Mechanisms contributing to the induction and storage of Pavlovian fear memories in the lateral amygdala

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The relative contributions of plasticity in the amygdala vs. its afferent pathways to conditioned fear remain controversial. Some believe that thalamic and cortical neurons transmitting information about the conditioned stimulus (CS) to the lateral amygdala (LA) serve a relay function. Others maintain that thalamic and/or cortical plasticity is critically involved in fear conditioning. To address this question, we developed a large-scale biophysical model of the LA that could reproduce earlier findings regarding the cellular correlates of fear conditioning in LA. We then conducted model experiments that examined whether fear memories depend on (1) training-induced increases in the responsiveness of thalamic and cortical neurons projecting to LA, (2) plasticity at the synapses they form in LA, and/or (3) plasticity at synapses between LA neurons. These tests revealed that training-induced increases in the responsiveness of afferent neurons are required for fear memory formation. However, once the memory has been formed, this factor is no longer required because the efficacy of the synapses that thalamic and cortical neurons form with LA cells has augmented enough to maintain the memory. In contrast, our model experiments suggest that plasticity at synapses between LA neurons plays a minor role in maintaining the fear memory.

[Supplemental material is available for this article.]
Impact of fear conditioning on the tone responsiveness of LAd neurons

There is a long-standing controversy in the field regarding the relative contributions of plasticity in the amygdala vs. CS afferent pathways to conditioned fear (Paré 2002). One view posits that thalamic and cortical neurons transmitting CS information to LA merely serve as sensory relays. Another stipulates that thalamic and/or cortical plasticity is critical for fear conditioning (for review, see Weinberger 2011). Thus, we first examined whether our model could reproduce the findings of Repa et al. (2001) when the CS responsiveness of thalamic and cortical neurons relaying tone information to LAd remained unchanged at 20 Hz (habituation level) through conditioning.

In these conditions (Fig. 3, empty triangles), the model could not reproduce the findings of Repa et al. (2001) (Fig. 3, solid circles). Indeed, whereas Repa et al. (2001) observed that 24% of LAd cells developed increased tone responses as a result of fear conditioning, only 9% (or 75 of 800) did so in the model experiment with thalamic and cortical tone-evoked firing rates unchanged at 20 Hz through conditioning. Moreover, when we considered cells that showed significant increases in tone responsiveness, as identified using the criterion of Repa et al. (2001), their average tone responses were much lower (<50%) than seen experimentally (Fig. 3). Various modifications to the model, including the distribution or strength of thalamic and cortical inputs, the Ca\(^{2+}\)-dependence of synaptic plasticity in LAd neurons, as well as the intrinsic connectivity within LAd, failed to alter this basic conclusion (Supplemental Fig. S2).

Since many studies reported that fear conditioning does increase the tone responsiveness of CS afferent neurons and that such changes play a critical role in the formation of conditioned fear memories (Quirk et al. 1997; Hennevin and Maho 2005; Letzkus et al. 2011; Weinberger 2011), we next ran the same “behavioral” protocol but increased the firing rate of thalamic inputs during the CS from 20 Hz to 40 Hz after one conditioning trial, and after six trials for cortical inputs, in keeping with prior experimental observations (Quirk et al. 1997). With this modification (Fig. 3, gray squares), the model successfully replicated the impact of fear conditioning and extinction training on the CS responsiveness of principal LAd cells, as described in Repa et al. (2001).

To assess the model’s ability to reproduce these earlier results (Fig. 4A1,A2, black curves), we relied on the criteria used by Repa et al. (2001) to analyze the behavior of LAd cells. That is, as a first step, model neurons were classified as plastic if their CS-elicted firing during any of the four trial conditioning blocks was higher than the habituation level using a t-test with a significance threshold of \(P < 0.05\). In a second step, a “persistence” value was calculated for each plastic cell by dividing the increase in CS responses during late conditioning (final eight trials) by that seen during early conditioning (first eight trials), both relative to habituation. As in Repa et al. (2001), plastic cells with persistence values of <0.75 were classified as “transient plastic” (TP) cells (Fig. 4A1), and those with persistence values > 0.75 were classified as “long-term plastic” (LP) cells (Fig. 4A2).

