Enhancing associative learning in rats with a computationally designed training

protocol

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Abstract

Associative learning requires the activation of protein kinases with distinct temporal dynamics. Learning protocols with computationally designed intertrial intervals (ITIs) that maximize the interaction between fast-activated protein kinase A (PKA) and slow-activated extracellular signal-regulated kinase (ERK) enhance nonassociative learning in *Aplysia*. Here, we tested whether an optimal learning protocol with irregular ITIs, predicted computationally to increase the overlap between PKA and ERK signaling in rat hippocampus, would enhance associative learning in mammals. We simulated ~1000 training protocols with irregular ITIs to identify this optimal protocol, predicted to induce stronger associative learning than standard protocols with fixed ITIs. With auditory fear conditioning, we showed that male adult rats exposed to the optimal conditioning protocol exhibited stronger fear memory retrieval and impaired fear memory extinction, compared to rats that received either massed or spaced conditioning protocols with fixed ITIs. With fear extinction, we likewise observed that fear conditioned rats exposed to the optimal extinction protocol showed improved extinction of contextual fear memory, compared to rats that received standard extinction protocols. Immunohistochemical staining confirmed that the optimal conditioning protocol increased phosphorylated cAMP responsive element binding (pCREB) protein levels in the dentate gyrus (DG) of the dorsal hippocampus. These findings demonstrate the capacity of a behavioral intervention driven by a computational model of memory-related signaling pathways to enhance associative learning in mammals, and may provide greater insight into strategies to improve cognition in humans.

47 **Keywords:** Associative memory, fear conditioning, fear extinction, long-term potentiation,

48 learning model, spaced learning.

Significance Statement

To test if associative learning in mammals can be enhanced by predetermined training

protocols, we used a computational model of the dynamics of essential biochemical

cascades for learning and memory to predict an optimal training protocol in rats. Using

auditory fear conditioning and fear extinction tests, we demonstrated that the optimal

protocol enhanced the acquisition and the extinction of fear memories. Immunostaining

confirmed that the optimal protocol induced greater levels of pCREB in rat hippocampus.

Our results suggest that a non-invasive behavioral intervention guided by computational

modeling can be used to enhance associative learning in mammals, with potential

relevance for facilitating memory formation as well as enhancing extinction-based

therapies in humans.

INTRODUCTION

Long-term memory (LTM) formation that lasts for days to years is believed to be mediated by synaptic plasticity including long-term potentiation (LTP) or its invertebrate analogue long-term facilitation (LTF), which require gene expression and protein synthesis (Martin et al., 2000; Kandel, 2001; Lynch, 2004; Alberini, 2009; Byrne and Hawkins, 2015). Studies in the last decades have investigated LTP/LTF and their underlying molecular processes as potential targets to enhance learning or restore memory deficits in laboratory animals. However, traditional interventions using systemic cognitive enhancers or intracerebral pharmacological manipulations (Sharif et al., 2021; Lauterborn et al., 2016; McGaugh and Petrinovich, 1965; Lynch et al., 2014; Fernandez et al., 2008) are either based on trial-and-error approaches in specific model systems or are highly invasive, making the translation of such interventions difficult for human applications. The recent development of neural recording and optogenetic techniques has enabled precise control of neurons based on their intrinsic firing rates in order to enhance learning or modify memories (Brown et al., 2012; Nabavi et al., 2014; Lee et al., 2017; Liu et al., 2012), but the use of these techniques in humans remains unforeseeable in the near future.

An alternative approach to enhance learning and memory is to develop computational models to predict optimal training protocols based on the intracellular molecular cascades that underlie LTM formation and LTP/LTF induction (Smolen et al., 2016; Smolen et al., 2020; Zhang et al., 2021). Previous studies have identified activation of PKA (Schacher et al., 1988; Goldsmith and Abrams, 1991; Muller and Carew, 1998) and of the mitogen-

activated protein kinase (MAPK) isoform ERK (Martin et al., 1997; Sharma and Carew,

2004; Sharma et al., 2003) as essential cascades for LTF. These two pathways converge

to phosphorylate transcriptional factors such as CREB, which subsequently induce

expression of multiple plasticity-related genes (Dash et al., 1990; Bartsch et al., 1995; Liu

et al., 2011). The observation that that these two pathways exhibit distinct kinetics of

activation (Muller and Carew, 1998; Philips et al., 2007; Zhang et al., 2021), suggests that

the temporal activity patterns, and activation overlap, of these pathways may constitute

an important target to enhance associative learning. Accordingly, our previous study

demonstrated that a computationally designed protocol with irregular ITIs predicted to

maximize the overlap of PKA and ERK activities enhances LTF and nonassociative

learning, specifically the long-term sensitization of the tail elicited siphon-withdrawal reflex

in *Aplysia* (Zhang et al., 2011).

Substantial similarities between molecular processes of LTF in invertebrates and LTP in

mammals make it plausible that the same strategies used to enhance LTF and

nonassociative learning in invertebrates could enhance LTP and associative learning in

mammals. PKA activation in rodent hippocampus and amygdala is required for LTP and

LTM (Schafe and LeDoux, 2000; Abel et al., 1997). Similarly, activation of ERK/MAPK

cascades and their cross-talk with PKA kinase are required for the phosphorylation of

CREB and LTP induction (Adams and Sweatt, 2002; Impey et al., 1998). Albeit with

different kinetics than observed in invertebrates, PKA and ERK activation in mammals

also differ considerably in temporal dynamics (Roberson and Sweatt, 1996; Vázquez et

al., 2000; Wang et al., 2014), providing a similar opportunity to maximize their overlap

with irregular ITIs in an attempt to enhance memory formation.

Here, we tested if the invertebrate LTF model (Zhang et al., 2011) can be adapted to

computationally design an optimal associative learning protocol in mammals. Empirical

data from the literature were used to model PKA and ERK dynamics in rat hippocampus

(Wang et al., 2014; Roberson and Sweatt, 1996; Vázquez et al., 2000), a critical brain

region implicated in the formation of associative memories in mammals (Milner et al.,

1998; Mayes et al., 2007; Nakazawa et al., 2002). Then we simulated ~1000 different

training protocols with irregular ITIs and identified an optimal protocol, predicted to induce

stronger memory than standard training protocols with fixed ITIs. Using auditory fear

conditioning and fear extinction paradigms in rats, we demonstrated that our

computationally designed protocol enhanced the acquisition as well as the extinction of

fear memories. Using immunohistochemical staining, we confirmed that the

computationally designed protocol induced greater levels of pCREB in the DG of the

dorsal hippocampus. This strategy may have potential clinical relevance for interventions

aiming at facilitating memory formation in psychiatric disorders associated with cognitive

impairment in humans, as well as to enhance extinction-based therapies in patients

suffering from fear-related disorders.

MATERIALS AND METHODS

Animals

All experimental procedures were approved by the Center for Laboratory Animal Medicine and Care of The University of Texas Health Science Center at Houston. National Institutes of Health guidelines for the care and use of laboratory animals were strictly followed in order to minimize any potential discomfort and suffering. A total of 120 male Long-Evans hooded adult rats (Charles Rivers Laboratories) 3-5 months of age and weighing 330-450 g were used. Rats were kept in a 12-hour light/12-hour dark cycle with food and water *ad libitum*. Experiments were conducted during the light phase. For the Optimal Extinction experiment, animals were maintained on a restricted diet of 18 g per day of standard laboratory rat chow to increase their motivation during the lever press training. Rats' weights were monitored weekly to make sure that all animals maintained their weight under food restriction.

Model Development

The mathematical model for the activation of kinase cascades critical for long-term memory (LTM) was modified from a previous model of the signaling cascades for the induction of long-term synaptic facilitation (Zhang et al., 2011). As mentioned above, the induction of LTM for fear conditioning is known for requiring the activation of multiple kinase cascades with different temporal dynamics. In the model, the cyclic AMP (cAMP)-PKA pathway is rapidly activated after training, whereas the Raf- MEK-ERK pathway is slowly activated, and both are required for the induction of LTM after fear conditioning and extinction (**Fig. 1A, Fig. 4A**).

PKA Pathway. The dynamics of cAMP activation upstream of PKA following

training and the cAMP-dependent activation of PKA are described by Eqs. 1-4. Inactive

PKA is a holoenzyme (PKA$_{RC}$, Eq. 2), consisting of regulatory (PKA$_R$, Eq. 3) and catalytic

(PKA$_C$, Eq. 4) subunits. In response to training, represented by the variable *Stim* in Eq.

