ARTICLE

Neural correlates of conceptual-level fear generalization in posttraumatic stress disorder

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Posttraumatic stress disorder (PTSD) may develop when mechanisms for making accurate distinctions about threat relevance have gone awry. Generalization across conceptually related objects has been hypothesized based on clinical observation in PTSD, but the neural mechanisms remain unexplored. Recent trauma-exposed military veterans ($n = 46$) were grouped into PTSD ($n = 23$) and non-PTSD ($n = 23$). Participants learned to generalize fear across conceptual categories (animals or tools) of semantically related items that were partially reinforced by shock during functional magnetic resonance imaging. Conditioned fear learning was quantified by shock expectancy and skin conductance response (SCR). Relative to veteran controls, PTSD subjects exhibited a stronger neural response associated with fear generalization to the reinforced object category in the striatum, anterior cingulate cortex, amygdala, occipitotemporal cortex, and insula ($Z > 2.3$; $p < 0.05$; whole-brain corrected). Based on SCR, both groups generalized the shock contingency to the reinforced conceptual category, but learning was not significantly different between groups. We found that PTSD was associated with an enhanced neural response in fronto-limbic, midline, and occipitotemporal regions to a learned representation of threat that is based on previously established conceptual knowledge of the relationship between basic-level exemplars within a semantic category. Behaviorally, veterans with PTSD were somewhat slower to differentiate threat and safety categories as compared with trauma-exposed veteran controls owing in part to an initial overgeneralized behavioral response to the safe category. These results have implications for understanding how fear spreads across semantically related concepts in PTSD.

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INTRODUCTION

Fear learning can involve making inferences about threat associations, which are influenced by prior knowledge and experience [1]. Evolutionarily, such threat-related assessments facilitated survival such as predator–prey interactions, but have been co-opted along with other related brain circuits by modern-day psychological stressors [2]. Generalization of conditioned fear is a basic, cross-species associative-learning process whereby fear acquired to a conditioned stimulus (CS+), which reliably predicts an aversive unconditioned stimulus (US), transfers to safe stimuli that resemble the CS+ in some way. Although such threat processing is typically adaptive, posttraumatic stress disorder (PTSD) may develop when mechanisms for making accurate distinctions about threat relevance have gone awry.

In PTSD, experience with objects and events associated with a traumatic event may lead to an overgeneralized spread of fear. Neural circuits that process conditioned fears in PTSD, relative to trauma-exposed veteran controls, overgeneralize to related stimuli based on their overlap with perceptual attributes of the CS+, such as the emotional intensity expressed on a fearful face [3] or the size of a geometric shape [4]. Generalizing fear along a gradient of intensity in facial expression, for instance, yielded greater neural

responses in PTSD than trauma-exposed veteran controls in several brain areas, including the thalamus, locus coeruleus, insula, fusiform gyrus, and primary visual cortex, as well as enhanced functional connectivity between the amygdala and calcarine cortex [3]. When generalizing fear to images of circles that increase parametrically in size relative to the fear-conditioned stimulus, PTSD patients showed a stronger response than veteran controls in the caudate, anterior insula, left ventral hippocampus, dorsomedial prefrontal cortex (PFC), and dorsolateral PFC [4].

In clinical presentation, patients with PTSD may report symptoms are triggered by stimuli or situations that share a conceptual association, but little or no perceptual overlap with details of the trauma. For instance, symptom triggers in military PTSD might include idiosyncratic reminders of the event including ethnic food from the region they were deployed, anniversary dates, or military holidays, and even the topic of war in general. Concept-based symptom provocation provides a seemingly vast source of potential triggers that could lead to overgeneralization of threat association in PTSD that extend beyond cues that bear physical resemblance to trauma details [5]. Arguably, the spread of fear that is based on semantic relationships and conceptual knowledge may apply more

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directly to aspects of real-world generalization mechanisms in PTSD than low-level perceptual attributes. Although some prior work has examined perceptual generalization in PTSD [3, 4], and conceptual generalization in non-clinical individuals [6, 7], the neural substrates mediating conceptually based fear generalization in PTSD remain unknown. Modeling this aspect of threat generalization in PTSD is important because in some cases, learning trauma associations may take place with exposure to only a single traumatic event that later generalizes to other stimuli along abstract conceptual features.

The present study sought to determine the neural mechanisms by which fear learning generalizes through category-level representations of object concepts in PTSD. Our behavioral protocol [8] exploits established organizational separations in cortical representations of animate and inanimate objects. One group of participants learned with experience, through partial reinforcement, that images of animals predicted an aversive electrical shock but that tools were safe, whereas a separate group learned the opposite contingency. This innovative paradigm presents trial-unique conditioned stimuli to achieve fear learning based on participants' prior knowledge about the conceptual relationships among objects (animals from tools). Each participant viewed the same images chosen to vary widely in perceptual attributes, but different reinforcement contingencies determined which category attained threat relevance.

Our prior work in non-trauma-exposed young adults showed that conceptual fear generalization engaged distributed brain regions (including amygdala, insula, midbrain, cingulate gyrus, portions of the frontal lobe, and posterior category-selective cortex), which showed greater activity to exemplars from the aversively reinforced object category (CS+) than the unreinforced category (CS-) [9]. Moreover, the multivariate activity patterns of the items from the reinforced category were more correlated to each other (relative to those from the unreinforced category) in the amygdala and occipitotemporal cortex, suggesting an aversive learning-induced change in their representation. We showed that the object typicality of exemplars influenced left hippocampal activation whereby early trials were modulated by typicality more than late trials. At last, there was a temporally graded pattern of task-based functional connectivity between the hippocampus and amygdala that was stronger during early training trials but then dissipated in late trials [9].