In the model, 24.7% (198/800) of principal LAd cells were plastic cells (LP and TP cells), which is similar to experimental observations (24/100; Repa et al. 2001) with 12% (96/800) being TP cells and 12.7% (102/800) being LP cells, again similar to experimental values (12% for each type; Repa et al. 2001). Moreover, whereas TP cells were largely found in LAdd (75 of 96) (Fig. 5), model LP cells were more widely distributed in LAd (47 and 55 of 102 in LAdd and LAdv, respectively) (Fig. 5), as again seen experimentally (5 and 7 of 12; Fig. 8 of Repa et al. 2001), and so providing validation for the model. Last, model TP and LP neurons reproduced the behavior of the corresponding cell types during extinction, as seen experimentally (Fig. 4A). That is, the tone responses of TP cells were at habituation levels except during the first block of extinction (\(P < 0.001\)) (Fig. 4A1). In contrast, LP cells retained higher levels of tone responsiveness throughout extinction (\(P < 0.001\)) (Fig. 4A2).

Although the above suggests that conditioning-induced increases in the CS responsiveness of thalamic and cortical neurons are required for the emergence of LP and TP cells (Figs. 3, 4B), it remains unclear whether the altered CS responsiveness of LAd neurons, once induced, remains dependent on an augmented signaling of the CS by thalamic and cortical neurons. To address this question, we ran the same simulation as above with the exception that the firing rates of thalamic and cortical CS inputs during the CS were returned to habituation levels during the recall test. Unexpectedly, this manipulation did not alter the average tone responses of TP cells (Fig. 4C1, squares) and LP cells (Fig. 4C2, squares), only causing a small and statistically insignificant decrease (10%, \(P = 0.33\)) in their tone responses. Overall, the above simulations indicate that increases in the CS responsiveness of thalamic and cortical neurons projecting to LAd are required for the development of the LAd plasticity but that, once induced, it is plasticity in LAd that mainly contributes to maintain the increased tone responsiveness of TP and LP cells.

Role of input connectivity in the formation of TP and LP cells

To determine whether the emergence of TP and LP cells was dependent on the particular connections they formed, we compared their intrinsic and extrinsic connectivity. Since TP and LP cells were more prevalent in LAdd vs. LAdv, respectively (Fig. 5; TP, red; LP, blue), we first compared the inputs to principal cells in
these two LAd regions. The reader is reminded that the model’s intrinsic connectivity was not set arbitrarily but adjusted to reproduce the pattern of intra-LAd connections inferred in a previous electrophysiological study (Samson and Pare 2006). This analysis revealed that the average ratio of excitatory to inhibitory connections for a LAd principal cell was significantly lower in LAdd (0.84 \pm 0.03) than in LAdv (1.1 \pm 0.04, \( P < 0.0001 \)). This difference was even more pronounced between TP cells in LAdd (1.07 \pm 0.05) and LP cells in LAdv (1.5 \pm 0.08).

However, TP and LP cells were similar in other respects. For instance, both types received at least one tone input and most received direct shock inputs (TP, 93%; LP, 96%). Yet, a higher proportion of the tone inputs originated from the thalamus for TP cells (86%) than LP cells (58%) and the opposite for cortical inputs (TP, 40%; LP, 91%; \( X^2 = 23.5, P < 0.0001 \)). Since previous unit recordings in rats have revealed that the CS responsiveness of auditory cortical neurons increases slowly during conditioning (Quirk et al. 1997), the latter difference probably contributed to the distinct temporal profiles seen in model TP and LP cells. Consistent with this, ablating cortical inputs to LAdv and LAdd neurons drastically reduced the tone responsiveness of LP cells (\( P < 0.001 \)) (Fig. 6A2), with only marginal effects on TP cells (\( P = 0.55 \)) (Fig. 6A1).

To test whether the differences in intrinsic connectivity between TP and LP cells play a role in shaping their spatial distribution and responsiveness, we carried out additional simulations using a random uniform connectivity (see details in legend of Fig. 6) or by selectively removing all connections from LAdd to LAdv (Fig. 6B). In TP cells, the random uniform connectivity resulted in an unchanged tone responsiveness during habituation but a marked increase by the end of conditioning (225 \pm 18%, \( P < 0.0001 \)) and late extinction (49 \pm 4%, \( P = 0.05 \)) (Fig. 6B1, empty circles and dashed line). In contrast, the tone responses...
of LP cells were reduced by 46 ± 3% (P < 0.05) during late conditioning and by 37 ± 3% (P < 0.01) during late extinction, with no significant change during habituation (Fig. 6B2, empty circles and dashed line). As explained below, these divergent changes in tone responsiveness of TP and LP cells were due to a shift in dorsoventral connectivity. Indeed, by converting the model’s connectivity from biology-based to random, the number of ventral to dorsal principal cell connections (which were sparse in the control case) increased, while dorsal to ventral connections (which were high in the control case) decreased.

In a second test, the model’s connectivity was returned to its initial state with the exception that we selectively removed all connections from LAd to LAdv. This resulted in a decrease in tone responses of LP cells during late conditioning (43 ± 3%, P < 0.001) and late extinction (36 ± 3%, P < 0.001), with no significant change during habituation (P = 0.2) (Fig. 6B2, triangles), and, as expected, no significant change in the tone responsiveness of TP cells (P = 0.79) (Fig. 6B1, triangles).