1, cAMP is activated (Eq. 1). Active cAMP binds to the regulatory subunit of PKA, leading

to the release of free, active catalytic subunit (Fig. 1A) (Eqs. 2-4). '*Stim*' represents the

neurotransmitters released by training that function to activate kinase cascades.

162
$$\frac{d[cAMP]}{dt} = \lambda \frac{[Stim]}{[Stim] + K_{Stim}} - k_{b,cAMP}[cAMP] \qquad \text{(Eq. 1)}$$

163
$$\frac{d[PKA_{RC}]}{dt} = k_{b,PKA}[PKA_C][PKA_R] - k_{f,PKA}[PKA_{RC}][cAMP]^2 \qquad \text{(Eq. 2)}$$

164
$$\frac{d[PKA_R]}{dt} = k_{f,PKA}[PKA_{RC}][cAMP]^2 - k_{b,PKA}[PKA_C][PKA_R] \qquad \text{(Eq. 3)}$$

165
$$\frac{d[PKA_C]}{dt} = k_{f,PKA}[PKA_{RC}][cAMP]^2 - k_{b,PKA}[PKA_C][PKA_R] \qquad \text{(Eq. 4)}$$

Parameter values are modified from (Zhang et al., 2011) to activate PKA immediately

after training, but with PKA activity quickly returning to the basal in ~5 min (Fig. 1B)

(Roberson and Sweatt, 1996; Vázquez et al., 2000): $\lambda = 14.6$ μM/min, $K_{STIM} = 85$ μM,

$k_{b,cAMP} = 4$ min^{-1}, $k_{f,PKA} = 20$ μM^{-2}min^{-1}, $k_{b,PKA} = 12$ μM^{-1}min^{-1}

ERK Pathway. The activation of ERK by training '*Stim*' is via sequential activation

of the upstream kinases Raf and MEK (Fig. 1A). Raf activates the MAP kinase kinase

MEK, MEK in turn activates the MAP kinase ERK. The differential equations describing

the activation of Raf, MEK, and ERK (Eqs. 5-12) are similar to those in (Zhang et al.,

2011). However, a discrete time delay in activation of Raf was removed from the

phosphorylation of Raf (Eq. 5). Instead, based on empirical data (Wang et al., 2014; Ajay

and Bhalla, 2004), parameters describing the activation of Raf were adjusted from (Zhang

et al., 2011) so that the ERK activation curve reached the peak around 20 min post-

training (Fig. 1B).

$$\frac{d[Raf^p]}{dt} = k_{f,Raf}[Raf][Stim] - k_{b,Raf}[Raf^p]$$ (Eq. 5)

$$[Raf] = [Raf]_{total} - [Raf^p]$$ (Eq. 6)

$$\frac{d[MEK]}{dt} = \frac{k_{b,MEK}[MEK^p]}{[MEK^p] + K_{MEK,2}} - \frac{k_{f,MEK}[Raf^p][MEK]}{[MEK] + K_{MEK,1}}$$ (Eq. 7)

$$\frac{d[MEK^{pp}]}{dt} = \frac{k_{f,MEK}[Raf^p][MEK^p]}{[MEK^p] + K_{MEK,1}} - \frac{k_{b,MEK}[MEK^{pp}]}{[MEK^{pp}] + K_{MEK,2}}$$ (Eq. 8)

$$[MEK^p] = [MEK]_{total} - [MEK] - [MEK^{pp}]$$ (Eq. 9)

$$\frac{d[ERK]}{dt} = \frac{k_{b,ERK}[ERK^p]}{[ERK^p] + K_{ERK,2}} - \frac{k_{f,ERK}[MEK^{pp}][ERK]}{[ERK] + K_{ERK,1}}$$ (Eq. 10)

188
$$\frac{d[ERK^{pp}]}{dt} = \frac{k_{f,ERK}[MEK^{pp}][ERK^{p}]}{[ERK^{p}] + K_{ERK,1}} - \frac{k_{b,ERK}[ERK^{pp}]}{[ERK^{pp}] + K_{ERK,2}}$$ (Eq. 11)

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$$[ERK^{p}] = [ERK]_{total} - [ERK] - [ERK^{pp}]$$ (Eq. 12)

Parameter values: $k_{f,Raf}$ = 0.001 $\mu M^{-1} min^{-1}$, $k_{b,Raf}$ = 0.05 min^{-1},   *[Raf]total* = 0.5 μM,

$k_{f,MEK}$ = 0.41 min^{-1}, $k_{b,MEK}$ = 0.04 $\mu M/min$, $K_{MEK,1}$ = 0.20 μM,

$K_{MEK,2}$ = 0.19 μM,   *[MEK]total* = 0.5 μM, $k_{f,ERK}$ = 0.41 min^{-1},

$k_{b,ERK}$ = 0.12 $\mu M/min$, $K_{ERK,1}$ = 0.19 μM,

$K_{ERK,2}$ = 0.21 μM,   *[ERK]total* = 0.5 μM.

As in (Zhang et al., 2011), a variable '*inducer*' was used to quantify the overlap of

activation between PKA and ERK, which together regulate the gene expression

necessary for the induction of LTM.

198   *inducer* = $k_{inducer}$ *[PKAC] [ERKpp]*   (Eq. 13)

where $k_{inducer}$ = 1 μM^{-1}.

To determine which protocols could more effectively activate *inducer,* four-trial protocols

with three ITIs, each ranging from 2–20 min in steps of 2 min, were simulated. Combining

these permutations yielded 10^3 protocols. For each protocol, the maximal (peak) overlap

between PKA and ERK was quantified as the peak level of *inducer* produced. Also, the

value of *Stim* was varied from 100 to 300 μM to represent weak and strong trainings and

to test the robustness of protocols. Based on the maximal overlap, a protocol with ITIs of

8, 8, and 16 min was selected and was denoted optimal partial conditioning (OPC). We

assumed that fear conditioning and extinction activate the same signaling pathways, so

we predicted this protocol would similarly enhance extinction.

Numerical methods. Fourth-order Runge-Kutta integration was used for

integration of all differential equations with a time step of 3 s. Further time step reduction

did not lead to significant improvement in accuracy. The steady-state levels of variables

were determined after at least one simulated day, prior to any manipulations. The model

was programmed in XPPAUT (http://www.math.pitt.edu/~bard/xpp/xpp.html) (Ermentrout,

2002) and simulated on Dell Precision T1700 microcomputers.

Behavioral Tasks

Optimal Conditioning

Apparatuses

Two distinct chambers (context A and context B) positioned inside sound attenuating

boxes were used during the Optimal Conditioning experiments. Context A consisted of a

small operant chamber (34 cm high x 25 cm wide x 23 cm deep, 200 lux, Med Associates,

see schematic drawing in Fig. 2A top) with one of the two aluminum walls covered by

black adhesive paper and two transparent acrylic walls, and a metal grid floor beneath

which a microcentrifuge tube (Eppendorf) containing 50 μl of 10% amyl acetate (Sigma-

Aldrich) was positioned. Context B consisted of a larger acrylic operant chamber (40 cm

high x 50 cm wide x 26 cm deep, 20 lux, Med Associates, see schematic drawing in Fig.

2A bottom) with one of its walls covered by a black and white stripped paper, and a floor

made of a white acrylic board beneath which a microcentrifuge tube (Eppendorf)

containing 50 μl of deionized water was positioned.