Importantly, these neural circuits have been broadly implicated in functional studies of fear conditioning, symptom provocation, and emotion regulation in PTSD [10–13]. These studies generally report a hyper-reactive salience network in PTSD, including the amygdala, anterior cingulate cortex, and insula, coupled with poor contextual and cognitive regulation of this circuit via the hippocampus and PFC, respectively. Although it remains debated, structural abnormalities in some of these fronto-limbic regions may confer predispositions to develop PTSD following a trauma, in addition to trauma-induced pathological changes [14, 15]. Symptom provocation and fear conditioning studies further implicate interactions between these regions and posterior sectors of the brain that represent sensory features of environmental stimuli and more complex face, scene, and object representations [3, 16, 17].

Thus, we predicted that learning to generalize fears across object category exemplars in PTSD, relative to trauma-exposed veteran controls, would more extensively modify representations along the occipital temporal cortex, related association cortices, as well as areas implicated in simpler forms of fear conditioning, such as the amygdala. We further predicted that generalization of aversive learning to object categories would more strongly modulate functional connectivity between the amygdala and occipitotemporal cortex in the PTSD compared to trauma-exposed veteran controls.

MATERIALS AND METHODS

Participants

The participants ($N = 46$) were recruited from February 2014 through August 2016, from a repository of over 3800 US military veterans [18]. All the participants served since 11 September 2001 (9/11), and most were deployed to the military conflicts in Iraq and/or Afghanistan. All the participants provided written informed consent to participate in procedures approved by the Institutional Review Boards at Duke University Medical Center and the Durham VA Medical Center. Participants underwent screening for inclusion in the study. Exclusions included Axis I diagnosis (other than MDD or PTSD), contraindication to MRI, moderate/severe traumatic brain injury, past or current alcohol dependence, past or current substance dependence, current alcohol abuse, current substance abuse, neurological disorders, and age over 65 years. However, past alcohol or substance abuse was permissible. Abuse criteria met in the past 12 months constitutes current abuse. We administered the Clinician Administered PTSD Scale [19] based on DSM-5, Alcohol Use Disorders Test (AUDIT) [20], Combat Exposure Scale (CES), Childhood Trauma Questionnaire (CTQ) [21], Trauma Life Events Questionnaire (TLEQ) [22], Beck Depression Inventory (BDI-II) [23], psychotropic medication use based on self-report, and Drug Abuse Screening Test (DAST) [24]. Participants' demographic and clinical features are reported by diagnostic group in Table 1. Greater levels of combat exposure, lifetime trauma exposure, depressive symptoms, and serotonergic medication use, in the PTSD group compared with the control group are addressed in follow-up analyses (Supplemental Tables 1–5).

Stimuli

Images of the objects used in the study were obtained from the website www.lifeonwhite.com and from publicly available resources on the internet. The stimuli were the same as in our previous publication that contains a complete list of stimuli and healthy participants' category typicality ratings [8]. We avoided the use of highly threat-relevant images, such as weapons associated with combat trauma and animal exemplars associated with specific phobias (snakes/spiders), to mitigate the potential arousal bias evoked by these objects [25]. The presentation of stimuli was controlled using Presentation Software (Neurobehavioral Systems, Albany, CA, USA).

Procedure overview

The scanning session contained four phases: (1) a functional localizer of animal and tool exemplars, (2) pre-conditioning resting state, (3) fear generalization task based on conceptual categories of animals and tools, and (4) post-conditioning resting state. Twenty-four hours later, subjects returned for a recognition memory test and provided typicality ratings of the category exemplars.

Animal-tool localizer

The animal, tool, and phase-scrambled localizer contained 12 alternating blocks (separated by 11-s fixation cross) with 10 images/block presented for 750-ms (separated by a 250-ms blank screen). This phase was used to identify category-specific cortex independently of the fear generalization task itself. Localizer images were not reused for fear learning. Analysis of the category localizer included separate regressors for animals, tools, and phase-scrambled blocks. Single-subject contrast images for each event versus baseline were used for group-level random-effects contrasts of animals versus tools and tools versus animals. Activations were masked by (animals + tools) > phase-scrambled images. The fear learning response was interrogated within ROI spheres defined by category localizer. Subjects assigned to animal CS+ were analyzed separately from subjects assigned to tool CS+.

Table 1. Clinical and demographic features of sample.

Group	Control (<i>n</i> = 23) mean ± SD	PTSD (<i>n</i> = 23) mean ± SD	Statistic	<i>p</i> value
Age	38.65 ± 8.07	41.83 ± 8.37	<i>t</i> = 1.31	0.20
Race (AA), <i>n</i> (%)	10 (43.5)	12 (56.5)	$\chi^2 = 2.19$	0.34
White, <i>n</i> (%)	14 (60.8)	9 (39.1)	<i>n/a</i>	<i>n/a</i>
Multiracial <i>n</i> (%)	0 (0)	1 (100)	<i>n/a</i>	<i>n/a</i>
female gender; <i>n</i> (%)	3 (13.0)	7 (30.4)	$\chi^2 = 2.04$	0.15
CAPS (DSM-5)	1.39 ± 2.86	46.64 ± 32.03	<i>t</i> = 6.60	<0.0001
^a Trauma chronicity (y)	13.57 ± 10.14	14.43 ± 8.78	<i>t</i> = 0.30	0.86
BDI-II	5.25 ± 7.76	16.14 ± 12.25	<i>t</i> = 3.63	0.001
TLEQ	9.75 ± 8.21	19.77 ± 13.15	<i>t</i> = 3.13	0.003
Combat exposure	3.00 ± 5.47	8.23 ± 8.85	<i>t</i> = 2.43	0.02
Child trauma (CTQ)	63.83 ± 10.77	66.00 ± 10.71	<i>t</i> = .68	0.50
AUDIT	2.25 ± 1.62	4.09 ± 4.20	<i>t</i> = 1.93	0.064
DAST	0.25 ± 0.61	0.55 ± 1.18	<i>t</i> = 1.08	0.29
Serotonergic <i>n</i> (%)	0 (0)	12 (54.5)	$\chi^2 = 17.71$	<0.0001
Antipsychotic <i>n</i> (%)	0 (0)	0 (0)	$\chi^2 = 0$	1.0
Mood stabilizer <i>n</i> (%)	0 (0)	0 (0)	$\chi^2 = 0$	1.0
Animal CS+ <i>n</i> (%)	12 (52.2)	12 (52.2)	$\chi^2 = 0$	1.0
Right handed <i>n</i> (%)	22 (95.6)	21 (91.3)	$\chi^2 = 3.02$	0.22

^aTrauma chronicity is based on information provided by participants on the Trauma Life Events Questionnaire (#24) to indicate the number of years since the last occurrence of the trauma deemed to be the most distressing [23].
SD standard deviation, AA African American, AUDIT Alcohol Use Disorders Test, BDI-II Beck Depression Inventory II, CES Combat Exposure Scale, CAPS Clinician Administered PTSD Scale for DSM-5, CTQ Childhood Trauma Questionnaire, CS+ conditioned stimulus, DAST Drug Abuse Screening Test, TLEQ Trauma Life Events Questionnaire.

