Arithmetic and local circuitry underlying dopamine prediction errors

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Dopamine neurons are thought to facilitate learning by comparing actual and expected reward\(^1\).\(^2\). Despite two decades of investigation, little is known about how this comparison is made. To determine how dopamine neurons calculate prediction error, we combined optogenetic manipulations with extracellular recordings in the ventral tegmental area while mice engaged in classical conditioning. Here we demonstrate, by manipulating the temporal expectation of reward, that dopamine neurons perform subtraction, a computation that is ideal for reinforcement learning but rarely observed in the brain. Furthermore, selectively exciting and inhibiting neighbouring GABA (\(\gamma\)-amino butyric acid) neurons in the ventral tegmental area reveals that these neurons are a source of subtraction: they inhibit dopamine neurons when reward is expected, causally contributing to prediction-error calculations. Finally, bilaterally stimulating ventral tegmental area GABA neurons dramatically reduces anticipatory licking to conditioned odours, consistent with an important role for these neurons in reinforcement learning. Together, our results uncover the arithmetic and local circuitry underlying dopamine prediction errors.

Associative learning depends on comparing predictions with outcomes\(^3\). When outcomes match predictions, learning is not required. When outcomes violate predictions, animals must update their predictions to reflect experience. Dopamine neurons are thought to promote this process by encoding reward prediction error, or the difference between the reward an animal receives and the reward it expected to receive\(^4\) (see Supplementary Information).

Despite extensive study, how dopamine neurons calculate prediction error remains largely unknown. Theories of reinforcement learning predict that dopamine neurons perform subtraction, simply calculating actual reward minus predicted reward (or, in temporal difference models, the value of the current state minus the value of the previous state)\(^5\). However, dopamine neurons could also perform division, an equally fundamental and arguably more common neural computation\(^6\). The arithmetic underlying prediction errors has never been investigated.

To probe how dopamine neurons calculate prediction error, we recorded from the ventral tegmental area (VTA) (Extended Data Figs 1a and 2a–c) while mice (\(n = 5\)) performed a classical conditioning task with two interleaved trial types (Fig. 1a). In roughly half of the trials, we delivered reward unexpectedly, in the absence of any cue. In these trials, both the timing and size of reward were unexpected. In the other half of the trials, an odour cue predicted the timing of reward, but the size was still unexpected. By comparing responses to these two trial types, we could determine how temporal expectation modulates individual dopamine neurons across a range of firing rates. The light-gated ion channel, channelrhodopsin (ChR2), was expressed selectively in dopamine neurons, enabling us to identify neurons as dopaminergic on the basis of their responses to light\(^7\) (Extended Data Fig. 3a–g).

Consistent with previous results\(^8\), dopamine neurons increased their responses with increasing reward size (example neuron in Extended Data Fig. 4a). Much like sensory neurons in response to stimuli of increasing intensity, dopamine neurons showed a gradual, monotonic response, well-fit by a saturating Hill function (orange trace in Fig. 1c; note that VTA GABA neurons do not show the same monotonic response, Extended Data Fig. 5).

When reward was temporally expected, the responses of dopamine neurons were suppressed (\(P < 0.001\), \(t\)-test; example neuron in Extended Data Fig. 4a; population in Fig. 1b). To determine the nature of the suppression, we performed two complementary analyses. First, we fitted dopamine responses both with subtractive and with divisive models (Extended Data Fig. 6a). We found that subtraction was a significantly better fit (\(P < 0.001\), bootstrap; Fig. 1c and Extended Data Fig. 4b). Second, we plotted the effect of temporal expectation across reward sizes and measured the slope. A divisive process would produce a positive slope, as division should have a larger effect on larger dopamine responses. In contrast, subtraction would produce a slope near zero. We found the latter; regardless of reward size, the odour cue simply shifted the dose–response curve by a constant amount (\(P > 0.05\), linear regression, Fig. 1d). This subtractive pattern held not just for the population, but also for 35 out of 40 individual neurons (Extended Data Fig. 4c). Thus, consistent with classic theories of reinforcement learning, dopamine neurons appear to be performing subtraction (specifically, output subtraction\(^1\); see Extended Data Fig. 6a).

Having established the computation, we next wished to determine the input that dopamine neurons subtract. A variety of biological models have been proposed to explain the neural circuit required to calculate prediction errors. Some of these models have situated the calculation at the level of the dopamine neurons\(^9\),\(^10\), while others have suggested that the calculation happens upstream, for instance in the lateral habenula\(^11\),\(^12\), and is then relayed to dopamine neurons. Recently, we demonstrated that GABAergic neurons in the VTA encode reward expectation, showing sustained responses that vary with the timing and size of expected reward\(^13\). Although these neurons are known to synapse onto nearby dopamine neurons\(^14\) and appear to play a role in conditioned behaviour\(^14\),\(^15\), there has been no direct evidence that dopamine neurons use the VTA GABA signal for prediction-error calculations. Furthermore, although some models of prediction-error calculations call for a ramping expectation function\(^16\),\(^17\), which resembles VTA GABA activity, others call for phasic, precisely timed expectation signals\(^18\),\(^19\). Our study allows us to distinguish between these possibilities.

