - 04	4.2E – 02	AZU1, LAPTM5, LIPA, MMD, IFI30, CPA3, CTSC, CTSB, ASAH1, GBA
GOTERM_CC_FAT	GO:0005764 \sim lysosome	10 (5.4%)	4.9	2.0E - 04	4.2E – 02	AZU1, LAPTM5, LIPA, MMD, IFI30, CPA3, CTSC, CTSB, ASAH1, GBA
KEGG_PATHWAY	hsa04060:cytokine-cytokine receptor interaction	13 (7.0%)	3.2	5.0E – 04	4.5E – 02	TNFRSF21, IL18R1, IL18, CCR1, CXCR1, PF4, CCL5, TGFB1, PPBP, CX3CR1, CSF2RB, IL2RG, IL3RA
GOTERM_MF_FAT	GO:0019955 \sim cytokine binding	8 (4.3%)	7.0	1.4E – 04	4.9E – 02	IL18R1, CCR1, CX3CR1, CSF2RB, CXCR1, IL2RG, TRIM16, IL3RA
Reactome	Term	Count (%)		р	FDR-corrected p	Gene corresponding to dysregulated exon
	Cytokine-cytokine receptor interaction	13 (7.0%)		1.0E – 05	3.0E - 03	IL18, TGFB1, PPBP, CX3CR1, CCL5, IL3RA, CXCR1, CSF2RB, TNFRSF21, CCR1, PF4, IL2RG, IL18R1
	Lysosome	7 (3.8%)		3.0E - 04	3.5E – 02	NAPSA, LAPTM5, ASAH1, CTSC, CTSB, GBA, LIPA
	Interleukin signaling pathway	5 (2.7%)		3.0E-04	4.7E – 02	IL16, IL18, STAT5B, IL3RA, CXCR1

cases by Neylan et al. (2011), but was found to be downregulated prior to deployment among US Marines who would later develop PTSD and down-regulated among current PTSD cases within the present analysis, which utilized many of the same subjects as Glatt et al. (2013). Also discussed above, the present study was the second to implicate DDX17 and FAM175B transcripts. The paucity of overlapping findings across studies may be accounted for by a number of factors, including the high risk for Type 1 error inherent when the number of subjects is small and statistical correction for multiple observations is inadequate. Other factors could also contribute to discrepant findings, including differences in trauma exposure (military combat, catastrophic event), sample type (PBMC vs. whole blood), the timing of sampling with respect to trauma exposure and disease onset, or differences in PTSD treatment effects across studies.

Comparing annotation-enrichment results across studies provided no additional consensus on the nature of geneexpression dysregulation in PTSD; we performed DAVID and Reactome analyses on individual transcript lists provided by each of the reviewed studies (Mehta et al., 2011; Neylan et al., 2011; Sarapas et al., 2011; van Zuiden et al., 2012a; Yehuda et al., 2009; Zieker et al., 2007), but few studies demonstrated significant enrichment of terms and no common terms were observed across studies. However, when these transcript lists were combined with the present data to create a single list, significant enrichment was observed for a number of terms related to cytokine signaling, lysosomal activity, and other immune-cell activities (Table 8). It is apparent that genes involved in cellular immunity are reliably and disproportionately represented among those that are dysregulated in PTSD cases. This is supported by a large body of evidence for dysfunctional cellular immune processes in individuals with PTSD, which we recently reviewed in depth (Baker et al., 2012b). Our review of the collective evidence suggests that systemic inflammation and deleterious health consequences in PTSD are strongly linked. Given this evidence, treatment strategies to reduce inflammation or modulate cell-mediated immunological processes may be of value to pursue in preclinical models of PTSD.

In conclusion, as the development of PTSD following initial trauma exposure remains quite variable and unpredictable, we sought to identify readily assessable biomarkers to aid in early diagnosis based on evaluations of bloodbased gene-expression among Marines participating in the MRS. Our analyses converged on a diverse group of genes and exons that appeared to be differentially expressed in peripheral blood cells from individuals with PTSD. Reduced expression of two genes involved in ROS metabolism were predictive of PTSD diagnostic status, while altered exon expression within a larger and more heterogeneous group of transcripts also predicted the PTSD diagnosis with apparently high accuracy. If blood-based biomarkers (such as the panels of genes and exons identified here) can be validated in additional cohorts exposed to a wider variety of traumatic stressors, then they may serve as useful adjuncts to the prevailing gold-standard behavioral diagnostic systems (Brewin, 2005; Ozer et al., 2003). Enabling clinicians to more confidently diagnose PTSD at earlier stages would be particularly important in groups such as these Marines, for whom it is known in advance that exposure to serious trauma is highly likely. This may also prove highly relevant for first-responders, such as police, fire, and emergency medical teams, for whom a regular part of their job is also exposure to potentially traumatic situations. Furthermore, blood-based biomarkers may help clinicians identify instances of determination of fitness for duty so that support services and limited resources can be made available to those individuals with the greatest need.

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Conflict of interest statement

The authors report no conflicts of interest

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Blood-based gene-expression biomarkers of post-traumatic stress disorder among deployed marines: A pilot study 23

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