Fear conditioning task

Event-related fear conditioning during functional magnetic resonance imaging (fMRI) was achieved by presenting one category (e.g., animals) designated as CS+ (50% of exemplars co-terminated with the US) and the non-reinforced category (e.g., tools) served as the CS− (Fig. 1). Importantly, every trial contained a unique basic-level exemplar that was never repeated. Subjects were not instructed on the CS−US contingencies and were not informed that images would be presented only once in the MRI session. A 6-ms electrical shock applied to the right-wrist served as the US. Shock administration was controlled using the STM-100 and STM-200 modules connected to the MP-150 BIOPAC system (BIOPAC systems, Goleta, CA, USA). The level of the shock was calibrated prior to the start of the experiment using an ascending staircase procedure to reach a level deemed “annoying but not painful” by the subject [26].

Each run contained 20 trials (10 animals, 10 tools) presented in a pseudorandomized order, with no more than two objects from the same category occurring in a row. We used four different stimulus presentation orders to counterbalance presentation of animal and tool exemplars across participants (Fig. 1). Each trial was 6-s in duration, during which time subjects rated expectancy for US delivery on a scale anchored between 0 (certainty the shock will

not occur) and 100 (certainty the shock will occur) using an MRI-compatible joystick. Subjects were instructed that the final location of the rating bar at trial offset would be recorded as their response. A fixation cross was displayed for 10–12 s duration (average = 11 s) following the offset of each trial. One category (e.g., animals) was designated the CS+, and 50% of exemplars from this category co-terminated with the US. The other object category (e.g., tools) served as the CS−, and none of its exemplars were reinforced with a US. Category assignment was counter-balanced between PTSD and control groups, such that approximately half of the participants (*N* = 24; 12 PTSD, 12 veteran controls) were conditioned to animals, whereas the remaining subjects (*N* = 22; 11 PTSD, 11 veteran controls) were conditioned to tools. The selection of CS+ items that were paired with shock was random and counterbalanced between subjects.

Recognition memory test and typicality ratings

Twenty-four hours later, subjects returned to a different laboratory setting outside the MRI facility where they conducted a follow-up recognition test of memory for the CS+ and CS− items and rated each animal and tool picture for its typicality to the superordinate category. Category typicality ratings were made for each item on a 10-point scale (0 = highly atypical, 10 = highly typical). Participants provided memory ratings for all of the exemplars that were encountered as CS+ and CS− stimuli during fear learning, as well as a matched set of new exemplar foils belonging to the same superordinate categories (new tool/new animal), by indicating new or old and the confidence of this response on a continuous visual scale. In addition, participants rated whether the stimulus was followed by a shock to assess source memory. One subject was missing source memory ratings.

Skin conductance response (SCR) methods and analysis

Psychophysiological recording electrodes were placed on palmar surface of the left hand, and the data were analyzed as previously described [9] using procedures in the Supplemental Materials. We collected these data as an additional index of fear conditioning. We obtained usable data from 36 participants (19 veteran controls, 17 PTSD) who elicited non-zero shock responses. Ten additional participants were excluded from analysis: six participants had zero shock responses, and four participants lacked measurable data because of technical issues with the equipment, possibly owing to excessive noise associated with the scanner environment. The CS+ and CS− responses were baseline corrected, normalized to the shock response within subjects to account for possible group differences in overall arousal, and square-root transformed to reduce skewness.

fMRI parameters

High-resolution T1-weighted structural MRI scans were acquired with a 3D-FSPGR sequence on a General Electric 3 T MR750 scanner (1-mm³ voxels, TR/TE/flip angle = 7.84 ms/2.98 ms/12°, FOV = 256 mm², 166 slices, and 1 excitation). Functional images were acquired parallel to the anterior commissure-posterior commissure line using a SENSE spiral-in sequence (matrix = 64²; FOV = 256 mm²; 34 interleaved slices with no gap; 3.75 mm³ voxels; TR/TE/flip angle = 2000 ms/27 ms/60°). Functional imaging data were preprocessed using FSL version 5.0.9 (FMRIB Software Library; Oxford University, UK) [27] including brain extraction, motion correction, slice-timing correction, affine registration, 5-mm FWHM smoothing, and time series filtering <0.01 Hz to minimize low-frequency drift.

fMRI univariate and task-based functional connectivity analysis

Only fMRI data from the task period is presented; resting-state data were not analyzed.