Procedures

On day 0, rats were placed in context A for a 20-min familiarization session. Next, rats

were assigned to three experimental groups by matching their baseline freezing and

locomotor activity during the familiarization session: Full Conditioning (FC), Partial

Conditioning (PC), and Optimal Partial Conditioning (OPC). On day 1, rats were placed

into context A and exposed to one nonreinforced habituation tone (3 kHz, 75 dB, 30 s)

followed by distinct fear conditioning protocols (44 min duration). The FC group received

eight presentations of a conditioned stimulus (CS, 3 kHz tone, 75 dB, 30 s) that co-

terminated with an unconditioned stimulus (US, footshock, 0.7 mA, 0.5 s), with fixed ITIs

of 270 s. The PC group received four similar CS-US pairings with the same ITIs of 270 s,

and remained in the chamber until the end of the session. The OPC group received four

similar CS-US pairings with ITIs of 8, 8, and 16 min. On day 2, rats were placed in context

B and given two CS presentations (ITI of 150 s, 7 min duration) in the absence of US to

test the retrieval of tone-associated fear memory in a novel context. On day 3, rats were

returned to context A for an extinction training session where they received twelve CS

presentations (ITIs of 150 s, 37 min duration). The first four CSs were compared to the

last four CSs to assess extinction learning within the same session. On day 4 and 5, rats

were placed back in context A and B, respectively, and exposed to an extinction retrieval

session (similar to day 2). On day 29 and 30, rats were placed back in context A and B,

respectively, and exposed to a spontaneous recovery session (similar to day 2). On day

1 and day 2, rats in the same group were trained simultaneously in four chambers with

the group order counterbalanced to avoid interference from different protocols in

neighboring chambers. On the following days, rats were simultaneously tested in four

chambers regardless of the group assignment. Each rat was tested in the exact same

chamber across the days. Footshocks, tones, intertrial intervals, and session duration

were controlled by an automated video tracking system (ANY-maze, Stoelting), which

also quantified the percentage of time freezing, distance traveled, average speed and

maximum speed. All rats passed the criteria of 20% of freezing during at least one CS of

the fear conditioning session and the first two CSs of the extinction session.

Optimal vs. Spaced Partial Conditioning

Apparatus and Procedures

The same apparatus and contexts described above were used. Following familiarization

on Day 0, rats were assigned to two experimental groups by matching their baseline

freezing and locomotor activity during the familiarization session: Optimal Partial

Conditioning (OPC) and Spaced Partial Conditioning (SPC). The same procedures

described above were used here, except that: *i)* the footshock intensity was reduced from

0.7 mA to 0.5 mA to decrease the possibility of ceiling effects; *ii)* the SPC group received

four CS-US pairings with ITIs of 11 min and 10 s. The same automated video tracking

system (ANY-maze, Stoelting) described above was used for protocol control and

behavioral quantification. One rat that failed the criteria of 20% of freezing during at least

one CS of the fear conditioning session and the first two CSs of the extinction session

was excluded from the analyses.

Optimal Extinction

Lever-press training

Rats were placed in an acrylic/aluminum operant chamber (34 cm high x 25 cm wide x

23 cm deep, Med Associates, see schematic drawing in Fig. 4A) and trained to press a

lever for sucrose on a fixed ratio of one pellet for each press. Next, animals were trained

in a variable interval schedule of reinforcement that was gradually reduced across the

days (one pellet every 15 s, 30 s, or 60 s) until they reached a minimum criterion of 10

presses/min after 7 days of training. All sessions lasted 30 min and were performed on

consecutive days. Sucrose pellet delivery, variable intervals, and session duration were

controlled by an automated video tracking system (ANY-maze, Stoelting).

Apparatus and Procedure

On day 8, rats were placed into the same chamber where they had previously undergone

lever presses training. Animals were exposed to five nonreinforced habituation tones (3

kHZ, 75 dB, 30 s duration) followed by seven CS-US pairings (ITIs of 150 s, 37 min

duration). The footshock intensity was increased from 0.7 mA to 1.0-1.2 mA to result in

stronger fear acquisition and consequently higher freezing levels during the extinction

training session in the next day. Rats were assigned to three experimental groups by

matching their freezing and lever presses during the fear conditioning session: Full

Extinction (FE), Partial Extinction (PE), and Optimized Partial Extinction (OPE). On day

9, rats were returned to the same chamber for a fear extinction session (39 min duration).

The FE group received twelve CSs with ITIs of 150 s; the PE group received four CSs

with ITIs of 150 s and remained in the chamber until the end of the session; and the OPE

group received four CSs with ITIs of 8, 8, and 16 min. On day 10 and 35, rats were placed

into the same chamber and received two CSs with an ITI of 150 s to test the strength of

fear extinction memory during extinction retrieval and spontaneous recovery tests,

respectively. The same automated video tracking system (ANY-maze, Stoelting)

described above was used for protocol control and behavioral quantification. Two rats

that never reached the criterium of 20% of freezing during at least one CS of the fear

conditioning session and the first two CSs of the extinction session were excluded from

the analyses.

Immunohistochemistry

Rats were perfused with 200 ml of 0.1 M potassium phosphate buffer (KPBS) followed by

500 ml of 4% paraformaldehyde (PFA) in 0.1 M sodium phosphate buffer 15 min after day

1 of conditioning. Brains were removed from the skull, transferred to a 20% sucrose

solution in PFA for 24 h, and stored in a 30% sucrose solution in PFA for another 24 h.

Next, coronal brain sections (40 µm thick) were cut with a cryostat (CM1860, Leica) and

stored in an antifreeze solution overnight. Sections were blocked in 20% normal goat

serum, 0.3% Triton X-100, and 3% hydrogen peroxide in KPBS at room temperature for

1 h. Brain sections were then incubated with anti-pCREB serum raised in rabbit (1:1000;

EMD Millipore 06-519) overnight at room temperature. After sections were washed with

KPBS five times, they were incubated with biotinylated goat anti-rabbit IgG antibody

(1:200, Vector Labs, BA-1000-1.5) for 2 h, followed by incubation in ABC kit (1:50, Vector

Labs, PK-6100) for 90 min and DAB-Ni solution (Vector Labs, SK-4100) for 25 min.

Sections were washed with KPBS and mounted on Superfrost Plus slides. After three

days, sections were serially dehydrated with alcohol (Epredia; 50 and 70% for 5min each,

95 and 100% for 10 min each) and defatted with Citrisolv (Decon Labs) for 30 min.

Sections were then cover-slipped with mounting medium (Permount, Fisher Scientific)

before being analyzed microscopically.

Microscopy and image analysis

Images were generated with a Nikon microscope (Eclipse NiE Motorized Upright

Microscope) equipped with a fluorescent lamp (X-Cite, 120 LED) and a digital camera

(Andor Zyla 4.2 PLUS sCMOS). Counts of pCREB-positive cells were performed at 20 x

magnification. Cell detection and quantification for the immunohistochemistry experiment

were performed automatically using QuPath (Bankhead et al., 2017). pCREB-positive cell

detection was performed using QuPath's optical density sum with a requested pixel size

of 0.5 µm. Nucleus parameters included a background radius of 6µm, median filter radius

of 0 µm, sigma of 1.5 µm, minimum area of 2 µm^2, and maximum area of 250 µm^2. Cell

expansion was set to 0 µm. Intensity parameters included a threshold of 0.15 and

maximum background intensity of 5, and intensity threshold parameters were threshold

1 of 0.2, threshold 2 of 0.4, and threshold 3 of 0.6. pCREB-positive cells were

automatically counted with a circular region of interest (0.228 mm^2) in each subfield of rat

dorsal hippocampus including CA1, CA2, CA3, and DG according to the Paxinos rat brain

atlas (Paxinos and Watson, 2014), and averaged for both hemispheres at 2-3 different

antero-posterior levels from -2.4 mm to -4.2 mm from bregma. The density of pCREB-

positive cells (cells per mm^2) was calculated by dividing the number of pCREB-positive

cells by the area of the circular region of interest.

Quantification and statistical analysis

Rats were recorded with digital video cameras (Logitech C920) and behavioral indices were measured using automated video-tracking system (ANY-maze). Lever presses per minute were calculated by measuring the number of presses during the 30 s cue (or 30 s pre-cue) multiplied by two. All graphics and numerical values reported in the figures are presented as mean ± standard error of the mean (SEM). Grubbs' tests were used to identify outliers ($p < 0.05$) in each experiment and one rat in Fig. 2 was removed after identified by the test. Shapiro-Wilk test were performed to determine whether the data have a normal distribution. F tests before pair-wise comparison and Brown-Forsythe tests before multiple group comparison were performed to test if groups have the same variance. Statistical significance for parametric distributions was determined with paired Student's t-test, Welch's t-test, one-way analysis of variance (ANOVA) or two-way repeated-measures ANOVA followed by Tukey post-hoc comparisons, whereas Kruskal-Wallis test followed by Dunn's post-hoc test was used for non-parametric distributions (Prism 7), as indicated for each experiment. Sample size was based on estimations from previous literature and experience.