Since we know the normal firing patterns of VTA GABA neurons during classical conditioning\(^3\), our strategy was to mimic this firing and determine whether it induces subtraction of dopamine neuron responses. In a separate set of mice (\(n = 5\)), ChR2 was expressed selectively in VTA GABA neurons, enabling us to stimulate these neurons while recording from putative dopamine neurons (Extended Data Figs 1b and 2d–f). Much like the previous task, we unexpectedly delivered rewards of various size (Fig. 2a). In half of the trials, reward was delivered alone; in the other half, reward was delivered during 40 Hz VTA GABA stimulation.

First we confirmed that ChR2 stimulation efficiently excited VTA GABA neurons (\(P < 0.001\), paired \(t\)-test), adding about ten spikes per second to the baseline firing rate of the neurons (example neuron in Extended Data Fig. 4d; population in Fig. 2b). This laser-evoked
stimulation ($P < 0.05$; see Methods and Extended Data Fig. 6h, i). We conclude that VTA GABA activation mimics the effect of temporal expectation on putative dopamine neurons.

Although we show that VTA GABA activity can account for expectation-like changes in dopamine responses, this does not demonstrate that VTA GABA neurons normally play such a role. To strengthen the causal link between VTA GABA activity and dopamine prediction-error coding, we inhibited VTA GABA neurons during their normal period of activity and asked whether this disrupts dopamine prediction errors. In a separate group of mice ($n = 7$), the light-sensitive inhibitory proton pump archaerhodopsin (ArchT)$^{21}$ was expressed selectively in VTA GABA neurons (Extended Data Figs 1c, e–g and 9a–c). Mice were trained in a two-odour classical conditioning task, in which odour A predicted reward with $10\%$ probability and odour B predicted reward with $90\%$ probability (Fig. 3a). In $25\%$ of the trials, we delivered green laser to activate ArchT and inhibit VTA GABA neurons for $1\ s$ around reward outcome.

We first confirmed that laser stimulation significantly suppressed expectation-related activity in putative VTA GABA neurons ($P = 0.001$, t-test; individual neurons in Extended Data Fig. 4f, h; population in Fig. 3b). Next, we assessed how inhibiting VTA GABA neurons modified dopamine activity. Normally, putative dopamine neurons had reduced reward responses when a cue predicted reward delivery ($P < 0.001$, paired t-test; Fig. 3c). Inhibiting VTA GABA neurons partially reversed this expectation-dependent reduction (individual dopamine neurons in Extended Data Fig. 4g, i; population in Fig. 3d). Thus, when VTA GABA neurons are inhibited, dopamine neurons respond as if reward is less expected. This change was specific to phasic reward responses, and not due solely to a shift in baseline activity (Extended Data Figs 7c–h and 10). Combined with our ChR2 experiment, these results suggest that VTA GABA neurons play a causal role in dopamine prediction-error coding. In particular, they help provide the burst-cancelling expectation signal long anticipated by models of reinforcement learning.$^{16,18,22}$

In Figs 2 and 3 we report that VTA GABA excitation and inhibition modulates dopamine prediction-error responses. However, our unilateral optogenetic manipulation did not modify mouse behaviour (Extended Data Fig 8b, c). To determine whether the VTA GABA expectation signal is important for learning, we designed an additional experiment with bilateral manipulation. In a separate group of mice ($n = 6$), ChR2 was expressed selectively in VTA GABA neurons bilaterally. The mice performed a four-odour classical conditioning task, in which odour A was associated with large reward, odours B and D were associated with small reward, and odour C was associated with no reward (Fig. 4a). After training, odour D trials were paired with VTA GABA stimulation. Importantly, the odour–reward associations

**Figure 1 | Expectation triggers subtraction of dopamine neuron responses.** a, Dopamine (DA) identification recording setup (left) and task (right). ITI, inter-trial interval. b, Dopamine neuron firing rates (mean ± s.e.m. across neurons) for unexpected (orange) or temporally expected (black) reward. ***$P < 0.001$, t-test. c, Dopamine neuron responses (mean ± s.e.m.) to different reward sizes. Orange line, fit for unexpected reward. Dotted black line, divisive transformation. Solid black line, subtractive transformation. Subtraction was a better fit (**$P < 0.001$, bootstrap; see Methods and Extended Data Fig. 6e). d, Difference between unexpected and expected reward responses (mean ± s.e.m.) as a function of reward size.