The main analysis for interrogating the generalization response used whole-brain voxelwise analysis that contrasted responses to all exemplars from the reinforced category (CS+) versus those to

Fig. 1 Subjects were presented with 80 unique exemplars of animals (40) and tools (40) for 6 s each. Exemplars were never repeated during the fear acquisition session. Half the objects from one category (CS+) co-terminated with a 6-ms electrical shock unconditioned stimulus (US), whereas objects from the other category were never reinforced (CS−). Each trial was separated by a 6–8 s jittered inter-trial interval. Subjects were not instructed on the CS–US contingency and were not informed that images were presented only once. In this example, tools served as the CS+ and animals served as the CS− (category assignment was counterbalanced across subjects). Subjects learned through experience to generalize their conditioned fear response from the reinforced exemplars of the CS+ category (e.g., tools as shown here) to the other exemplars from the same object category, thus forming a broad threat response to the entire category.

the unreinforced category (CS−) between the PTSD and trauma-exposed control groups. The contrasts used whole-brain voxelwise regression with the FSL FMRI Expert Analysis Tool (FEAT). Shock delivery was modeled as a regressor of no interest (50% of CS+ trials) along with the customary six parameters specifying head motion. Statistical contrasts for CS+ > CS− trials (and the reverse contrast of CS− > CS+ trials) were corrected for multiple comparisons, which was performed on whole-brain data with the FSL two-step cluster threshold procedure to identify clusters of contiguous voxels with $Z > 2.3$ and then test the significance of resulting cluster(s) at a corrected threshold of $p < 0.05$ according to Gaussian random field theory [28, 29]. The results of the first-level contrast (CS+ > CS−) were subjected to between-group contrasts for PTSD > control and control > PTSD. For our second-level analysis, we used FMRIB's Local Analysis of mixed-effects (FLAME 1 + 2). Eklund and colleagues reported the false positive rate of FLAME 1 is similar to using a cluster-forming p value threshold of 0.001 [29]. All voxels from FLAME 1, which are close to threshold according to the selected contrasts and thresholds are processed with FLAME 2. This process involves implicit estimation of the mixed-effects variance using the Metropolis–Hastings Markov Chain Monte Carlo simulation. In response to peer-review, we compared the results of randomize and TFCE to FLAME 1 + 2. Randomize and TFCE yielded more voxels than FLAME 1 + 2, which was consistent with a recent report by Han and colleagues, who compared TFCE with AFNI ClustSim (method most similar to FLAME 1 + 2) [30].

Task-modulated functional connectivity measures were generated for each exemplar in the reinforced category using an interaction term between trial-specific onset times as the task regressor and the activation time course of the seed region as the physiological regressor. We adopted the generalized psychophysiological interaction (gPPI) approach, which provides improved model fit compared with standard PPI [31]. Mean and maximum connectivity strengths within each target region from the single-subject level were input into a model that contrasted groups (control, PTSD).

Representational Similarity Analysis (RSA)

RSA is one type of multi-voxel pattern analysis approach to fMRI in which a similarity structure is derived and compared across spatially distributed patterns of fMRI activity. Visual responses were compared between groups by considering the similarity structure of neural responses of exemplars from each object category [32]. See Deen et al. [33] and Supplemental Material for analysis details. In brief, we constructed a general linear model with each regressor containing a single trial. Voxels identified with the animal localizer were omitted for subjects who learned with an animal CS+ for the purpose of generating an ROI to test generalization to animal exemplars. Similarly, voxels identified with the tool localizer were omitted for subjects who learned with

a tool CS+ for the purpose of generating an ROI to test fear generalization of tool exemplars. Voxel patterns were computed in each subject for the entire occipitotemporal cortex, which is where distributed representations of these object categories reside, excluding voxels that showed a main effect of the CS+ category in the localizer task. This exclusive mask was created to avoid biasing the RSA results with voxels that reflect a main univariate category effect. Mean activation from each ROI provided trial-wise (80 × 80) representational similarity matrices (RSM) of correlations between all the possible pairwise combinations of the animal and tool exemplars for each subject. Kendall–Tau rank-order correlations were used to determine whether orderings were more consistent within diagnostic groups than between groups separately for animal–animal, tool–tool, and animal–tool exemplar correlations. Differences between PTSD–PTSD, control–control, and PTSD–control rank-order correlations and zero were calculated with exact permutation testing.

RESULTS

Clinical and demographic information

Detailed clinical and demographic information is reported by group in Table 1.

SCRs

A 2 × 2 (group × CS type) analysis of variance (ANOVA) was computed on the SCR responses in 36 (78.3%) participants (19 veteran controls, 17 PTSD) with usable data (see Methods). A significant main effect (Fig. 2a) of CS type ($F(1,34) = 14.701$, $p = 0.001$, $\text{partial-}\eta^2 = 0.302$) showed higher magnitude responses to the CS+ (mean = 0.380; SD = 0.248) relative to the CS− (mean = 0.282; SD = 0.222). The interaction of group and CS type was non-significant ($F(1,34) = 0.001$, $p = 0.991$, $\text{partial-}\eta^2 = 0.001$).

Shock expectancy

A 2 × 2 (group × CS type) mixed ANOVA was computed on the shock expectancy ratings averaged overall four trial runs (Fig. 2c). Greater shock expectancy ratings to the CS+, relative to the CS−, indicated successful fear conditioning ($F(1,44) = 61.233$, $p < 0.001$, $\text{partial-}\eta^2 = 0.582$). The main effect of group ($p = 0.415$), and interaction of group and CS type ($p = 0.103$) were non-significant (Fig. 2b). Importantly, the shock expectancy results were not confounded by significant group differences in shock voltage levels set by the participants themselves (PTSD subjects: mean = 51.204, SE = 4.173; veteran controls: mean = 43.936, SE = 3.329; $t(44) = 1.383$, $p = 0.17$).

Because most of the learning took place within Run 1, we further interrogated the temporal pattern of responding across early (first-five trials) and late (last-five trials) phases of Run 1 to acquire a fine-grained picture of initial learning rates. Linear trend analysis of Run 1 showed an increasing linear trend to the CS+

Fig. 2 Physiological and behavioral indices of conceptual fear generalization. **a** Mean skin conductance response (SCR) for CS+ and CS- in Control and PTSD groups averaged across all trials. **b** Compared with the trauma-exposed control group, the PTSD group demonstrated delayed contingency learning, as measured by shock expectancy ratings. On average, subjects' differential shock expectancy ratings began to diverge from the 2nd trial of each category exemplar in the control group but only from the 7th trial onward in the PTSD group. Linear trend analysis confirmed a strong linear pattern of threat learning (CS+) only in the PTSD group, whereas both groups showed strong linear patterns for safety learning (CS-). **c** All four runs confirm that learning was achieved in the first run and was maintained for the duration of the experiment. Dark green represents PTSD CS- and that light-red indicates PTSD CS+. Error bars represent mean $\pm$ SEM. * $P < 0.05$, two-tailed t tests. CONT = control.