Data availability

All data that support the findings presented in this study are available from the corresponding author on reasonable request. Source codes will be submitted to the ModelDB database (McDougal et al., 2015), and to GitHub (https://github.com/Owenxz/Zhang-XO-Enhanced-Learning-2022.git).

RESULTS

A computational model based on PKA and ERK dynamics in rats identified an optimal partial fear conditioning protocol

The model of (Zhang et al., 2011) was adapted to simulate the dynamics of PKA and ERK during the process of LTP induction in rat hippocampus (**Fig. 1A**). In this model, the stimulus (*Stim*), which represents tetanic stimuli, rapidly activates PKA via the cAMP pathway and slowly activates ERK via the Raf-MEK pathway. The synergistic interaction between PKA and ERK pathways is quantified by a variable *inducer*, which corresponds to the efficacy of the stimulus in inducing LTP. ERK kinetics were described by differential equations (see Methods) with parameter values reproducing empirical findings that ERK activity reaches peak levels 15~20 min after BDNF treatment or tetanic stimuli in rat hippocampus acute slices (Wang et al., 2014; Ajay and Bhalla, 2004). Equations describing PKA kinetics simulated data showing that PKA is transiently activated within 2 min after LTP induction in slices from rat hippocampus or within 5 min *in vivo* after spatial discrimination task training (Roberson and Sweatt, 1996; Vázquez et al., 2000). We simulated fear conditioning in rats, in which the pairing of a conditioned stimulus (CS) with an unconditioned stimulus (US) is represented by *Stim*, and the conditioned responses are proportional to the peak value of *inducer*. We observed that a single CS-US pairing produced little overlap between PKA and ERK pathways (**Fig. 1B1**). We then simulated ~1000 partial fear conditioning protocols, with 4 trials and varying ITIs, to identify one predicted to have optimal ITIs for triggering downstream gene activation. To compare, we also simulated a Full Conditioning (FC) protocol with 8 trials and a fixed ITI of 270 s, and a Partial Conditioning (PC) protocol with 4 trials and the same ITI of 270 s, which resemble

previous protocols used for full and partial fear conditioning in rats (Detert et al., 2008;

Lonsdorf et al., 2017). Our simulation identified an Optimal Partial Conditioning (OPC)

protocol with 4 trials and irregular ITIs of 8, 8, and 16 min, which was able to produce

higher peak *inducer* than the standard PC protocol with 4 trials (**Fig. 1B2-1B4**).

Comparing OPC and PC in Fig. 1B3-B4, both protocols produced pERK curves with

similar peak levels (orange). However, PKA activity induced by the last trial of OPC had

a much larger overlap with the pERK curve than did PKA induced by PC. Therefore, OPC

induced a much higher peak level of *inducer* (purple). We therefore predicted the OPC

protocol would produce stronger long-term memory in rats than the standard PC protocol.

Figure 1. Computational simulations of PKA and ERK pathways predict an optimal protocol for fear conditioning. A, Schematic of the model. Stimulus (*Stim*, μM) activates PKA via cAMP and activates ERK via Raf-MEK. The variable *inducer* quantifies the PKA/ERK interaction (activity overlap). **B1**, Simulated time courses of activated ERK (pERK, μM) and activated PKA (PKAc, μM) in response to one trial of *Stim*. **B2**, Simulated time courses of pERK, PKAc, and *inducer* (nM) in response to an 8-trial protocol with regular ITIs of 4.5 min (full conditioning, FC). **B3**, Simulated time courses of pERK, PKAc, and *inducer* in response to a 4-trial protocol with regular ITIs of 4.5 min (partial conditioning, PC). **B4**, Simulated time courses

of pERK, PKAc, and *inducer* in response to a 4-trial protocol with computationally designed

intervals (optimal partial conditioning, OPC).

The optimal partial conditioning protocol induces stronger fear memory than a

standard partial conditioning protocol

To test if the computationally designed OPC protocol could enhance associative learning

in rats, we used a classical auditory fear conditioning paradigm. Because the

hippocampus is required for context encoding during fear conditioning (Phillips and

LeDoux, 1992; Selden et al., 1991; Maren et al., 2013), we used two different chambers

(context A and context B) to assess the contribution of the context to the auditory fear

memory. The two contexts differed in size, floor texture, visual decoration of the walls,

and odor cues (**Fig. 2A**). On Day 0, rats were pre-exposed to context A for 20 min to

familiarize with the chamber where fear conditioning took place in the following day. After

matching their baseline freezing and locomotor activity during the familiarization session,

rats were assigned to one of the three experimental groups described above in our

simulation experiments: Full Conditioning (FC), Partial Conditioning (PC), and Optimal

Partial Conditioning (OPC). No significant difference was found among their baseline

freezing level (FC: 3.05%, PC: 2.84%, OPC: 2.96%, one-way ANOVA, $F_{(2, 37)}$ = 0.018, *P*

= 0.982), average speed (FC: 0.023 m/s, PC: 0.025 m/s, OPC: 0.025 m/s, one-way

ANOVA, $F_{(2, 37)}$ = 0.062, *P* = 0.939), and maximum speed (FC: 0.439 m/s, PC: 0.424 m/s,

OPC: 0.446 m/s, one-way ANOVA, $F_{(2, 37)}$ = 0.044, *P* = 0.956). On Day 1, rats were placed

into context A and exposed to one nonreinforced habituation tone (3 kHZ, 75 dB, 30 s)

followed by distinct fear conditioning protocols (44 min duration, **Fig. 2B**). The FC group

received eight presentations of CS (3 kHZ tone, 75 dB, 30 s) that co-terminated with an

US (footshock, 0.7 mA, 0.5 s) with ITIs of 270 s. The PC group received four similar CS-

US pairings with the same ITIs of 270 s, and remained in the chamber until the end of the

session to equate the total context exposure time. The OPC group received four similar

CS-US pairings with ITIs of 8, 8, and 16 min. All groups reached high levels of freezing (>

65%, **Fig. 2C**) and reduced locomotion at the end of the fear acquisition session. On Day

2, rats were placed in context B and given two CS presentations (ITI of 150 s, 7 min

duration) in the absence of US to test the retrieval of tone-associated fear memory in a

novel context. Compared to the FC group, both PC and OPC groups exhibited less

freezing during the second CS presentation (Two-way repeated measures (RM) ANOVA

followed by Tukey's post-hoc test. $F_{(2, 37)}$ = 3.751, P = 0.033; FC vs. OPC: p = 0.029; FC

vs. PC: p = 0.002; OPC vs. PC: p = 0.527). Nevertheless, both FC and OPC groups

showed reduced average speed during the retrieval test compared to the standard PC

group, suggesting stronger fear retrieval in a novel context (**Fig. 2D;** One-way ANOVA

followed by Tukey's post hoc test. $F_{(2, 37)}$ = 10.87, P < 0.001; FC vs. OPC: p = 0.131; FC

vs. PC: p < 0.001; OPC vs. PC: p = 0.024). On Day 3, rats were returned to context A for

an extinction training session where they received twelve CS presentations (ITIs of 150

s, 37 min duration). Although no significant differences in freezing levels were observed

among the groups during each CS presentation (Two-way RM ANOVA. $F_{(22, 407)}$ = 1.169,

P = 0.271), a within session extinction analysis comparing the first four CSs to the last

four CSs revealed impaired extinction learning in the FC and OPC groups, when

compared to the PC group. This impairment in the acquisition of extinction in the FC and

OPC groups was characterized by sustained freezing levels across the session, which

differed from the PC group that showed reduced freezing levels from the first to the last

CSs presentation (**Fig. 2E left;** Paired Student's t-test. FC: t_{13} = 2.153, p = 0.051; PC: t_{11}

= 4.191, p = 0.001; OPC: t_{13} = 0.869, p = 0.400**)**. Interestingly, the OPC group exhibited

persistent freezing levels during the extinction training session as indicated by a reduced

relative change in freezing from the beginning to the end of the session, when compared

to the PC group (**Fig. 2E right;** One-way ANOVA followed by Tukey's post-hoc test. $F_{(2,}$

$_{37)}$ = 3.315, P = 0.047; FC vs. OPC: p = 0.487; FC vs. PC: p = 0.320; OPC vs. PC: p =

0.037). Despite the persistent freezing levels across the extinction training session

observed in the OPC group, no significant differences were found between the OPC and

the PC groups during either the extinction retrieval tests performed on Days 4 and 5 (Two-

way RM ANOVA. context A: $F_{(2, 37)}$ = 0.097, P = 0.907; context B: $F_{(2, 37)}$ = 2.984, P =

0.063) or the spontaneous recovery tests performed on Days 29 and 30 (**Fig. 2C;** Two-

way RM ANOVA followed by Tukey's post hoc test. context A: $F_{(2, 37)}$ = 3.259, P = 0.049;

FC vs. OPC: p = 0.027; FC vs. PC: p = 0.038; OPC vs. PC: p = 0.999; context B: $F_{(2, 37)}$

= 0.774, P = 0.469). Taken together, these data demonstrate that the computationally

designed OPC protocol is able to enhance fear acquisition in rats, thereby resulting in a

stronger fear memory that is more resistant to extinction.