**Figure 2 | Selective excitation of VTA GABA neurons mimics the effect of expectation.** a, GABA stimulation (stim.) recording setup (left) and task (right). b, Firing rate (mean ± s.e.m.) of putative VTA GABA neurons with (blue) and without (black) ChR2 stimulation. Light blue box, laser delivery. c, Firing rate (mean ± s.e.m.) of putative dopamine neurons. ***$P < 0.001$, t-test. d, Dopamine neuron responses (mean ± s.e.m.) to different reward sizes. Black line, fit for unexpected reward. Dotted blue line, divisive transformation. Solid blue line, subtractive transformation. Subtraction was a better fit ($P < 0.05$, bootstrap; see Extended Data Fig. 6g). e, f, Same as c and d, respectively, but in GFP-expressing control animals.
reduction remained significant on probe trials, where odour D was not though the reward was the same. Indeed, this is what we found: after and mice should lick less for odour D compared with odour B, even odour D. As a result, the expected value of odour D should decrease, always remained the same. Our hypothesis was that over time, laser interaction, mixed effects model.

Our study provides the first direct evidence for the arithmetic of dopamine prediction errors. Subtraction is an ideal process for prediction-error coding because it maintains a faithful separation between expected and unexpected rewards, even at the extremes of reward size (Extended Data Fig. 6a). Indeed, most, if not all, models of reinforcement learning have used subtraction to compute prediction error. However, although cortical pyramidal neurons appear capable of subtracting GABA input23–25 and modelling studies have explored the biophysics of this process26–28, surprisingly few examples of subtraction have been observed in natural settings in vivo19,30. Our finding that reward expectation reduces dopamine reward responses in a purely subtractive manner sheds light on how such a computation can emerge from a network of neurons, and may provide a framework for other prediction-related processes in the brain.

always remained the same. Our hypothesis was that over time, laser stimulation would reduce dopamine prediction-error responses for odour D. As a result, the expected value of odour D should decrease, and mice should lick less for odour D compared with odour B, even though the reward was the same. Indeed, this is what we found: after the laser was introduced, mice licked significantly less for odour D than for odour B (P < 0.001, laser × odour interaction, mixed linear model, Fig. 4b and Extended Data Fig. 8d). This reduction did not occur in a separate group of control mice (n = 6) that did not express ChR2 (Extended Data Fig. 8e). Although there was probably a direct effect of GABA stimulation on licking behaviour, as previously discovered14, this cannot account for the entire difference, because the reduction remained significant on probe trials, where odour D was not paired with laser stimulation (Fig. 4c).

Figure 3 | Selective inhibition of VTA GABA neurons modulates prediction errors. a, GABA inhibition (inhb.) recording setup (left) and task (right). b, Firing rate (mean ± s.e.m.) of putative VTA GABA neurons during odour B trials with (green) or without (black) laser delivery. ***P < 0.001, paired t-test. c, Firing rate (mean ± s.e.m.) of putative dopamine neurons when reward was delivered after odour A (orange) or odour B (black). ***P < 0.001, paired t-test. d, Same as b, but for putative dopamine neurons. ***P < 0.001, paired t-test.

Figure 4 | Bilateral excitation of VTA GABA neurons disrupts learned association. a, Schematic of optogenetic setup (left) and behavioural task (right). For a representative mouse (one of six mice injected with ChR2), anticipatory licks during each session (mean ± s.e.m. across trials) for odours A (black), B (dark grey), C (light grey), and D (blue). For sessions 12–17 (pale yellow), odour D was paired with laser. ***P < 0.001, laser × odour interaction, mixed effects model. c, Ratio of anticipatory licks for odour D versus odour B during laser sessions. Circles, mice injected with ChR2 (blue) or GFP (yellow). Open circles, probe trials, where laser was omitted after odour D. *P < 0.05; **P < 0.001; Wilcoxon rank-sum test.

Online Content Methods, along with any additional Extended Data display items and Source Data, are available in the online version of the paper, references unique to these sections appear only in the online paper.

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Supplementary Information is available in the online version of the paper.

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Author Information Reprints and permissions information is available at www.nature.com/reprints. The authors declare no competing financial interests. Readers are welcome to comment on the online version of the paper. Correspondence and requests for materials should be addressed to N.U. (uchida@mcb.harvard.edu).
METHODS

Animals. We used 33 adult male mice, backcrossed for more than 5 generations with C57BL/6J mice, that were heterozygous for Cre recombinase under the control of either the DAT gene (B6.SJ-Lcra1atm1(Cre)Jkk/J, The Jackson Laboratory)\(^\text{31}\) or the Vgat gene (B6.129S-Cre/J). We used 5 animals in the dopamine-identification task (Fig. 1), 7 in the GABA stimulation task (Fig. 2), 9 in the GABA inhibition task (Fig. 3), and 12 in the behavioural experiment (Fig. 4). Animals were housed on a 12 h dark/12 h light cycle (dark from 7:00 to 19:00) and performed the task at the same time each day. In the behavioural experiment, animals were randomly assigned to either the experimental or control groups, and the experiments were blinded to the assignment during all surgeries, behavioural sessions, and individual mouse analyses. All procedures were approved by the Harvard University Institutional Animal Care and Use Committee.