Last, we directly tested the hypothesis put forward by Repa et al. (2001), namely that TP and LP cells are involved in the initiation of learning vs. long-term storage of the fear memory, respectively. To this end, we ran simulations with the synaptic strengths of TP cells clamped at habituation levels (by disabling plasticity) during conditioning (Fig. 7A1). Although this manipulation only had a marginal effect on the number of LP cells (from 102 to 98), it greatly reduced their responses to the CS (Fig. 7A2) during late conditioning (41 ± 3% from control, P < 0.001) and late extinction (29 ± 3%, P < 0.001). Overall, these simulations support the model set forth by Repa et al. (2001) and indicate that the heterogeneous intrinsic connectivity of LAd plays a critical role in shaping the behavior of LAd neurons during fear conditioning.

Role of afferent and intrinsic LAd synapses in storing fear memories

Overall, the model experiments presented so far suggest that formation of auditory conditioned fear memories depends on several interacting factors, including (1) conditioning-induced increases in the tone responsiveness of afferent thalamic and cortical neurons, (2) a heterogeneous termination pattern of these cells in LAd, and (3) spatially differentiated intrinsic connections within LAd. However, the above tests did not address the question of which amygdala synapses are critical for storing the fear memory. As mentioned above, when the tone-evoked activity of thalamic and cortical neurons was artificially returned to habituation levels during the recall test, the number of plastic LAd cells as well as the magnitude of their tone-evoked responses were essentially unaltered (Fig. 4C). This indicates that a population of synapses within the amygdala maintains the memory.

Candidate synapses include those between tone afferents and LAd neurons as well as those between LAd neurons themselves. To identify the critical synapses, we next ran simulations where, after...
fear conditioning, the weight of various types of synapses was returned to habituation levels, one type at a time. We then examined the impact of these manipulations on the responsiveness of TP and LP cells when the CS was presented alone. As shown in Figure 7B, returning the weight of tone afferents to habituation levels had a dramatic impact, reverting the tone evoked responses of TP and LP cells to pre-training levels (paired t-tests: TP 66 ± 5%, P < 0.001; LP 80 ± 6%, P < 0.001). In contrast, the same manipulation at all amygdala synapses simultaneously caused a marginal increase in the tone responses of TP cells (26 ± 2%) and LP cells (11 ± 1%) (Fig. 7C). Only the former effect reached significance (paired t-test, P < 0.01). Importantly, restricting the latter tests to synapses involving plastic cells yielded the same results.

**Discussion**

We have developed a biophysically realistic model of LAd to shed light on fundamental, yet unresolved, questions regarding the mechanisms of Pavlovian fear memory formation, storage, and expression. These questions include the respective contributions of plasticity in LAd vs. CS afferent pathways, and the role of the intrinsic LAd network in the initiation and storage of fear memories. Below, we discuss the insights that emerged from our simulations.

**Initiation vs. storage of fear memory by different types of LAd neurons: contributing mechanisms**

During fear conditioning, the temporal profile of changes in CS responsiveness is not uniform in LAd (Repa et al. 2001). Some neurons, concentrated in LAdd, display augmented CS responses for only a few trials whereas others, typically in LAdv, show persistently increased CS responses that resist extinction. This led Repa et al. (2001) to speculate that the two cell types (transient [TP] vs. long-term [LP]) are involved in the initiation of learning vs. long-term storage of the fear memory, respectively. Our model could reproduce these observations including the two cell types, their incidence, as well as differential distribution in LAd. Our analyses revealed that a number of interacting factors contributed to the emergence of the two cell types in LAd.

**Selection of model parameters**

As cited in Materials and Methods, available biological data were used to constrain the single-cell models, connectivity, and the effects of neuromodulation, as in other similar models (e.g., Dyhrfjeld-Johnsen et al. 2007). However, synaptic strengths and plasticity parameters are not well understood presently. Two conditioning-induced factors affect afferent (thalamus/cortex to LAd) and intrinsic (within LAd) plasticity during conditioning: (1) change in excitability of thalamic and cortical cells and (2) change in neuromodulator concentrations. As we have demonstrated with the model, these two factors were critical for matching experimental data in Repa et al. (2001) for the numbers of plastic cells and the magnitudes of their tone responses. Various modifications to other parameters, such as distributions of afferents, plasticity parameters, or intrinsic connectivity within LAd, failed to reproduce the experimental data for the numbers of plastic cells and/or their tone responsiveness (e.g., Supplemental Figs. S2, S3) if only one effect was present. Furthermore, with both effects present, we did find a second model (model 2, see Supplemental Section S2.2) that was similar in all respects to the present model except for different intra-LAd parameters (initial intra-synaptic strengths and plasticity parameters). Model 2 outputs were also similar in all cases except for decreased inhibitory transmission in the amygdala after fear conditioning as has been reported in some experiments (Rea et al. 2009), in contrast to the small increase seen in model 1. Both thalamic/cortical plasticity and neuromodulator effects were critical for model 2 also. Taken together, all these constitute evidence that the network might not be inadvertently “tuned” in the process of development.
Intrinsic connectivity