shock expectancy ratings in the PTSD group ($r = 0.868$, $p = 0.0006$) but not the control group ($r = 0.152$, $p = 0.675$). By contrast, both groups showed decreasing linear trends in shock expectancy for the CS- (control: $r = 0.768$, $p = 0.001$; PTSD: $r = 0.843$, $p = 0.002$). The data were also analyzed in a $2 \times 2 \times 2$ (group $\times$ CS type $\times$ phase) mixed ANOVA (Fig. 2b). These results showed that learning progressed from early-phase trials to late-phase trials ($F(1,44) = 8.467$, $p = 0.006$, partial- $\eta^2 = 0.161$) and interacted at the non-significant trend-level with group ($F(1,44) = 3.776$, $p = 0.058$, partial- $\eta^2 = 0.079$). Expectancy ratings differed between groups in early ($t(44) = 2.770$, $p = 0.016$ partial- $\eta^2 = 0.149$; Bonferroni corrected), but not late ($t(44) = 0.092$, $p > 0.9$ partial- $\eta^2 < 0.001$; Bonferroni corrected), phases of Run 1. On average, PTSD patients were six trials slower to initially differentiate the CS+ from CS- based on shock expectancy ratings out of a total of 40 trials per trial type (80 total trials).

Recognition memory

Recognition memory for the CS items was measured by d' and analyzed with a 2×2 (group $\times$ CS type) mixed ANOVA. A significant influence of reinforcement during fear learning was found on subsequent memory discriminability ($F(1,44) = 7.956$, $p = 0.007$, partial- $\eta^2 = 0.156$), with greater recognition memory for CS+ (mean = 1.444, SE = 0.120) than CS- items (mean = 1.264, SE = 0.097). The specific object category assigned as the CS+ did not influence results ($F(1,44) = 2.377$, $p = 0.130$, partial- $\eta^2 = 0.051$). Diagnosis did not influence responding on this measure as either a main effect or in interaction with any of the other factors (p values > 0.05).

Source memory

Source memory for which items were paired with shock was analyzed with a $2 \times 2 \times 2$ (group $\times$ CS type $\times$ memoranda type) mixed ANOVA. A significant main effect of CS type (CS+ vs CS-; mean = 65.826, SE = 2.085 vs. 90.306, SE = 2.167) ($F(1,43) = 102.572$, $p < 0.001$, partial- $\eta^2 = 0.705$), a significant main effect of memoranda type (old vs new; mean = 71.688, SE = 1.799 vs mean = 84.444, SE = 2.033) ($F(1,43) = 65.354$, $p < 0.001$, partial- $\eta^2 = 0.603$), and a significant interaction between these variables ($F(1,43) = 65.797$, $p < 0.001$, partial- $\eta^2 = 0.605$). Examining this interaction, we found that the CS+ items were significantly more likely to be rated as paired with shock than the CS- items ($t(43) = 10.823$, $p < 0.001$ partial- $\eta^2 = 0.731$).

In addition, when examining only items presented as CS+ stimuli during fear learning (2×2 mixed ANOVA; reinforcement $\times$ group), we observed significantly higher endorsement of shock association for the items that were actually paired with a shock than those that were not, (paired with shock vs not paired with shock; mean = 46.093, SE = 4.014 vs mean = 62.625, SE = 3.737), ($F(1,43) = 35.859$, $p < 0.001$, partial- $\eta^2 = 0.449$). There was no main effect of diagnosis ($F(1,43) = 0.455$, $p = 0.504$, partial- $\eta^2 = 0.010$) and no interaction of shock pairing with diagnosis ($F(1,43) = 0.149$, $p = 0.702$, partial- $\eta^2 = 0.003$).

Typicality results

Subjects rated each exemplar (Supplemental Fig. 1) based on how representative the exemplar was of the superordinate category (animals or tools). There was a significant main effect of typicality ($F(1,43) = 10.615$, $p = 0.002$, partial $\eta^2 = 0.19$). Ratings for typicality were not influenced by shock pairing or diagnosis ($F(1,43) = 0.731$, $p = 0.397$, partial $\eta^2 = 0.017$).

fMRI results from the animal-tool functional localizer

Whole-brain animal-selective activation in lateral FFG and inferior occipital gyrus, extending into the posterior superior temporal sulcus, and tool-selective activation in medial FFG/parahippocampal gyrus and middle occipital gyrus (Supplemental Fig. 3 and Supplemental Material) formed an occipitotemporal ROI.

There were no between-group differences among subjects assigned to animal CS+ in either the animal- or tool-selective localizer-defined ROIs ($F(1,22) = 0.012$; $p = 0.912$) and no interaction of group and ROI ($F(1,22) = 0.735$; $p = 0.401$). Similarly, there were no between-group differences among subjects assigned to tool CS+ in either the animal-selective or tool-selective localizer-defined ROIs ($F(1,22) = 0.203$; $p = 0.657$) and no interaction of group and ROI ($F(1,22) = 0.458$; $p = 0.506$). Thus, PTSD was not associated with an altered response to broad stimulus categories with learned fear associations in visual perceptual brain regions.