Surgery and viral injections. All surgeries were performed under aseptic conditions with animals under either ketamine/medetomidine (60 and 0.5 mg kg\(^{-1}\), intraperitoneal, respectively) or isoflurane (1–2% at 0.5–1.0 min\(^{-1}\)) anaesthesia. Analgesia (ketoprofen, 5 mg kg\(^{-1}\) intraperitoneally; buprenorphine, 0.1 mg kg\(^{-1}\), intraperitoneally) was administered postoperatively. For the recording experiments, mice underwent two surgeries, both stereotactically targetting left VTA (from bregma: 3.0 mm posterior, 0.8 mm lateral, –4.5 mm ventral). In the first surgery, we injected 200–500 nl adeno-associated virus (AAV) to enable cell-type identification or manipulation (see below). After 2–4 weeks, we performed a second surgery to implant a head plate and microdrive containing six to eight tetrodes and an optical fibre, as described\(^\text{32}\). Recording sites are displayed in Extended Data Fig. 1. For the behavioural experiment, mice underwent a single surgery in which we injected 500 nl AAV into VTA bilaterally, and then implanted a headplate and a dual-optic fibre cannula (300 \(\mu\)m diameter, Doric Lenses) custom-designed to target bilateral VTA.

The viral infections differed among the four experiments. In the dopamine identification experiment (Fig. 1), we injected AAV (serotype 5) carrying an inverted ChR2 (H134R) fused to the fluorescent reporter enhanced yellow fluoroscent protein (eYFP) and flanked by double loxp sites\(^\text{33}\). We previously showed that expression of this virus in dopamine neurons is highly selective and efficient\(^\text{34}\). In both the GABA stimulation experiment (Fig. 2) and the behavioural experiment (Fig. 4), we injected the same AAV–FLEX–ChR2–eYFP construct or, for control mice, we injected AA5V–GFP (University of North Carolina Vector Core). Finally, in the GABA inhibition experiment (Fig. 3), we injected AAV (serotype 1 or 8) carrying an inverted Arch\(^\text{35}\) fused to the fluorescent reporter GFP and flanked by double loxp sites (University of North Carolina Vector Core). Expression of ArchT was almost 100% selective to GABA neurons and about 50% efficient, for both AAV1 and AA8V (Extended Data Fig. 1e–g). In both the ChR2 and ArchT experiments, no virus-expressing cell bodies were observed distant from the injection site (for example, in the striatum or the cortex), implying that the virus was not taken up by axons in the VTA and transported retrogradely to input areas.

Behavioural tasks. After more than 1 week of recovery, mice were water-restricted in their cages. Weight was maintained above 90% of baseline body weight. Animals were head-restrained and habituated for 1–2 days before training. Odours were delivered with a custom-made olfactometer\(^\text{35}\). Each odour was dissolved in mineral water; and in the GABA stimulation experiment, but instead of delivering a reward-predicting odour, we used a blue laser to activate VTA GABA neurons directly. The experiment included three randomly interleaved trial types. In trial type 1 (5% of trials), rewards were delivered unexpectedly, in the absence of laser stimulation. Reward sizes were chosen pseudo-randomly from the following set: 0.3, 1.2, 2.5, 5, or 10 \(\mu\)l. Each reward size was equally frequent. In trial type 2 (5% of trials), rewards were also delivered unexpectedly, but now in the presence of laser stimulation. The laser was delivered at 40 Hz for a total of 1 s, and reward was delivered in the middle of this period. In trial type 3 (90% of trials), laser was delivered at 40 Hz for a total of 1 s, but no reward was delivered. The reason for the prevalence of trial type 3 was to ensure that mice did not associate the laser (which they might have seen, despite attempts to mask the light by painting the fibre black) with reward delivery.

In the GABA inhibition experiment (Fig. 3), each trial began with one of two odours, selected pseudo-randomly. One odour predicted water reward with 10% probability and the other odour predicted water reward with 90% probability. In 25% of these trials, 1 s of continuous green laser was administered, beginning at odour offset and lasting until 0.5 s after reward was delivered. This encompassed both the delay between odour and reward (1–1.5 s) and the reward response period (1.5–2 s), which are the times in which VTA GABA neurons normally fire\(^\text{34}\). Laser stimulation did not affect licking behaviour (Extended Data Fig. 8c). At the beginning and end of each recording session, we delivered 1 s periods of green laser without any odours or rewards, to assess how GABA inhibition modulated dopamine baseline activity.

The behavioural experiment (Fig. 4) included four trial types, each associated with a different odour. The four trial types were pseudo-randomly interleaved and equally likely. Odour A was associated with big reward (5 \(\mu\)l), odours B and D were associated with small reward (2 \(\mu\)l), and odour C was associated with no reward. After training, when the mice consistently associated the odours with reward (as demonstrated by their anticipatory licking behaviour), blue laser was paired with odour D trials. Laser was delivered for 2.5 s, beginning 0.5 s after odour onset and ending 0.5 s after reward onset. The intensity of light was modulated in a ramping fashion (see below). After six to eight sessions using the laser, the laser was turned off for the remaining four or five sessions, allowing us to examine whether the effect of laser stimulation would persist even in the absence of laser. Additionally, to clarify whether behaviour changes reflected learning or a direct effect of VTA GABA stimulation on licking, we included probe trials in the final two or three laser-primed sessions. During these probe sessions, 10% of odour B trials randomly delivered laser stimulation, 10% of odour D trials randomly avoided laser stimulation.