Differences in the intrinsic connectivity of principal cells in LAdd and LAdv appeared to play a determining role in the emergence of TP and LP cells. Indeed, our model featured heterogeneous intrinsic connections based on the findings of a prior electrophysiological study (Samson and Paré 2006). As a result of this heterogeneous connectivity, the ratio of excitatory to inhibitory inputs to principal cells was lower in LAdd than in LAdv, and this difference was particularly marked between TP and LP cells. When this heterogeneous intrinsic connectivity was replaced with random uniform connections, the model could no longer reproduce the observations of Repa et al. (2001). LAdd to LAdv connections appeared to play a critical role in this respect since selective ablations of these connections greatly attenuated the CS responsiveness of LP cells. Moreover, in support of the notion that TP and LP cells are, respectively, involved in the initiation of learning vs. long-term storage of the fear memory, we observed that clamping the CS responsiveness of TP cells at habituation levels during learning blunted the increase in the CS responses of LP cells produced by fear conditioning.

Tone and shock inputs

Another aspect of the model’s connectivity that played a determining role was the distribution of tone and shock inputs. Compared to nonplastic cells, a higher proportion of TP and LP cells tended to derive their tone inputs from different sources: prevalently from the thalamus for TP cells vs. the cortex for LP cells. Since the CS responsiveness of auditory cortical neurons increases slowly during conditioning (Quirk et al. 1997), this difference is likely a major determinant of the contrasting temporal profiles of increased CS responsiveness seen in model TP and LP cells.

Contributions of the amygdala and CS afferent neurons to fear memory formation and storage

Despite incontrovertible evidence that fear conditioning leads to plasticity not only in the amygdala but also in CS afferent pathways (~25 papers from numerous laboratories), the role of thalamic and cortical neurons in fear memory formation is often overlooked in the literature (for review, see Weinberger 2011). In part, this is due to the difficulty of disentangling thalamocortical vs. amygdala contributions to fear memory. However, computational models allow experimentally impossible manipulations that can shed light on this question.

Here, we compared how our model LAd network behaved when the firing rates of thalamic and cortical neurons during the CS were clamped at habituation levels throughout the behavioral protocol or allowed to increase during conditioning, as observed experimentally. These simulations revealed that unless thalamic and cortical CS responsiveness increased during conditioning, the model could not reproduce learning-induced changes in the behavior of LAd neurons, as observed experimentally (Repa et al. 2001). While this suggests that an augmented thalamic and cortical CS responsiveness is required for the induction of amygdala plasticity, it is also required for fear expression after learning? To address this question, we allowed the CS responsiveness of thalamic and cortical neurons to increase during conditioning but returned it to habituation levels during the recall test. Unexpectedly, the CS responses of LAd neurons were marginally altered by this manipulation.

Overall, these results support the view that increases in the CS responsiveness of thalamic and cortical neurons are required for fear learning, but that later on they are no longer critical because the fear memory can be maintained by learning-induced changes in synaptic efficacy within the amygdala. To identify what populations of synapses are critical for storing Pavlovian fear memories, we next ran simulations where, after fear conditioning, the weight of various types of synapses was returned to habituation levels, one type at a time. This revealed that the critical sites of plasticity are tone inputs from the thalamus and cortex to LAd neurons. Indeed, suppressing plasticity at these synapses returned the tone responses of TP and LP cells to habituation levels. In contrast, abolishing plasticity at all synapses between amygdala neurons caused an unexpected increase in CS responsiveness. It may be that plasticity in the intrinsic LAd network leads to a potentiation of feedforward inhibition that limits the
size of the population of plastic cells. Such a phenomenon might contribute to maintain the specificity of fear memories. A challenge for future studies will be to test this prediction.

Finally, it should be noted that the model did not address the question of whether the amygdala is involved in altering the tone responsiveness of thalamic and/or cortical neurons in the first place. This possibility appears likely given the abundance of data indicating that the amygdala can facilitate synaptic plasticity in various structures of the brain either directly, or via the recruitment of modulatory systems (McGaugh 2002; Pare´ 2003).