Univariate fMRI results from the conditioning task

Activation for the effect of conditioned learning (CS+ > CS-) irrespective of CS identity (animal CS+ or tool CS+) was significantly greater in the PTSD than the trauma-exposed control group (Fig. 3 and Table 2) in the amygdala, ventral striatum particularly in the putamen, insula, anterior cingulate cortex (ACC), subgenual ACC (sgACC), and visual association areas including

Fig. 3 **fMRI correlates of conceptual fear generalization.** Data are averaged across all exemplars from the reinforced category compared with the unreinforced category (CS+ > CS-). **a** Relative to trauma-exposed veteran controls, PTSD patients exhibited enhanced activation in an extensive corticolimbic network, including the amygdala (amyg) [$x y z$] = 28, -2, -16, ventromedial prefrontal cortex (vmPFC) [$x y z$] = 6, 44, -14, striatum, particularly in the putamen (putam) [$x y z$] = 30, -8, -0, insula [$x y z$] = 38, -16, 12, rostral anterior cingulate cortex (rACC) [$x y z$] = 2, 36, -2, subgenual ACC (sgACC) [$x y z$] = -4, 18, -6, lingual gyrus (lingual) [$x y z$] = 16, -58, 0, and intracalcarine cortex (intracalcarine) [$x y z$] = 16, -82, 8. All results were significant with whole-brain correction (cluster threshold $Z > 2.3$; $p < 0.05$ corrected). **b** The bar graph offers a visualization of regional activation presented in the activation maps presented as Z scores. Error bars represent means $\pm$ SEMs.

Table 2. Brain regions involved in conceptual fear learning (CS+ > CS-) with greater activation in the PTSD group than the trauma-exposed control group.

Sub/cortical region	Number voxels 3.8 mm ³	Max Z statistic	Peak 3D coordinates		
			x	y	z
Rostral cingulate/vmPFC right	389	4.05	42	60	57
Putamen left	194	3.43	56	73	38
Planum temporale right	191	3.59	17	56	39
Frontal pole right	161	3.45	40	89	32
Posterior cingulate cortex left	148	3.33	48	40	50
Lingual gyrus/precuneus right	144	4.6	36	34	36
Postcentral right	144	3.33	31	47	67
Putamen right	140	3.68	29	59	34
Central operculum right	79	3.7	20	60	46
Precuneus left	51	3.47	58	35	41
Lateral occipital cortex left	41	3.01	68	24	47
Amygdala right	43	3.42	35	62	22
Intracalcarine cortex right	39	3.29	37	21	39
Right postcentral	39	3.26	28	52	57
Supramarginal right	35	3.04	21	51	64
Dorsal ACC left	30	3.51	49	89	47
Subgenual ACC right	24	3.61	41	71	31
Precentral right	162	3.43	44	49	68
Temporo-occipital fusiform left	23	3.17	65	33	37
Cuneus right	22	3.08	39	21	51
Amygdala right	21	3.81	30	62	29

All clusters are significant at $p < 0.05$ with whole-brain correction using FLAME 2.
vmPFC ventromedial prefrontal cortex, ACC anterior cingulate cortex.

lingual gyrus and precuneus. All results were significant after whole-brain correction (cluster threshold $Z > 2.3$; $p < 0.05$ corrected). The use of unique (non-repeated) exemplars for each CS+ presentation demonstrates that the neural response to the reinforced object category was the result of a generalized threat association to the entire conceptual category. Greater activation in the PTSD than veteran controls to the CS+ > CS- contrast, when each exemplar was modulated by its typicality rating, recapitulated the main fMRI results.

Effects of trauma exposure and clinical variables on neural activity
The pre-existing group differences in combat trauma, lifetime trauma, serotonergic medication use, and depression symptoms between the PTSD and Control groups (Table 1) meant that group differences in fMRI activation might be attributable to either PTSD or to one or more of these variables. Their inclusion as covariates offered inconclusive information because the severity of trauma exposure and diagnostic grouping are collinear and therefore difficult to interpret, according to Miller & Chapman [34]. The role of trauma/clinical variables on the univariate generalization response was further scrutinized vis-à-vis PTSD by assessing the strength of correlations between each covariate and activation separately in PTSD and Control groups in each of five key brain regions. Correlation of activation with lifetime trauma exposure (Supplemental Table 1), combat exposure (Supplemental Table 2), serotonergic medication (Supplemental Table 3), and depression symptoms (Supplemental Table 4), had non-significant interactions with diagnosis as assessed by the Fisher r-to-z test [35]. Re-analysis of the five brain regions with a covariate, generally revealed significant between-group differences in activation, as did the results without this covariate (Supplemental Table 5). Two exceptions were role of serotonergic medication on the insula and paracingulate gyrus, which produced a non-significant effect of diagnosis and a trend-level diagnosis effect on the amygdala. Finally, removing all subjects who were prescribed serotonergic medication from the analysis, yielded significant between-group differences in vmPFC ($p = 0.002$), putamen ($p = 0.01$), and trend-level differences in the paracingulate gyrus ($p = 0.095$), and insula ($p = 0.101$).

Functional connectivity results from the conditioning task

There was a significant temporal influence (across four runs) on task-related amygdala–hippocampal functional connectivity modulated by exemplar typicality ($F(1,44) = 6.954$, $p = 0.012$), but no main effect of group ($F(1,44) = 0.350$, $p = 0.557$) or no interaction of group and time ($F(1,44) = 0.025$, $p = 0.874$). Similarly, there was no interaction of diagnostic group and time for tool CS+ ($F(1,21) = 0.184$, $p = 0.906$), or animal CS+ ($F(1,21) = 0.076$, $p = 0.972$) exemplars. Functional connectivity between amygdala seed and target regions obtained from the GLM results revealed stronger category generalization (CS+ > CS−) in the PTSD group than trauma-exposed control group (Supplemental Fig. 2 and Supplemental Material) for left hippocampus ($t(44) = 2.552$, $p = 0.018$), and trend-level for sgACC ($t(44) = 1.835$, $p = 0.080$).

Representational similarity in occipitotemporal cortex

Visual responses were compared between PTSD and trauma-exposed control groups by considering the similarity structure of spatial patterns of neural response to category exemplars for both the animal and tool CS+ categories [32, 33]. The hypothesized higher dissimilarity in control-PTSD as compared to PTSD–PTSD and control–control was not demonstrated for animal CS+ (Supplemental Fig. 4a) and tool CS+ exemplars (Supplemental Fig. 4b) in the relevant perceptual cortex. Detailed results in Supplemental Material.