Electrophysiology. Recording techniques were based on a previous study\(^\text{9}\). Briefly, we recorded extracellularly from VTA using a custom-built, screw-driven microdrive containing six or eight tetrodes (Sandvik) glued to a 200 \(\mu\)m optic fibre (ThorLabs). Tetrodes were affixed to the fibre so that their tips extended 300–600 \(\mu\)m from the end of the fibre. Neural and behavioural signals were recorded with a DigiLynx recording system (Neuralynx) or a custom-built system using a multi-channel amplifier chip (RHA2116, Intan Technologies) and data acquisition device (PCIe-6351, National Instruments). Broadband signals from each wire were filtered between 0.1 and 9,000 Hz and recorded continuously at 32 kHz. To examine single-unit activity, signals were band-pass-filtered between 300 and 6,000 Hz and sorted offline using SpikeSort3D (Neuralynx) or MClast-3.5 (A. D. Redish). At the end of each session, the fibre and tetrodes were lowered by 40–80 \(\mu\)m to record new units the next day.

To be included in the data set, a neuron had to be well isolated (\(I_{\text{ratio}}\) (ref. 36) < 0.05) and recorded within 0.5 mm of a light-identified or putative dopamine neuron, to ensure that it was recorded in VTA. Recording sites were also verified histologically with electrolytic lesions using 10–15 s of 30 \(\mu\)A direct current.

Laser delivery. To identify neurons as dopaminergic or GABAergic, we used ChR2 to observe laser-triggered spikes\(^\text{35,36}\). The optical fibre was coupled with a diode-pumped solid-state laser with analogue amplitude modulation (Laserlight Technologies). At the beginning and end of each recording session, we delivered trains of ten blue (473 nm) light pulses, each 5 ms long, at 1,10, and 20 Hz, with an intensity of 5–20 mW mm\(^{-2}\) at the tip of the fibre. Spike shape was measured using a broadband signal (0.1–9,000 Hz) sampled at 32 kHz.
In the GABA stimulation experiment (Fig. 2), we used the same blue laser to deliver 40 pulses (duration 5 ms, 40 Hz) during selected trials. In the GABA inhibition experiment (Fig. 3), we used one of two methods of laser delivery. For seven mice (Extended Data Fig. 9a–c), we used an electronic shutter (Vincent Associates) to deliver 1 s intervals of continuous green laser (532 nm, Laserglow Technologies), with an intensity of –50 mW mm−2 at the tip of the fibre. For a separate group of two mice (Extended Data Figs 9d–f and 10), we instead modulated laser intensity in an analogue fashion, beginning at zero intensity 0.5 s after odour onset, smoothly increasing, or decreasing to peaking at 50 mW mm−2 at reward delivery, and then gradually decreasing over the next 0.5 s. This ramping protocol was also used for the behavioural experiment (Fig. 4), with a 473 nm laser (OptoEngine) and beam splitter (Doric Lenses) to deliver blue light bilaterally. The ramping intensity profile was chosen to approximate the response pattern of VTA GABA neurons.

Data analysis. Peristimulus time histograms were constructed using 1 ms bins and then convolved with a function resembling a postynaptic potential, (1 − exp(−t)/2) × (exp(−t)/20)), for time t in milliseconds. Average firing rates in response to reward were calculated using a 600 ms window after reward onset for the dopamine identification and GABA stimulation experiments, and a 500 ms window after reward onset for the GABA inhibition experiment. These windows were chosen to reflect the full duration of the neural response to reward. Window sizes ranging from 300 to 1,000 ms were attempted and gave qualitatively similar results. To calculate reward response, we subtracted baseline firing (averaged over 1 s before trial onset). Calculating the baseline using different windows (for example, 600 ms before reward onset) did not change the results. To ensure reliability, analyses of particular trial types only included neurons that were recorded during at least five presentations of that trial type.

To identify neurons as dopaminergic or GABAergic, we used the stimulus-associated spike latency test6 to determine whether light pulses significantly changed a neuron’s spike timing (Extended Data Fig. 3). We used a significance value of P < 0.001. To ensure that spike sorting was not contaminated by light artefacts, we also calculated waveform correlations between spontaneous and light-evoked spikes, as described8. All light-identified neurons had Pearson’s correlation coefficients > 0.9. In all three recording experiments, we identified putative dopamine and GABA neurons on the basis of their firing patterns through an unsupervised clustering approach (Extended Data Figs 2 and 9), similar to a previous study9. Briefly, receiver-operating characteristic (ROC) curves for each neuron were calculated by comparing the distribution of firing rates across trials in 100 ms bins (starting 1 s before expected reward and ending 1 s after expected reward) to the distribution of baseline firing rates (1 s before trial onset). Principal component analysis was calculated using the singular value decomposition of the area under the ROC. Hierarchical clustering was then done using the first three principal components of the area under the ROC using a Euclidean distance metric and complete agglomeration method. As described8, this method produced three clusters: one with phasic excitation to reward (type 1), one with sustained excitation to reward expectation (type 2), and one with sustained suppression to reward expectation (type 3). Type 1 neurons were classified as putatively dopaminergic. Forty out of 43 light-identified dopamine neurons fell into this cluster; the other three light-identified dopamine neurons showed phasic suppression to reward and were clustered as type 3. Since these three dopamine neurons showed qualitatively different responses than the others, they were not included in the data set. Note that although we focus on identified dopamine neurons, our main findings are identical if we include all putative dopamine neurons (Extended Data Fig. 6b, c).