Figure 2. Computationally designed protocol enhanced fear conditioning in rats. **A,**

Schematics of the enhanced fear conditioning experimental procedures. Top: tests that were

conducted in Context A. Bottom: tests that were conducted in Context B. **B,** Schematics of the

fear conditioning protocols for FC (n = 14), PC (n = 12), and OPC (n = 14) groups. Following a

habituation (Hab.) tone, rats received multiple trials of a CS (3 kHz tone, 30 s) that co-terminated

with a US (footshock, 0.7 mA, 0.5 s). **C,** Freezing levels during CS presentations of each group

across the experiment. Two-way Repeated-Measure ANOVA for each day followed by Tukey's

post hoc test. Letters a, b, and c indicate pairwise post hoc tests with $p < 0.05$: a, FC vs. OPC; b,

FC vs. PC; c, OPC vs. PC. **D,** OPC group shows lower average speed compared to PC group.

One-way ANOVA followed by Tukey post hoc test. Letters b and c indicate pairwise tests with p

< 0.05: b, FC vs. PC ; c, OPC vs. PC. **E,** OPC group is resistant to extinction while PC group

shows significantly more extinction. Left, the freezing levels during the first 4 and last 4 CS

presentations. Paired Student's t-test. Letter e indicates test with $p < 0.05$: PC, Last 4 CS vs. First

4 CS. Right, normalized change of the freezing level during extinction, as indicated by the

difference of the freezing levels between last 4 and first 4 CS presentations as a percentage of

the freezing level of the first 4 CS presentations ((Last 4 CS – First 4 CS)/First 4 CS). One-way

ANOVA followed by Tukey's post hoc test. * $p<0.05$. Letter c indicates pairwise test with $p < 0.05$:

OPC vs. PC. Data shown here and in the next illustrations as mean ± SEM.

The optimal partial conditioning protocol induces stronger fear memory than a

spaced partial conditioning protocol

Previous studies have demonstrated that spaced learning protocols result in stronger

memories than standard (i.e., massed) learning protocols in both humans (Cepeda et al.,

2006; Raman et al., 2010) and rodents (Anderson et al., 2008; Jiang et al., 2019; Scharf

et al., 2002; but also see Cain et al., 2003). Thus, an alternative explanation for the

augmented fear memory observed in the OPC group in Fig. 2 could be simply a trial-

spacing effect, because our standard PC group used massed CS-US pairings. In

simulations, both protocols produced similar peak values of the pERK curves, but different

overlap of PKA and ERK activities, suggesting a way in which appropriately spaced

protocols might advantageously enhance overlap of kinase activities with distinct

dynamics. Indeed, there was a broad range of effective 4-trial protocols (**Fig. 3**), with one

protocol with fixed 11 min 10 s ITIs producing a nearly identical peak level of inducer as

the OPC protocol, and a higher peak level of inducer than other protocols with fixed ITIs

from 1 min to 21 min (**Fig. 3A, Fig. 4A1**). We termed this protocol the spaced partial

conditioning (SPC) protocol. We also examined the effects of different stimulus intensities

on predicting optimal training trials. The 1,000 different protocols were re-simulated with

a reduced intensity of stimulus. Interestingly, the protocol with 4 trials and irregular ITIs

of 8, 8, and 16 min still produced the greatest peak level of inducer (**Fig. 3B, Fig. 4A2**).

Equally interesting was that with weaker stimuli, the OPC protocol became relatively more

effective than the SPC protocol. Therefore, we predicted that empirically, the irregular

intervals of the OPC may produce more effective conditioning than the SPC protocol.

Figure 3. **Histogram of peak levels of *inducer* from 1,000 protocols. A**, Standard intensity

stimulus (*Stim* = 300 μM). The range of peak levels of *inducer* (0 – 220 nM) was subdivided into

14 bins, and the number of simulations that produced a peak concentration of *inducer* in each

subdivision was plotted. The arrows indicate which bins contained the peak concentrations

produced by the partial conditioning, spaced partial conditioning (SPC), and optimal partial

conditioning (OPC) protocols. **B,** Reduced intensity stimulus (*Stim* = 200 μM). The range of peak

levels of *inducer* (0 – 160 nM) was subdivided into 14 bins, and the number of simulations that

produced a peak concentration of *inducer* in each subdivision was plotted. Insets below the main

plots illustrate in detail the difference in peak *inducer* for SPC *vs*. OPC, which is negligible in **A**.

To test this prediction, we empirically compared the OPC and SPC protocols using a

reduced shock intensity (0.5 mA instead of the 0.7 mA used in Fig. 2) (**Fig. 4B**). Rats

were assigned to OPC or SPC groups by matching their baseline freezing and locomotor

activity during the familiarization session. No significant difference was found between

their baseline freezing level (SPC: 3.72%, OPC: 5.21%, Unpaired Student's t-test, t_{16} =

0.988, *P* = 0.337), average speed (SPC: 0.016 m/s, OPC: 0.015 m/s, one-way ANOVA,

$F_{(2, 37)}$ = 0.062, *P* = 0.939), and maximum speed (FC: 0.439 m/s, PC: 0.424 m/s, OPC:

0.446 m/s, one-way ANOVA, $F_{(2, 37)}$ = 0.044, *P* = 0.956). During the fear acquisition, the

SPC group received 4 CS-US pairings with ITIs of 11 min and 10 s, whereas the OPC

group was exposed to the same protocol as in Fig. 2 (*i.e.*, 4 CS-US pairings with ITIs of

8, 8, and 16 min) (**Fig. 4C**). We found that the difference in the ITIs between the OPC

and SPC groups was sufficient to result in distinct levels of fear memory acquisition. Rats

exposed to OPC showed higher CS freezing levels during the third CS-US pairing of the

fear acquisition session (**Fig. 4D Day 1;** Two-way RM ANOVA followed by Tukey's

planned comparison. $F_{(4, 64)}$ = 1.550, *P* = 0.339; OPC vs. SPC: p = 0.019). In addition,

fear memory retrieval was increased when rats in the OPC group were returned to context

A for an extinction training session on Day 3, as indicated by higher freezing during the

first two CS presentations, compared to the SPC group (**Fig. 4D Inset;** Welch's t-test.

$t_{12.43}$ = 2.227, p = 0.045). Both OPC and SPC groups exhibited the same levels of freezing

by the end of the extinction training session (**Fig. 4D Day 3;** Two-way RM ANOVA

followed by Tukey's post hoc test. $F_{(11, 176)}$ = 2.997, P = 0.001; Tone 12, OPC vs. SPC:

p > 0.999), as well as during the extinction retrieval tests performed in context A (**Fig. 4D**

Day 4; Two-way RM ANOVA. $F_{(2, 37)}$ = 0.097, P = 0.907) or context B (**Fig. 4D Day 5;**

Two-way RM ANOVA. $F_{(2, 37)}$ = 2.984, P = 0.629). However, fear memory spontaneously

recovered when rats in the OPC group were retested in context A approximately 3 weeks

later, as indicated by higher freezing during the first CS presentation compared to SPC-

trained rats (**Fig. 4D Day 29;** Two-way RM ANOVA followed by Tukey's planned

comparison. $F_{(2, 32)}$ = 1.809, P = 0.180; OPC vs. SPC: p = 0.030). In summary, these data

suggest that the enhancement in fear memory acquisition observed with our

computationally designed OPC protocol cannot simply be attributed to a trial-spacing

effect or differences in the delay to remove the animals from the chamber, but it is rather

associated with the maximized overlap between PKA and ERK signaling.