The results of our model experiments lead to the following two predictions. First, that interference with plasticity of thalamic and cortical CS input neurons will prevent the emergence of plastic LAd cells and formation of the conditioned fear memory. Second, that if activity of CS input neurons is normal during training but returned to preconditioning levels during testing, the number and distribution of TP and LP cells, as well as the expression of conditioned fear, will be normal. While we recognize that testing these predictions will be challenging, recent advances in optogenetics and genetic engineering (see Supplemental Section S2.3) suggest that it will soon be possible.

Materials and Methods

We developed a biophysical model of LAd. A brief overview is presented here. See Supplemental Material, Sections S1–S3, for details.

Single-cell models

To reproduce the diversity of spike frequency adaptation seen in principal LA neurons (for review, see Sah et al. 2003), we modeled three types of regular spiking principal cells, with high (type-A), intermediate (type-B), or low (type-C) spike frequency adaptation, due to the differential expression of a Ca^{2+}-dependent K^-current. LA principal cells contain local GABAergic interneurons that exhibit various firing patterns, even among neurochemically homogeneous subgroups (Lang and Pare´ 1998; Rainnie et al. 2006; Sosulina et al. 2006; Woodruff and Sah 2007; Jansow et al. 2009). However, the majority displays a fast-spiking pattern, which was reproduced in the model.

The principal cell model had three compartments representing a soma (diameter 24.75 μm, length 25 μm), an apical dendrite (diameter 2.5 μm, length 296 μm) on which synapses were placed, and another dendrite (diameter 5 μm, length 119 μm) that helped match passive properties. Values of specific membrane resistance, membrane capacitance, and cytoplasmic (axial) resistivity were, respectively, R_m = 55 KΩ·cm^-2, C_m = 1.2–2.4 μF/cm^2, and R_f = 150–200 Ω·cm. Leakage reversal potential (E_L) was set to ~67 mV. The resulting V_rest was ~69.5 mV, input resistance (R_in) was ~150 MΩ, and t_m = 30 msec, which is within the range reported in previous physiological studies (Washburn and Moises 1992; Faber et al. 2001). All compartments had the following current sources: leak (I_L), voltage-gated persistent muscarinic (I_MA), high-voltage activated Ca^{2+} (I_Ca), spike-generating sodium (I_Na), potassium delayed rectifier (I_DR), and A-type potassium (I_A) (Li et al. 2009; Power et al. 2011). In addition, the dendrites had a hyperpolarization-activated nonspecific cation (I_h) current and a slow apamin-insensitive, voltage-independent afterhyperpolarization current (I_AHP) (Power et al. 2011). See Supplemental Sections S1.4 and S1.5 as well as Supplemental Tables S1 and S2 for current equations and densities.

Figure 1A shows the voltage response of the three principal cell models to depolarizing (two left panels) and hyperpolarizing (right panel) current injection. The three model cells could reproduce previous experimental observations (Sah et al. 2003) including the temporal dynamics of repetitive firing produced by membrane depolarization as well as their responses to membrane hyperpolarization.

The interneuron model had two compartments, a soma (diameter 15 μm, length 15 μm) and a dendrite (diameter 10 μm, length 150 μm). Each compartment contained a fast Na^+ (I Na) and a delayed rectifier K^+ (I_DR) current with kinetics (Durstewitz et al. 2000) that reproduced the much shorter spike duration that is characteristic of fast-spiking cells. The passive membrane properties were as follows: R_m = 20 KΩ·cm^-2, C_m = 1.0 μF/cm^2, R_f = 150 Ω·cm, and E_dr = −70 mV. As shown in Figure 1B, the interneuron model could reproduce the nonadapting repetitive firing behavior of fast spiking cells, as observed experimentally (Lang and Pare´ 1998; Woodruff and Sah 2007).

Network structure and connectivity

It was estimated that there are 24,000 principal cells in LA (Tuunanen and Pitkainen 2000). To keep computation times practical while capturing the complexity of the intra-LAd network, we modeled a scaled down (30:1) version of LA that included 800 principal cells (types A, B, and C in the ratio 50:30:20). Because the proportion of interneurons to principal cells is 20:80 (McDonald and Augustine 1993), the model included 200 interneurons. Principal cells and interneurons were distributed randomly in a realistic tridimensional representation of the horn-shaped LA (Fig. 2A).