Type 2 neurons were classified as putatively GABAergic. Of 14 identified GABA neurons, 11 were clustered as type 2; the other three were inhibited by reward and were clustered as type 3. Again, these three GABA neurons were not included in the data set. Unlike type 1 neurons, type 2 neurons did not respond to either expected or unexpected reward in a consistently size-dependent fashion (Extended Data Fig. 5). This contrasts with their delay activity, which increases with increasing reward expectation10. The distribution of neurons across mice for all recording experiments is provided in Supplementary Table 1. To determine the dose–response of dopamine neurons and see whether expectation caused a subtractive or divisive effect (Fig. 1c), we based our analysis on a previous study30. We first fitted a hyperbolic ratio function (Hill function) to the unexpected reward data:

\[
f(r) = f_{\text{max}} \left( \frac{r^\alpha}{r^\alpha + \sigma^\alpha} \right)
\]

The function had two free parameters: \(f_{\text{max}}\), the saturating firing rate; and \(\sigma\), the reward size that elicits half-maximum firing rate. We chose an exponent of 0.5 after fitting the data with exponents ranging from 0.1 to 2.0 (in steps of 0.1), and finding the exponent with the lowest mean squared error. Note that the Hill function is not the only possible function that could fit our data. For example, the power function \(f(r) = ar^k\), where \(a = 3.73\) and \(k = 0.39\), was also excellent. However, this function does not saturate, so we thought it was less likely to represent neuronal responses. Our conclusions do not depend on the exact function chosen to fit the data.

After fitting the unexpected reward data, we explored what simple transformation could best mimic the effect of expectation. We tested four options: input subtraction, input division, output subtraction, and output division (Extended Data Fig. 6a). Specifically, we evaluated the following four models:

- **Input subtraction**: \(f(r) - f_{\text{max}} \left( \frac{r-E}{r^\alpha + \sigma^\alpha} \right)\)
- **Input division**: \(f(r) = f_{\text{max}} \left( \frac{r^\alpha}{r^\alpha + \sigma^\alpha} \right)\)
- **Output subtraction**: \(f(r) = f_{\text{max}} \left( \frac{r^\alpha}{r^\alpha + \sigma^\alpha} - E \right)\)
- **Output division**: \(f(r) = \frac{1}{E} \left[ \frac{r^\alpha}{r^\alpha + \sigma^\alpha} \right]\)

In each case, we used the \(f_{\text{max}}\) and \(\sigma\) values determined by the unexpected reward data. The only new parameter was the expectation factor \(E\), which we fitted separately for each of the four models. Output subtraction consistently gave the best fit (lowest mean squared error), for the population and most individual neurons. The next best model was generally output division. We statistically compared model fits using a bootstrapping analysis: we resampled the data 1,000 times and determined for each resample the mean squared error for both output subtraction and output division. We calculated the \(P\) value by counting the number of resamples when the mean squared error was better for output division than for output subtraction (for example, if 1 resample out of 1,000 preferred output division over output subtraction, \(P = 0.001\); Extended Data Fig. 6c). These steps were repeated for putative dopamine neurons in the GABA stimulation experiment (Fig. 2d).

As a complementary analysis to determine whether expectation had a subtractive or divisive effect on dopamine reward responses, we calculated the difference between unexpected and expected reward responses for different reward sizes (Fig. 1d). We then ran a linear regression to determine whether the slope of this difference was significantly different from zero. A slope of zero would be consistent with output subtraction, as expectation would have the same effect on all responses. A slope greater than zero would be consistent with output division, as expectation would have a larger effect on larger responses. All but five of the light-identified dopamine neurons had a slope different from zero (Extended Data Fig. 4c).

In our GABA stimulation and inhibition experiments, we wanted to ensure that laser delivery affected phasic dopamine responses in addition to shifting baseline dopamine activity. First, we identified putative dopamine neurons that did not significantly change their baseline firing upon laser delivery. To do so, we compared firing rates in the 0.5 s before reward delivery on laser trials versus no-laser trials. Neurons with \(P > 0.05\) (Wilcoxon rank-sum) were identified as unaffected by laser delivery. In both the GABA stimulation and GABA inhibition experiments, these neurons continued to be affected at the time of reward (Extended Data Fig. 7a, c). Second, we recorded from putative dopamine neurons while manipulating VTA GABA activity outside the task (Extended Data Fig. 7c, g). This gave us an unbiased sense of how VTA GABA stimulation or inhibition affected dopamine baseline responses. We then subtracted these laser-alone trials from trials where laser was delivered during reward (Extended Data Fig. 7b, f). Any remaining change at the time of reward should not be due to a baseline shift.