Figure 4. **Optimal partial conditioning protocol induced stronger fear memory than spaced**

partial conditioning protocol in rats. **A**, *Inducer* peak levels from the FC, PC, and OPC

protocols, compared to peak levels from partial conditioning protocols of 4 trials with regular ITIs

varying from 1 to 21 min, using standard intensity stimulus (**A1**) or reduced intensity stimulus (**A2**).

The peak *inducer* values of FC, PC, and OPC are labeled by dashed lines (black: FC; red: OPC;

purple: PC). The blue curve gives peak *inducer* values for the protocols with regular ITIs, and the

curve peaks at the dark blue dot and arrow, representing the SPC protocol with equal ITIs of 11

min and 10 s. **B,** Schematics of the fear conditioning procedures. Top: tests that were conducted

in Context A. Bottom: tests that were conducted in Context B. **C,** Schematics of the fear

conditioning protocols for OPC (n = 8), and SPC, (n = 10, ITI = 670 s) groups. Following a

habituation (Hab.) tone, rats received 4 trials of a CS (3 kHz tone, 30 s) that co-terminated with a

US (footshock, 0.5 mA, 0.5 s). **D,** Freezing levels during CS presentations of each group across

the experiment; Two-way repeated-measure ANOVA for each day followed by Tukey's pairwise

post hoc test, * $p < 0.05$. Inset: the average freezing level during the first two CS presentations;

Welch's t-test, * $p < 0.05$.

The optimal partial conditioning protocol induces greater levels of phosphorylated

CREB in DG than a spaced partial conditioning protocol

The computational model predicts that the increased overlap of PKA and ERK activation

in OPC protocol leads to greater LTP induction in rat hippocampus. One of the

mechanisms that may enhance LTP is increased convergence of the PKA and ERK

pathways to induce phosphorylation of CREB (Impey et al., 1998), a transcription factor

required for synaptic plasticity and LTP (Liu et al., 2011; Bourtchuladze et al., 1994; See

review by Kida, 2012). We therefore used immunohistochemistry to quantify the levels of

pCREB in rat dorsal hippocampus following fear conditioning in OPC versus SPC groups.

We repeated the same experimental design used in **Fig 4,** as the observed augmentation

in fear acquisition was predicted to correlate with differences in pCREB levels between

the two groups following fear conditioning (**Fig. 5A**). Rats were assigned to OPC or SPC

groups by matching their baseline freezing and locomotor activity during the

familiarization session. No significant difference was found between their baseline

freezing level (SPC: 5.90%, OPC: 8.53%, Unpaired Student's t-test, t_5 = 0.677, p = 0.528),

average speed (SPC: 0.014 m/s, OPC: 0.015 m/s, Unpaired Student's t-test, t_5 = 0.239 p

= 0.820), and maximum speed (SPC: 0.272 m/s, OPC: 0.367 m/s, Unpaired Student's t-

test, t_5 = 1.697, p = 0.150). During the fear acquisition session, rats exposed to the OPC

or SPC protocols showed similar levels of freezing during the CS presentations (**Fig. 5B**,

Two-way RM ANOVA. $F_{(4, 20)}$ = 0.848 , P = 0.5112. Inset, Unpaired Student's t-test, t_5 =

2.405, p = 0.061), most likely due to the reduced sample sizes. However, OPC rats

reached significantly higher levels of CS freezing during the end of the conditioning

session (Unpaired Student's t-test, t_5 = 2.769, p = 0.039). Approximately 15 min after the

end of fear acquisition session, rats were perfused and their brains were collected for

subsequent analyses. Immunohistochemistry revealed high levels of pCREB expression

across different hippocampal subfields in both OPC and SPC groups, with the highest

expression level being observed in the DG (**Fig. 5C**; Two-way ANOVA followed by

Tukey's post-hoc test. $F_{(3, 20)}$ = 4.533, P = 0.014; CA1 vs. DG: p < 0.001; CA2 vs. DG: p

< 0.001; CA3 vs. DG: p < 0.001). Interestingly, rats exposed to the OPC protocol exhibited

increased pCREB expression in the DG when compared to the SPC group (**Fig. 5C-D;**

Two-way ANOVA followed by Bonferroni's post-hoc test. $F_{(3, 20)}$ = 4.533, P = 0.014; OPC

vs. SPC: p = 0.004). Given that increased pCREB levels in the DG has been associated

with the induction efficiency of LTP in rodents (Davis et al., 2000), our immunostaining

results support the prediction that the augmentation in fear memory acquisition observed

with the OPC protocol is mediated by higher levels of LTP induction in the rat

hippocampus.

Figure 5. **Immunohistochemistry quantification of pCREB in hippocampus of rats exposed**

to optimal versus spaced partial conditioning protocols. **A,** Schematic of the

immunohistochemistry procedure. **B,** Freezing levels during CS presentation for OPC (n = 3) and

SPC (n = 4) groups during fear conditioning; Two-way RM ANOVA followed by Bonferroni

correction. Inset, averaged freezing level for the 4 CS presentations; Student's t-test. **C,** Average

pCREB density across hippocampal subfields for OPC (n = 3) and SPC (n = 4) groups; Two-way

ANOVA followed by Tukey's Bonferroni correction. *p < 0.05. **D,** Representative images of pCREB

immunostaining in distinct subfields of the dorsal hippocampus (left) in the SPC group (middle)

and OPC group (right).

The optimal partial protocol also enhanced fear extinction

Extinction is a new learning that temporarily inhibits the initial associative memory (de

Oliveira Alvares and Do-Monte, 2021; Quirk and Mueller, 2008). Because previous

studies suggest that the molecular mechanisms underlying fear extinction learning are

similar to those underlying fear acquisition (Berman and Dudai, 2001; Flood et al., 1977;

Radulovic and Tronson, 2010; Santini et al., 2001; Szapiro et al., 2003; Tronson et al.,

2008), we hypothesized that the computationally designed protocol would also enhance

the acquisition of extinction, suppressing the original fear memory (**Fig. 6A-B1**). We

therefore used the OPC protocol with 4 trials and irregular ITIs of 8, 8, and 16 min as an

Optimal Partial Extinction (OPE) protocol (**Fig. 6B4**). We compared the OPE protocol with

a Full Extinction (FE, **Fig. 6B2**) protocol using 12 trials and ITIs of 150 s, and a Partial

Extinction (PE, **Fig. 6B3**) protocol using 4 trials and the same ITIs of 150 s, which

resemble previous protocols used for full and partial fear extinction in rats (Santini et al.,

2004; Holmes and Quirk, 2010; Do-Monte et al., 2015). Similar to the simulation results

for the conditioning protocols, the OPE protocol triggered higher peak *inducer* than the

PE protocol with equal ITIs of 150 s, and was comparable to the FE protocol (**Fig. 6B**).

We thus predicted the OPE protocol would result in stronger extinction of fear memory

than the standard PE protocol.

Figure 6. Prediction of enhanced protocol for fear extinction. A, Schematic of the model. B1, Simulated time courses of activated ERK (pERK, μM) and activated PKA (PKAc, μM) in response to one trial of *Stim* (μM). B2, Simulated time courses of pERK, PKAc, and *inducer* (nM) in response to 12-trial protocol with regular intervals of 2.5 min (full extinction, FE). B3, Simulated time courses of pERK, PKAc, and *inducer* in response to 4-trial protocol with regular intervals of 2.5 min (partial extinction, PE). B4, Simulated time courses of pERK, PKAc, and *inducer* in response to 4-trial protocol with computationally designed intervals (optimal partial extinction, OPE).

To test this prediction, we designed an experiment comparing the efficacy of different

extinction protocols (**Fig. 7A**). We used conditioned suppression of reward-seeking

behavior as an additional measure of fear memory because lever-press suppression is

more sensitive than freezing during fear extinction paradigms (Mast et al., 1982; Sotres-

Bayon et al., 2012). In addition, lever presses help to maintain a constant level of activity

so that freezing can be more reliably measured along the session (Quirk et al., 2000).

Since lever presses for reward are trained in a specific context, and extinction memory is

context-dependent (Podlesnik et al., 2017; Goode and Maren, 2014), we used only one

context in this experiment so that lever press suppression can be consistently assessed

across sessions and the extinction memory can be retrieved after the extinction learning.