DISCUSSION

The present study extends knowledge about fear learning mechanisms in PTSD by demonstrating how acquired threat associations generalize across exemplars that belong to a conceptual category. Unlike traditional fear conditioning paradigms that examine repeated reinforcement to a single item, here a novel stimulus was introduced on each trial. In order to successfully predict reinforcement contingencies, participants had to generalize threat associations across the subset of reinforced exemplars to the superordinate category level. Modeling this aspect of threat generalization in PTSD is important because in some cases learning trauma associations may take place with exposure to only a single traumatic event that later generalizes to other stimuli along abstract conceptual features.

Patients with PTSD had a stronger generalization of the neural response than trauma-exposed control veterans in a fronto-limbic network including the amygdala, insula, striatum, ventromedial PFC, and rostral ACC. Although this network has generally been implicated in fear learning in PTSD [3, 4], here we show that its function extends to the generalization of fear to conceptual-level object categories. Activity in these regions was especially tuned to the more typical exemplars of the reinforced category in PTSD. In addition, we reveal enhanced functional coupling of the amygdala with the hippocampus to support category-based fear generalization in PTSD. Contrary to our hypothesis, univariate analysis showed no PTSD-associated differences in activation of the category-specific perceptual cortex during generalized fear learning of semantic categories. The multivariate RSA conducted on ventral occipitotemporal cortex also showed negative results associated with PTSD that were consistent with the univariate analysis. Nevertheless, whole-brain analysis successfully showed that PTSD patients more strongly engaged specific posterior areas to generalize their fear learning in the left temporo-occipital fusiform cortex, left lateral occipital cortex (superior division), cuneus, and right intracalcarine cortex (Table 2). Thus, generalization learning functions, which rely on visual and other constructive processes in these posterior areas, are biased by PTSD only in these circumscribed regions. This PTSD bias extends to visual association areas such as precuneus and lingual gyrus, which have established functions in the processing and early

interpretation of complex visual objects and scenes [36–38]. The present results extend our earlier work demonstrating greater engagement in patients with PTSD of the right fusiform gyrus and right intracalcarine cortex in generalizing fear from shock-reinforced fearful faces [3] to conceptual forms of generalization across object exemplars.

PTSD and trauma-exposed veterans showed similar enhancement shock expectancy ratings to the reinforced category, although the PTSD group was slower in their ability to behaviorally learn the category discrimination. In contrast to prior work on fear generalization in PTSD, where a single CS+ and single CS− are presented repeatedly during learning, here we present 40 unique stimuli of each trial type that were never repeated. Since shock expectancy ratings reflect, in part, declarative knowledge of the reinforcement parameters, this group difference may simply reflect impaired declarative aspects of early learning in PTSD on this complex paradigm. This is not an under-generalization effect in PTSD because inspection of the data by trial type shows that PTSD subjects gave higher shock expectancy ratings to the safe CS− compared with veteran controls on early trials, which partly accounts for a lack of early differential learning. Indeed, this observation could be leveraged as additional evidence for initial overgeneralization of shock expectancy to an innocuous cue (although such evidence should not be over-interpreted). Accordingly, linear trend analysis confirmed a strong linear pattern of threat learning only in the PTSD group, whereas both groups showed strong linear patterns for safety learning.

Both groups exhibited a similar boost in 24-hr delayed recognition memory for the exemplars from the reinforced category, and they both had intact source memory for the specific items that were followed by shock reinforcement. In combination with the shock expectancy ratings, these behavioral results help to rule out behavioral performance differences between groups as a potential confound when interpreting fMRI differences between groups. However, our results raise questions about the interpretation of group differences in fMRI activation that are not accompanied by more widespread behavioral differences. One possibility is that the magnitude of fMRI differences we observed in the PTSD group are too subtle to manifest in physiological learning indices and memory biases. A second possibility is that distributed brain processes are able to compensate for localized differences in activation to maintain a similar level of behavioral performance. Both hypotheses will require further investigation to confirm or disprove.

The stronger generalization of neural fear response in PTSD to broad conceptual categories such as animals and tools is fundamentally different than previous reports of generalization to perceptual gradients such as affect intensity of fearful faces [3] or the size of geometric shapes [4]. Generalization to semantically defined categories requires a higher level of abstraction that is based on binding the function and taxonomy of specific exemplars to an object category in contrast to the more basic generalization processes from these prior studies, which involve perceptual level binding of geometric dimensions and intensity of fearful facial expressions to threats. This level of abstraction during fear learning involves the recruitment of cortical association areas that are not required when learning to generalize fear associations to concrete perceptual attributes, as exemplified by broader engagement of posterior and midline cortical regions in the present study as compared to Morey et al. [3] and Kaczurkin et al. [4]. Although there are many possible interpretations of this activity, we speculate that our task placed higher demands in the PTSD veterans on relating threats to existing knowledge structures, which may recruit the vmPFC and related structures as part of a broader semantic/schema network [39, 40]. These demands differ from those in perceptual generalization studies in PTSD, for which the vmPFC exhibits an activation pattern that signals the generalization of the safety cue (CS−) [3, 4].

Generalization across abstract conceptual categories is a more naturalistic model for humans and bears greater clinical relevance to the environmental cues that trigger re-experiencing symptoms of PTSD, which typically possess only a vague connection to the original trauma material [5, 41]. Indeed, generalization based purely on perceptual gradients of intensity or magnitude have limited clinical relevance in which generalization phenomena may span emotional, spatial, semantic, perceptual, and cognitive constructs [6, 7]. The highly developed association cortices in humans, which represent an evolutionary rerouting of cortical information flow, may facilitate generalization across a broad range of stimuli that have conceptual similarities even when they lack a common set of perceptual features [42]. Thus, the ability to process complex conceptual information at an abstract level generally helps humans with more judicious threat assessment than other animals, but comes at a penalty when it elicits a rather broad fear generalization response.