Interestingly, the baseline shift may have been an artefact of the type of laser stimulation we applied. In a separate experiment (\(n = 2\) mice, Extended Data Fig. 9d–l), we applied the laser so that light intensity would ramp up rather than remain constant over the course of a trial, more closely mimicking the physiologically responses of VTA GABA neurons. We found that this ramping stimulation successfully inhibited putative VTA GABA neurons (\(P = 0.001\), t-test, Extended Data Fig. 10a, b) and increased reward responses in putative dopamine neurons (\(P < 0.001\), t-test, Extended Data Fig. 10c, d) without causing a baseline shift.
To assess how well VTA GABA stimulation mimics odour expectation, we also directly compared the magnitude of change in dopamine responses in both experiments. In the odour-based experiment (Fig. 1), the average suppression of dopamine reward responses was 52.7%, compared with 43.5% for the VTA GABA stimulation experiment (Fig. 2; \(P < 0.05\), t-test). This difference may be accounted for by variation among putative dopamine neurons in their response to laser. Although 40 out of 45 putative dopamine neurons were suppressed by GABA stimulation, 5 were activated, perhaps through disinaptic disinhibition, as VTA GABA neurons are known to synapse onto each other as well as onto dopamine neurons\(^2\). In addition, there may be other neurons, besides VTA GABA neurons, that help suppress dopamine responses when reward is expected.

Although we focus on changes in the magnitude of dopamine neuron responses, this was not the only effect of reward expectation. Notably, the latency to peak response was also extended, from an average of 67.6 ms to 95.4 ms (\(P = 0.001\), t-test). The latency increased in 37 out of 40 dopamine neurons that we recorded. The downstream consequences of this change in latency remain to be elucidated.

Comparisons were performed with t-tests (for population data) or Wilcoxon rank-sum tests (for individual neuron data), with corrections for multiple comparisons (Bonferroni or Tukey). Correlations were done with Pearson's rho. Values less than 0.05 were considered significant, unless otherwise noted. Given pilot data showing effects of optogenetic manipulation of about 2 spikes per second, with variability of about 3 spikes per second, 36 neurons were required for 80% power to detect the effect. Given about ten neurons of each type per mouse, we aimed for at least four mice per experiment. Analyses were done with Matlab (Mathworks).

In the behavioural experiment (Fig. 4), the strength of the learned association between each odour and reward was estimated by counting the number of anticipatory licks over the 2 s from odour onset to reward delivery. For the analysis in Fig. 4b and Extended Data Fig. 8d, e, we excluded data from probe trials. Population results were examined using a mixed-effects linear model. The fixed effects included trial type and a binary variable indicating whether the session included laser delivery. The random effect was mouse identity. The outcome of interest was an interaction between trial type and laser. Results were robust to different choices of window for counting anticipatory licks.

**Immunohistochemistry.** After recording for 4–8 weeks, mice were given an overdose of ketamine/medetomidine, exsanguinated with saline, and perfused with 4% paraformaldehyde. Brains were cut in 100 µm coronal sections on a vibratome and immunostained with antibodies to tyrosine hydroxylase (AB152, 1:1,000, Millipore) to visualize dopamine neurons and 49,6-diamidino-2-phenylindole (DAPI, Vectashield) to visualize nuclei. Viral expression was determined through eYFP fluorescence. Slides were examined to verify that the optic fibre track was among VTA dopamine neurons and in a region expressing the virus. For the GABA inhibition experiment, two Vgat-tetTomato mice were injected with AAV–FLEX–ArchT–GFP to determine the selectivity and efficiency of ArchT expression in VTA GABA neurons (Extended Data Fig. 1e–g). One mouse was injected with AAV serotype 1 and the other with AAV serotype 8. For the figure, brightness and contrast were adjusted in Adobe Photoshop.

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Extended Data Figure 1 | Recording sites and ArchT expression.

a–d, Schematic of recording locations for mice used in the dopamine identification task (a, n = 5), the GABA stimulation task (b, n = 7), the GABA inhibition task (c, n = 9), and the behavioural task (d, n = 12). b, Red, experimental mice expressing ChR2 in VTA GABA neurons (n = 5). Blue, control mice expressing GFP in VTA GABA neurons (n = 2). c, Red, mice in which laser was delivered at continuous intensity (n = 7). Blue, mice in which laser was delivered with ramping intensity (n = 2). d, Red, experimental mice expressing ChR2 in VTA GABA neurons (n = 6). Blue, control mice expressing GFP in VTA GABA neurons (n = 6).