By comparing the responses of LA cells to local applications of glutamate at various positions with respect to recorded neurons, Samson and Pare´ (2006) inferred general principles of connectivity among principal cells, as well as between local-circuit and principal neurons. In particular, Samson and Pare´ (2006) determined that excitatory connections between principal cells prevalently run ventrally and medially with significant lateral divergence. In contrast, inhibitory connections prevalently run mediolaterally in the horizontal plane and have no preferential directionality in the coronal plane. Samson and Pare´ (2006) also recognized that principal LA neurons located along the external capsule (in the “shell” region of LA) form different connections than those found more medially (in the “core” region of LA; shell thickness of 100 μm). In the shell region, inhibitory neurons only affect nearby principal neurons whereas excitatory connections between principal cells are spatially less limited. Although not providing precise connectivity data, this information could be used to infer critical estimates about directionality and ratio of excitation to inhibition. Using the directionality information from the Samson and Pare´ (2006) study, and a third of the connectivity numbers, a model principal cell had, on average, 21.4 mono- and 40.6 disynaptic excitatory inputs, and 22.2 monosynaptic inhibitory inputs. These average figures result from the implementation of directionally heterogeneous connections, as described below.

Coronal plane

Within a 100-μm coronal slice, principal shell neurons excite shell cells located 300–400 μm more ventrally with a 30% probability (note that these are for monosynaptic connectivity) (Fig. 2B1). Core to shell connections occur with a much lower probability (2%). In addition, principal shell neurons are inhibited by more dorsally located interneurons (23% connectivity for cells within 300 μm) (Fig. 2B2). In the core region, excitatory connections between principal cells have a greater extent in the lateromedial direction (50–300 μm, 2%–6% connectivity) (Fig. 2B3) than in the mediolateral direction (50–200 μm, 5% connectivity) (Fig. 2B3), whereas inhibitory connections have similar strengths in all directions (interneurons formed inhibitory inputs with 10%–24% of principal cells at a distance of 50–600 μm) (Fig. 2B4).

Horizontal plane

Within a 100-μm horizontal slice, connections were set in the following manner. Connection probability increased with distance for lateromedial connections, and the opposite for mediolateral connections (see details in Fig. 2B5). As to inhibitory connections, they prevalently run in the mediolateral direction with 8%–20% connectivity in the range 50–600 μm and 5%–20% connectivity...
in the lateromedial direction within a distance of 50–600 μm (Fig. 2B6). Principal cells project to all interneurons within a spherical radius of 100 μm.

Activity-dependent synaptic plasticity

Fear conditioning induces changes in the efficacy of synapses conveying CS inputs to LA (McKernan and Shinnick-Gallagher 1997; Rumpel et al. 2005). Also, activity-dependent long-term potentiation (LTP) can be induced at thalamic (Huang et al. 2000; Tsvetkov et al. 2004; Humeau et al. 2005) and cortical (Huang and Kandel 1998; Tsvetkov et al. 2002; Humeau et al. 2003, 2005) inputs to principal LA neurons. Furthermore, it has been shown that fear conditioning and LTP share a common set of mechanisms (Miserendino et al. 1990; Kim et al. 1991; Campeau et al. 1992; Huang and Kandel 1998; Bauer et al. 2002), and that fear conditioning occludes LTP of cortical inputs to LA cells (Tsvetkov et al. 2002). These findings, coupled to the fact that intramygdala infusion of NMDA receptor antagonists blocks the induction of conditioned fear in vivo and of LTP in vitro, provided the basis for the hypothesis that NMDA receptor-mediated LTP represents a cellular substrate of fear conditioning (for review, see Pape and Paré 2010). Induction of LTP at thalamic and cortical inputs to principal cells has been found to depend on post-synaptic depolarization allowing influx of Ca\(^{2+}\) ions via NMDA receptors (Bauer et al. 2002; Tsvetkov et al. 2002) and/or voltage-dependent L-type Ca\(^{2+}\) channels (Weisskopf et al. 1999; Bauer et al. 2002; Humeau et al. 2005). In addition, it was reported that excitatory glutamatergic synapses from the thalamus or cortex onto interneurons exhibit NMDA-receptor-dependent potentiation (Bauer and LeDoux 2004). This potentiation is also AMPA-receptor-dependent because AMPA receptors on inhibitory neurons lack the GluR2 subunit, making them calcium-permeable (Mahanty and Sah 1998). It was further shown that inhibitory inputs onto pyramidal cells are modifiable via a Ca\(^{2+}\)-dependent mechanism (Bauer and LeDoux 2004). In addition to LTP, long-term depression (LTD) can be readily induced at excitatory amygdala synapses by low-frequency stimulation of the lateral nucleus at 1 Hz for 15 min (Wang and Gean 1999). To reproduce these experimental findings, all AMPA synapses in the model were endowed with long-term postsynaptic plasticity except for those delivering shock or background inputs. Also, all GABA synapses had long-term plasticity. This form of plasticity was implemented using a learning rule that uses the concentration of postsynaptic calcium at each modifiable synapse (Kitajima and Hara 1997; Shouval et al. 2002; Li et al. 2009). Calcium entered postsynaptic pools at excitatory synapses via NMDA receptors (and AMPA receptors for interneurons) and voltage-gated calcium channels (VGCCs). Similarly, calcium for pools at inhibitory synapses came from postsynaptic intracellular stores and VGCCs (Li et al. 2009). For both types of synapses, the synaptic weight decreased when the calcium concentration of the pool was below a lower threshold and increased when it exceeded an upper threshold. Equations and details related to the learning rule are provided in Supplemental Section S1.6.