Our understanding of the precise role of the amygdala in fear generalization is evolving. A delicate balance within subregions of the central nucleus of the amygdala appears to regulate fear generalization [43]. Neuronal activity in the medial portion of the central nucleus (the primary output pathway for initiating fear behaviors) is tonically inhibited by the lateral portion of the central nucleus. Accordingly, changes in tonic activity can tip the balance from expression of fear for only the CS+, toward generalization and expression of fear in response to the CS− or other related stimuli. In healthy humans, the amygdala has a key role in the generalization of fear along perceptual gradients [9, 44] in part owing to its functional connectivity to the hippocampus and higher-level sensory cortices. In PTSD, the amygdala has been implicated in perceptual forms of fear generalization across faces expressing different intensities of fear [3] and across circles varying in size [4]. However, these prior studies showed somewhat different patterns of amygdala connectivity in PTSD. In Morey et al. [3], amygdala connectivity with the calcarine (and thalamus at a trend-level) was greater in PTSD than veteran controls to generalized faces expressing more fear than the CS+, and amygdala connectivity to the vmPFC was greater in PTSD to generalized faces expressing less fear than the CS+. In Kaczurkin et al. [9], amygdala connectivity to the hippocampus was greater in PTSD than veteran controls to geometric stimuli resembling threats; other connectivity patterns were found in an activation seed that encompassed both the left amygdala and ventral hippocampus, so it is unclear whether the remaining connectivity patterns are in the amygdala proper and are not discussed further here. Our amygdala–hippocampal connectivity results are consistent with Kaczurkin et al. [4], but may reflect different processes. Our mechanistic study in healthy subjects [9] showed that hippocampal activity signaled the more typical category exemplars to provide a foundation of fear learning at the superordinate category level. We hypothesize that the hippocampus is providing information to the amygdala through enhanced functional coupling to the reinforced category. These processes are not engaged during perceptual forms of generalization, but the hippocampus may provide common functions for generalization across studies.

Computational modeling could be applied to yield valuable knowledge and insights about how stimulus value, prediction errors, learning rates, and attentional or associability parameters from different exemplar categories contribute to the clinical phenomenology of PTSD. Several design features of our experimental paradigm, however, limit the kinds of modeling that could be applied and inferences that could be drawn from such an endeavor. First, we intermixed reinforced and generalized stimuli during learning. This design feature precludes modeling learning parameters to a CS+ and then using those parameters to predict generalization during a subsequent test. In addition, we used trial-unique stimuli, which precludes applying

classic associative-learning models that track parameters based on a stimulus' prior reinforcement history. Finally, unlike prior fear generalization studies that manipulate a single perceptual feature, here the exemplars differ across numerous perceptual dimensions (color, shape, size) as well as conceptual knowledge (threat relevance, typicality) that would have to be jointly considered in terms of their relationship to learning parameters, and these relationships might differ across the two categories tested (animals, tools). Nonetheless, future studies should be designed specifically to test conceptual similarity gradients using Bayesian hierarchical modeling or other approaches that have been applied to mechanistically understand conditioned fear response generalization across a perceptual gradient [45].

Our results have implications for developing new exposure-based treatments to target overgeneralization of fear to conceptual categories [6, 7]. For instance, treatment for spider phobia with extinction therapy successfully generalized to help patients with untreated cockroach phobia to reduce subjective fear, approach distance, and psychophysiological arousal [46]. These treatments might focus on reducing fear reactivity not just to the specific triggers, but also semantically related stimuli. For example, in vivo exposure exercises could use stimulus categories partitioned according to semantic or other abstract feature dimensions and a hierarchy of feared stimuli, with exposures to stimuli resembling the feared stimulus added at each stage. In addition, patients could undergo discrimination training, whereby patients learn to differentiate trauma cues indicative of genuine danger from benign cues with inconsequential resemblance to the trauma.

Limitations

The reinforcement of a single category was accomplished by administering an electrical shock on 50% of trials, which raises the concern that activation was confounded by exposure to shock. However, this is unlikely because the stimulus onset preceded shock onset by 6-s, which is sufficiently long for separating the hemodynamic response between these two stimuli, and because the effect of shock was explicitly controlled in the data analysis. Nonetheless, there remains a possibility that unaccounted variance contributed to the signal contained in the CS+ > CS− fMRI contrasts. The pre-existing group difference in lifetime trauma exposure, combat exposure, serotonergic medication, and depression symptoms between the PTSD and control groups meant that group differences in outcomes could be partially attributable to either PTSD or other clinical variables such as trauma exposure; however, these variables lacked significant effects when explicitly tested for their contributions. Although the present results focused on animals and tools as conceptual categories owing to their known representation in the human brain, future work should examine other categories that have greater ecological validity for PTSD. Finally, future work should examine generalization at different levels of specificity within the hierarchy of categorical knowledge.

CONCLUSIONS

Fear in PTSD is associated with enhanced neural response to a learned representation of threat based on an established conceptual knowledge of the relationship between basic-level exemplars. PTSD may be one outcome of an evolutionary reality that misidentifying a dangerous stimulus as safe is far costlier to survival than incorrectly identifying a safe stimulus as dangerous. Real-world situations rarely involve simple unimodal stimuli, but rather involve complex objects and situations that can be represented across a variety of dimensions. One approach to capture the complexity of human fear generalization is to incorporate theoretical knowledge from other psychological disciplines that examine generalization of human knowledge,

including Bayesian models of inferential reasoning, categorization, and the organization of conceptual knowledge [47].

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AUTHOR CONTRIBUTIONS

R.A.M. wrote the manuscript, contributed to data analysis, interpretation of the data, and acquisition of data, gave final approval for the version to be published and is accountable for all aspects of the work including accuracy and integrity. C.C.H. made substantial contributions to data analysis, interpretation of the data, acquisition of data, and gave final approval for the version to be published. D.S. contributed to data analysis, interpretation of the data, and acquisition of data, gave final approval for the version to be published. M.A.M. Workgroup contributed to the acquisition of data, gave final approval for the version to be published. J.E.D. made substantial contributions to the conception and design, contributed to the interpretation of the data, gave final approval for the version to be published. K.S.L. made substantial contributions to the conception and design, interpretation of the data, and writing the manuscript, gave final approval for the version to be published, and is accountable for all aspects of the work including accuracy and integrity.

ADDITIONAL INFORMATION

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