Rats were initially trained to press a lever to receive sucrose pellets in a variable interval

schedule of 60 s. After 7 days of training, rats reached the same levels of lever pressing

(~10 presses/min) and locomotor activity. On Day 8, they underwent a fear conditioning

session which included five nonreinforced habituation tones (3 kHZ, 75 dB, 30 s duration)

followed by seven CS-US pairings (ITIs of 150 s, 37 min duration). In simulations, the

OPE group showed peak levels of *inducer* comparable to the FE group and higher than

the PE group when strong stimuli were used (**Fig. 6**). We therefore increased the

footshock intensity from 0.7 mA to 1.0 mA, which resulted in stronger fear acquisition and

consequently higher freezing levels during the extinction session. This stimulus intensity

also helped to avoid a "floor effect" in the following tests. On Day 9, rats were assigned

to three experimental groups by matching their freezing and lever presses during the fear

conditioning session: Full Extinction (FE), Partial Extinction (PE), and Optimal Partial

Extinction (OPE) (One-way ANOVA. $F_{(14, 231)} = 0.404$, $P = 0.973$). No significant difference

was found among their freezing level (FE: 3.20%, PE: 4.43%, OPE: 3.07%, Kruskal-

Wallis's test $H(2)$ = 0.188, P = 0.910), average speed (FE: 0.036 m/s, PE: 0.036 m/s,

OPE: 0.037 m/s, one-way ANOVA, $F_{(2, 33)}$ = 0.042, P = 0.958), and lever presses (FE:

11.13 presses/min, PE: 9.64 presses/min, OPE: 9.603 presses/min, one-way ANOVA, $F_{(2,}$

$_{33)}$ = 0.480, P = 0.622). On Day 10, rats were returned to the same chamber for a fear

extinction session (39 min duration). The FE group received twelve CSs with ITIs of 150

s; the PE group received four CSs with ITIs of 150 s and remained in the chamber until

the end of the session; and the OPE group received four CSs with ITIs of 8, 8, and 16

min (**Fig. 7B**). Our results revealed that at the end of the CS presentations, freezing levels

reduced from ~50% to ~20% in the FE groups, whereas the PE and OPE groups

maintained the same levels of freezing throughout the four CSs (**Fig. 7C**, Day 8). Similarly,

the FE group showed a significant increase in lever presses (from 0.1 to 1.5 press/min)

during the CS presentations from the beginning to the end of the session, whereas CS

lever presses remained unaltered in the PE and OPE groups (paired Student's t-test; FE,

first 4 CS lever presses: 0.08 presses/min, last 4 CS lever presses: 1.5 presses/min, t_{11}

= 2.837, p = 0.016; PE, first 4 CS lever presses: 0 presses/min, last 4 CS lever presses:

0.18 presses/min, t_{10} = 1.000, p = 0.341; OPE, first 4 CS lever presses: 0.46 presses/min,

last 4 CS lever presses: 1.23 presses/min, t_{12} = 1.237, p = 0.239). However, the OPE

group showed a significant increase in lever presses during the 30-s periods that

preceded the CS presentations (pre-CS) when comparing the first two pre-CS periods to

the last pre-CS period of the extinction training session (**Fig. 7D&E**; Paired Student's t-

test; FE: t_{11} = 0.994, p = 0.341; PE: t_{10} = 1.111, p = 0.293; OPE: t_{12} = 3.773, p = 0.003),

suggesting enhanced within-session extinction of contextual fear memory (Morgan and

LeDoux, 1995). On Days 10 and 35, rats were returned to the same chamber and

received two CSs with an ITI of 150 s to test the strength of fear extinction memory during

extinction retrieval and spontaneous recovery tests, respectively. No significant difference

in CS freezing or CS lever presses were observed among the three groups during these

tests (**Fig. 7C, Day 10&35;** Two-way RM ANOVA. Day 10: Freezing, $F_{(2, 33)} = 0.207$, $P =$

0.814; Lever Presses, $F_{(2, 33)} = 0.069$, $P = 0.933$; Day 35: Freezing, $F_{(2, 33)} = 0.535$, $P =$

0.591; Lever Presses, $F_{(2, 33)} = 1.085$, $P = 0.350$). However, the OPE group showed

increased pre-CS lever pressing rate either in the first pre-CS period (**Fig. 7D;** Two-way

RM ANOVA followed by Tukey's planned comparison. $F_{(2, 33)} = 2.897$, $P = 0.069$; Tone 1,

a, FE vs. PE, p = 0.791; b, FE vs. OPE, p = 0.025; c, OPE vs. PE, p = 0.005) or when

averaging the two pre-CS periods during the spontaneous recovery test (**Fig. 7F;** Kruskal-

Wallis's test followed by Dunn's post hoc test. $H(2) = 6.414$, $P = 0.041$; a: FE vs. OPE, p

= 0.531; b: FE vs. PE, p = 0.709; c: OPE vs. PE, p = 0.035), when compared to the FE

and PE groups on day 35. These results suggest that, although the tone-associated

memory was similar among the groups, our computationally designed OPE protocol was

able to enhance the extinction of contextual fear memory, as indicated by enhanced

within-session extinction of conditioned suppression and by reduced conditioned

suppression approximately 3 weeks after the extinction training session.

Figure 7. Computationally designed protocol enhanced fear extinction in rats. **A.** Schematic

of the enhanced fear extinction experimental procedures. Rats were trained to constantly press a

lever to retrieve sucrose pellets in a variable interval 60 (VI60) schedule, where the average

interval between each sucrose delivery is 60 s. **B,** Schematics of the fear extinction protocols for

FE (n = 12), PE (n = 11), and OPE (n = 13) groups, during which rats received multiple trials of a

CS (3 kHz tone, 30 s). **C,** Freezing levels during CS presentations of each group across the

experiment. No significant difference between groups was found by Two-way RM ANOVA. **D,**

Lever press rates during the 30 s of each CS presentation (pre-CS period) of each group across

the experiment. Two-way repeated-measure ANOVA for each day followed by Tukey's pairwise

post hoc test. Letters a and c indicate pairwise tests with $p < 0.05$: a, FE vs. OPE ; c, OPE vs. PE.

E, OPE group shows significant increase of pre-CS lever presses comparing the last CS

presentation to the average lever press rate of the first two CS presentations. Paired Student's t-

test. Letter f indicates test with $p < 0.05$: OPC, Last 4 CS vs. First 4 CS. **F,** OPE group shows

higher lever presses during pre-CS period compared to PE group in spontaneous recovery test.

One-way ANOVA followed by Tukey's post hoc test. Letter c indicates pairwise comparison with

* $p<0.05$: OPE vs. PE.

DISCUSSION

Studies on nonassociative learning in invertebrates have provided significant insights

about the molecular mechanisms of learning and memory due to their simplicity and

tractability (Byrne and Hawkins, 2015; Kandel, 2001; Carew and Sahley, 1986). Long-

term nonassociative learning in different regions of the nervous system or across different

species shares many intracellular molecular cascades including cAMP-PKA and Raf-

MEK-ERK signaling (Bartsch et al., 1995; Martin et al., 1997). These two pathways are

also required for associative learning in invertebrates and mammals (Hawkins and Byrne,

2015; Schafe and LeDoux, 2000; Adams and Sweatt, 2002). The conservation of these

mechanisms makes these molecular cascades potential candidates for a universal

intervention to enhance learning and memory. We have previously demonstrated that a

computationally designed protocol that maximizes the interaction between PKA and ERK

pathways enhances LTF and nonassociative learning in *Aplysia* (Zhang et al., 2011).

Here, we extended the feasibility of this computational approach to associative learning

in mammals by adapting the simplified mathematical model used in *Aplysia* to simulate

the dynamics of PKA and ERK in rat hippocampus based on empirical data available in

the literature (Wang et al., 2014; Roberson and Sweatt, 1996; Vázquez et al., 2000). After

simulating ~1000 different training protocols of auditory fear conditioning, we identified an

optimal protocol with irregular ITIs which resulted in stronger fear acquisition and

extinction-resistant memory formation in rats, when compared to either massed or spaced

training protocols with equal ITIs. Immunohistochemical staining demonstrated that the

augmented fear memory induced by the OPC protocol was associated with increased

expression levels of pCREB in the DG subfield of the dorsal hippocampus. Using a

separate fear extinction paradigm, we also showed that an optimal training protocol was

sufficient to induce enhanced extinction of contextual fear memory in rats, when

compared to standard or partial extinction protocols with equal ITIs. Our results

demonstrate the power of a simplified model of intracellular signaling cascades in guiding

associative learning across species, attesting the essential role of the interaction between

PKA and ERK pathways in both nonassociative and associative learning. In addition, our

findings highlight the potential clinical relevance of using computational modeling to

enhance associative learning in patients with memory deficits, as well as to enhance

extinction-based exposure therapies in patients with anxiety disorders.