All model AMPA and GABA synapses also exhibited short-term presynaptic plasticity, with short-term depression at interneuron to principal cell and principal cell to interneuron connections modeled after the experimental findings of Woodruff and Sah (2007) in the basolateral amygdala. In the absence of experimental data about the short-term dynamics of synapses between principal LA neurons, we relied on data obtained between regular spiking cortical cells to implement short-term frequency-dependent depression (Markram et al. 2004). For convenience in modeling, these were implemented by multiplying the synaptic conductances of the relevant AMPA and GABA synapses after each spike with appropriate frequency-dependent factors (Varela et al. 1997). See Supplemental Section S1.7 for equations.

Neuromodulator effects

Neuromodulators have long been implicated in fear and anxiety, and are known to regulate Pavlovian fear learning and synaptic plasticity in LA (Tully and Bolshakov 2001; Bissiere et al. 2003). Conditioned aversive stimuli alter the activity of ventral tegmental area and locus coeruleus neurons (Feenstra et al. 2001), which in turn modulate fear and anxiety through their widespread forebrain projections, including to the amygdala (Guarracci and Kapp 1999). Therefore, the model incorporated the effects of dopamine (DA) and norepinephrine (NE) on LA cells, based on prior experimental reports, as outlined in Supplemental Section S1.1.

Inputs

LA receives auditory inputs directly from the posterior intralaminar nucleus and the medial sector of the medial geniculate nucleus (LeDoux et al. 1985; Turner and Herkenham 1991; Shimonaga et al. 1994; Linke et al. 2000; Woodson et al. 2000) as well as indirectly via temporal auditory cortical fields (Mascagni et al. 1993; Romanski and LeDoux 1993a,b; Shi and Cassell 1997). Moreover, the same posterior thalamic regions that send auditory inputs to LA also receive inputs from the spinothalamic tract (LeDoux et al. 1987) and probably send convergent CS and US inputs to LA. Accordingly, the model was endowed with CS and US inputs represented as glutamatergic synapses acting at AMPA and NMDA receptors. Based on the experimental literature, we modeled two types of CS inputs: thalamic and cortical, both at 20 Hz during habituation and 40 Hz after conditioning (Bordi and LeDoux 1994; Quirk et al. 1997). These two tone inputs were differentially distributed in LAd, as described below.

Background synaptic inputs

LA projection neurons have low spontaneous firing rates in control conditions (Gaudreau and Paré 1996). To reproduce this, Poisson-distributed, random excitatory background inputs were delivered to all model cells, resulting in average spontaneous firing rates of 0.7 Hz for principal cells and 7.2 Hz for interneurons.

Tone and shock inputs

Auditory fear conditioning is commonly thought to depend on the convergence of inputs relaying information about the CS (tone) and US (footshock) in LAd (for review, see Pape and Paré 2010). In the model, the CS and US inputs were represented by glutamatergic synapses acting via AMPA and NMDA receptors. The frequency of thalamic and cortical tone inputs during habituation was set to 20 Hz (Quirk et al. 1997). The tone inputs also included noise represented by random Poisson spikes with an average frequency of 2 Hz. The density of both cortical and thalamic tone inputs to LAd was determined iteratively (see Supplemental Section S1.2). The following distribution of inputs was used for the simulations described in the Results section: uniform total tone density throughout LAd with 70% of the LAd cells receiving thalamic and 35% receiving cortical tone projections, and the opposite for LAdv, i.e., 35% of LAdv cells receiving thalamic and 70% receiving cortical tone projections. The shock inputs continued to be distributed uniformly to 70% of LAd cells.

Conditioning protocol used in simulations

The schedule of tone and shock inputs in the simulations was based on in vivo studies (Quirk et al. 1995; Repa et al. 2001). We scaled down the timing of the auditory fear conditioning protocol so that it would be suitable for computational studies (Fig. 2C). The protocol included three phases (habituation, conditioning, and extinction), comprised of 8, 16, and 20 trials, respectively. Each trial featured a 0.5-sec tone CS followed by a 3.5-sec gap. Only during conditioning, a shock was administered 100 msec prior to the end of the tone, so that they co-terminated. In light of evidence that fear conditioning leads to plasticity in CS afferent pathways (for review, see Weinberger 2011), the frequency of thalamic and cortical tone inputs was increased to 40 Hz after the first and sixth conditioning trials, respectively. Simulations were performed using the NEURON modeling package (Carnevale and Hines 2006).
Model tuning and validation

To obtain the full model, several aspects had to be developed: the single-cell models, the properties of glutamatergic and GABAergic synapses, the intrinsic connections between LAd cells, plasticity of these connections, neuromodulatory inputs, and, finally, extrinsic inputs conveying shock and CS information. For each of these aspects, the following approach was used. (1) The experimental literature was mined, both to constrain the modeling and derive criteria to assess whether the model successfully reproduced the particular phenomena being modeled. (2) Iterative “tuning” of model parameters was performed until the model’s behavior matched experimental observations. (3) Validation of the model took two forms. First, the ability of the model component to reproduce earlier experimental observations, as described in the previous sentence. Second, the ability of the model component to reproduce a set of experimental observations it was not designed to reproduce.

Here, it should be noted that because the model includes 1000 neurons connected by nearly 40,000 synapses, it was not possible to tune each component to achieve reproduction of the results of Repa et al. (2001). For instance, development of the single-cell models was based on biophysical studies of the intrinsic properties of LAd neurons. The design of single-cell models aimed to reproduce these observations without forethought as to the consequences of the model’s properties for fear conditioning. Similarly, the 40,000 synapses connecting these cells were adjusted individually in anticipation of their potential impact on the responsiveness of LAd neurons during fear conditioning. Rather, probabilistic gradients of connectivity were implemented to reproduce the findings of an in vitro study that inferred how the intrinsic LA network is organized based on responses evoked by local pressure applications of glutamate (Samson and Paré 2006). Surprisingly, the approach used to adjust properties of synaptic plasticity did not aim to predispose the network to reproduce the differential distribution of plastic cells reported in Repa et al. (2001). Instead, properties of plasticity were adjusted so that the synapses could undergo activity-dependent plasticity, as observed experimentally, while maintaining overall network stability by avoiding runaway behavior. Overall, ~75% of the model was based on lower-level details, very remote from the higher levels of conditioning-induced tone responses. Therefore, model tuning up to the point of assembling the entire network could not be done to achieve a specific impact on higher level responses.

The assembled network model was then tuned in two steps. In the first step, the weights of the synapses (tone, shock, intrinsic connections) and the background input parameters were adjusted to reproduce experimentally observed spontaneous firing rates for principal cells and interneurons. In the second step, the model was subjected to the training protocol (Fig. 2C), and the plasticity (learning in AMPA and GABA synapses) and neuromodulation (Dopamine) parameters used in the first step were adjusted to reproduce both the numbers of plastic (TP and LP) cells and their tone responses as reported in experiments (Repa et al. 2001). Adjustments to parameters at this step were consistent with biological bounds where available in the literature (see Supplemental Tables S1–S4). Note that step 2 does not affect step 1. Iteration was involved between the two steps. Surprisingly, the tuning was not difficult and the model output matched data in Repa et al. (2001) very well. Specifically, average values of model tone responses in trial blocks were within 98% of the experimental values (Fig. 4A1,A2). Also, the fraction of model plastic cells (198/800 = 25%) compared well with experimental estimates (24%). Given the tuning process described above, we feel that it would not have been possible to predict that the model would reproduce these experimental observations, let alone so closely.

We validated the full model by testing its ability to reproduce a set of experimental observations it was not designed to reproduce: the impact of blocking NA and DA receptors on the acquisition of conditioned fear (Bissiere et al. 2003; Rodrigues et al. 2009). That is, we blocked DA and NE effects individually, and found that both interfered with the acquisition of fear memory in LAd (Supplemental Fig. S4), consistent with experimental reports (Bissiere et al. 2003; Rodrigues et al. 2009). As additional validation of the model, we found that the distribution of model TP and LP cells (Fig. 5) matched experimental reports in Repa et al. (2001) even though we performed no model tuning to achieve this effect. Furthermore, to ensure that the model outputs were consistent across random runs using the same network size, we ran three additional cases with different (1) distribution of cells (types A, B, C) in the 3D space, (2) connectivity, and (3) delays, all selected with different random seeds. All these cases yielded consistent numbers for the total plastic cells in the control case (avg. 196 ± 6), and for the individual cell groups: TP cells (93.6 ± 4.1), LP cells (101.7 ± 1.7). Moreover, all the other behaviors discussed were similar, suggesting that the model output is robust to all these parameters.

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References


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