Pioneering studies have demonstrated that training protocols using spaced ITIs result in

stronger memory acquisition than those using standard massed ITIs, a well-established

phenomenon described as the "spacing effect" (see review by Smolen et al., 2016). Data

demonstrating this effect in animals and humans illustrate the relationship between the

duration of the ITIs and the strength of memory formation as an inverted-U-shape curve

(Verkoeijen et al., 2005; Barela, 1999), consistent with our simulated results (**Fig. 1**).

Previous findings have attributed this spacing effect to the different efficacy of massed vs.

spaced protocols in inducing LTP/LTF through plasticity-related signaling. For instance,

spaced training protocols in rodents lead to increased activation of ERK (Ajay and Bhalla,

2007), followed by CREB phosphorylation and the expression of downstream genes,

which have not been reported following massed training protocols in the same species

(Genoux et al., 2002). A computational model based on slice electrophysiology from rat

hippocampal data has found that the peak of ERK activation aligns with the optimal ITI

for LTP induction (Ajay and Bhalla, 2004). Although that study provided a possible

mechanism for the trial-spacing effect and a potential approach to identify more effective

protocols for associative learning, it relied on fixed ITIs that are not necessarily the optimal

intervals to induce synaptic plasticity and LTM. In fact, our simulations showed that an

optimal protocol with irregular ITIs was superior to most spaced protocols with the same

number of trials and equal ITIs. In the fear conditioning experiment, the optimal protocol

induced stronger and persistent fear memory in rats, when compared to a spaced protocol

predicted to be the most efficient among protocols with equal ITIs. These results suggest

that the trial-spacing effect in mammals can be at least partially explained by enhanced

overlap between PKA and ERK pathways, which are critical for CREB activation (Impey

et al., 1998). More importantly, our findings demonstrate that protocols that include

irregular ITIs can be a better approach to enhance learning than previous protocols using

equal ITIs.

The enhanced performance of our optimal protocol (OPC) compared to a spaced protocol

(SPC) during fear conditioning appears constrained by the intensity of stimuli. The model

predicts higher performance for OPC when weak stimuli are used, but comparable

performance when strong stimuli are used (**Fig. 4A**). Drawing a line to distinguish weak

and strong footshock intensities is not straightforward because the relationship between

fear memory and footshock intensity is neither monotonic nor linear (Davis and Astrachan,

1978). Nevertheless, we observed clear differences in the efficacy of the optimal fear

conditioning protocol when using footshocks of different intensities. When we compared

the optimal partial conditioning protocol to a massed partial conditioning protocol using a

standard footshock intensity (0.7 mA, **Fig. 2**) commonly used in previous studies (Abel et

al., 1997), we found only small differences in the conditioned responses (CS freezing)

between the two protocols. However, when we used a lower footshock intensity (0.5 mA,

Fig. 4) to compare the optimal partial conditioning protocol to a spaced partial

conditioning protocol, we found a significant increase in CS freezing in the OPC group

during fear acquisition, retrieval, and spontaneous recovery tests. Considering the

robustness of the spacing effect, it is possible that differences between the OPC and the

massed PC groups at the higher footshock intensity were masked by a "ceiling effect".

These data suggest that our approach could be more beneficial for learning protocols

relying on relatively weak stimuli. Additional studies need to be conducted to understand

the mechanism by which the intensity of stimuli determines the enhanced performance of

protocols with irregular ITIs compared to protocols with fixed ITIs.

The association between the CS and US is primarily mediated by the lateral amygdala

where LTP induces enhanced CS responses (see review by Maren, 2005; Johansen et

al., 2011). Nevertheless, another important and distinguishable component of fear

memory formation is the context in which the association has occurred. In the fear

conditioning experiment, enhanced CS freezing was only observed in the same context

where fear conditioning took place, suggesting that the memory facilitation effect induced

by our optimal protocol is context dependent. Similarly, in the fear extinction experiment,

enhancement by the optimal protocol was only observed during the pre-CS lever pressing,

a more sensitive index of contextual fear memory during extinction and spontaneous

recovery (Morgan and LeDoux, 1995; Padilla-Coreano et al., 2012; Woods and Bouton,

2008; Sotres-Bayon et al., 2012; Mast et al., 1982). These results are consistent with

current empirical data from rat hippocampus, on which our model is based. The

hippocampus plays major roles in context encoding during fear conditioning, fear

extinction, and the time-dependent reappearance of fear following extinction training (*i.e.*,

spontaneous recovery) (see reviews by Maren et al., 2013; Bouton et al., 2021).

Accordingly, our immunohistochemical results showed greater pCREB expression

specifically in the DG of the dorsal hippocampus, a region implicated in the acquisition

and extinction of contextual fear memory (Kheirbek et al., 2013; Bernier et al., 2017).

Increasing CREB expression in DG neurons using a viral approach has been

demonstrated to improve contextual fear acquisition (Restivo et al., 2009), whereas

chemogenetically or optogenetically inactivating DG neurons that overexpress CREB

disrupts contextual fear retrieval (Park et al., 2016).

Another hippocampal subregion implicated in contextual fear is the CA1 (Hunsaker and

Kesner, 2008). Although the OPC and SPC groups displayed similar levels of pCREB

expression in CA1, the single post-training time point we used do not suffice to rule out

the possibility of distinct pCREB dynamics in CA1 compared to DG. Indeed, prior reports

have shown that CA1 pCREB expression peaks at 30 min and disappears at 90 min

following fear conditioning (Sindreu et al., 2007; Chaaya et al., 2019), a timeframe that

differs from the 15 min post-training interval we used to detect group differences in DG

pCREB expression. Further experiments comparing distinct post-training intervals may

better reveal the dynamics of pCREB expression in different hippocampal subregions, as

well as in non-hippocampal areas involved in fear acquisition such as the lateral amygdala.

The observation that the dynamics of ERK pathways differ across brain regions involved

in fear memory, with peaks occurring at 20 min after stimulation in the hippocampus

versus 60 min after fear conditioning in the lateral amygdala (Schafe et al., 2000; Di

Benedetto et al., 2009; Wang et al., 2014), suggests that the model may be able to predict

optimal protocols targeting specific components of fear memory based on the dynamics

of intracellular signaling cascades in distinct brain regions. Future studies will test this

possibility by modeling the molecular cascades in the lateral amygdala and the medial

prefrontal cortex to preferentially target the acquisition and extinction of CS-associated

memories, respectively.

It is worth noting that the model is rather simplified. We did not consider many other

important molecular cascades critical for LTP and memory formation. For example,

calcium/calmodulin-dependent protein kinase II (CaMKII) and protein kinase C (PKC) are

also required for LTP induction (Smolen et al., 2020; Malenka et al., 1989; Wang et al.,

2016). In addition, we built the model with empirical data of PKA and ERK dynamics from

the literature, which are based on *ex vivo* analyses and have limited temporal resolution.

However, despite these limitations, the simplified model using essential biochemical

cascades for learning and memory was sufficient to enhance activation of the LTP-related

transcription factor CREB in the DG and facilitate associative learning in three different

experiments in mammals. A more complex model that incorporates a wider range of

intracellular and extracellular processes based on *in vivo* real-time data would likely

strengthen the predictive ability of simulations. Further experiments should also

investigate whether the memory enhancing effects of the optimal protocol vary across

subjects of different sexes and ages, as well as its efficacy in animal models of cognitive

impairment. Together, our results suggest the possibility of using similar model-driven,

non-invasive behavioral approaches in preclinical and clinical studies aimed at enhancing

learning or restoring memory deficits in humans.

Author Contributions: X.O.Z, D.S.E., and C.E.C performed and analyzed the behavioral

experiments. X.O.Z and C.E.C performed and analyzed the immunohistochemistry

experiments. Y.Z. implemented the computational model and ran all simulations. P.S.

helped design and implement the computational model. F.H.D-M and J.H.B supervised

and contributed to all aspects of this study. All the authors participated in the design of

the experiments. X.O.Z. and F.H.D-M interpreted the data and prepared the manuscript

with comments from all the co-authors.

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