Magenta, Vgat-tdTomato; green, ArchT–GFP. Open arrow, neuron expressing Vgat-tdTomato but not ArchT–GFP. Closed arrow, neuron expressing both Vgat-tdTomato and ArchT–GFP. Scale bar, 10 μm. f, Selectivity of infection to GABA neurons: percentage of ArchT–GFP-expressing neurons (n = 131 neurons for AAV1 and 165 neurons for AAV8) that were positive for Vgat-tdTomato. Filled bars, Vgat-tdTomato mouse injected with AAV1–FLEX–ArchT–GFP. Empty bars, Vgat-tdTomato mouse injected with AAV8–FLEX–ArchT–GFP. g, Efficiency of infection: percentage of Vgat-tdTomato-expressing neurons (n = 278 neurons for AAV1 and 283 neurons for AAV8) that were positive for ArchT–GFP.

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Extended Data Figure 2 | Neuron classification for dopamine identification and GABA stimulation experiments. a–c, Dopamine identification experiment. d–f, ChR2-expressing animals in GABA stimulation experiment. g–i, GFP-expressing control animals in GABA stimulation experiment. a, d, g, Responses of all VTA neurons recorded in the tasks. Each row reflects the area under the ROC values for a single neuron in the second before and after delivery of expected reward. Baseline is taken as 1 s before odour onset. Yellow, increase from baseline; cyan, decrease from baseline. Light-identified neurons are denoted by an asterisk to the left of each column. b, e, h, The first three principal components of the area under the ROC curves. These values were used for unsupervised hierarchical clustering, as shown in the dendrogram on the right. c, f, i, Average firing rates for the three clusters of neurons in each task. Odour was delivered for 1 s, followed by a 0.5 s delay and then reward delivery.
Extended Data Figure 3 | Light identification of dopamine and GABA neurons. 

a, Raw signal from one example light-identified dopamine neuron. Blue bars, light pulses. 
b, For the same neuron, mean waveforms for spontaneous (black) and light-evoked (blue) action potentials. 
c, For the same neuron, raster plots for 20 Hz (left) and 50 Hz (right) laser stimulation. Each row is one trial of laser stimulation.
d, Histogram of log \( P \) values for each neuron recorded in the dopamine identification experiment \((n = 170)\). The \( P \) values were derived from the stimulus-associated spike latency test (see Methods). 
e, Neurons with \( P < 0.001 \) and waveform correlations > 0.9 were considered identified (filled bars). 
f, For light-identified neurons, probability of spiking (e) and latency to first spike (f) after laser pulses at different frequencies. Orange circles, mean across neurons. 
g, Histogram of mean latencies (left) and latency standard deviations (right) in response to laser stimulation for all light-identified dopamine neurons in the variable-reward task. 
h–n, Same conventions as a–g, but for neurons recorded in the GABA stimulation task \((n = 102)\).
Extended Data Figure 4 | Individual neuron analysis from all recording experiments. a–c, Results from dopamine identification experiment (Fig. 1). d, e, Results from GABA stimulation experiment (Fig. 2). f–i, Results from GABA inhibition experiment (Fig. 3). a, Raster plots (top and middle) and firing rate (bottom) of representative dopamine neuron in response to unexpected (orange) or temporally expected (black) reward. ***P < 0.001, t-test. b, For the same neuron, responses (mean ± s.e.m. across trials) to each reward size. Orange line, fit for unexpected reward. Dotted black line, divisive transformation. Solid black line, subtractive transformation. c, Individual neuron regression slopes for the analysis in Fig. 1d. Empty bars, slope not different from zero (P > 0.05). Filled bars, P < 0.05. Triangle, mean slope. d, e, Firing rate of representative VTA GABA (d) and putative dopamine (e) neuron with (blue) and without (black) ChR2 stimulation. Light blue box, laser delivery. f, g, Firing rate of representative VTA GABA (f) and putative dopamine (g) neuron during odour B trials with (green) or without (black) laser delivery. h, i, Histogram of putative GABA (h) and dopamine (i) neuron responses to laser delivery. Filled bars, significant effect of laser (P < 0.05, Wilcoxon rank-sum); empty bars, P > 0.05. Triangle, mean.
Extended Data Figure 5 | VTA GABA activity does not vary consistently with reward size. a–c, Putative GABA neurons in the dopamine identification experiment (Fig. 1). a, b, Average firing rate of putative GABA neurons to unexpected (a) or temporally expected (b) rewards of various sizes. c, Population responses (mean ± s.e.m. across putative GABA neurons) for different reward sizes. Orange, unexpected reward. Black, temporally expected reward. Responses were averaged over a 600 ms window after reward delivery.

d–f, Putative GABA neurons in the GABA stimulation experiment (Fig. 2). d, e, Average firing rate of putative GABA neurons to rewards of various sizes, delivered with (e) or without (d) optogenetic GABA stimulation. f, Population responses (mean ± s.e.m. across putative GABA neurons) for different reward sizes. Blue, reward with laser stimulation. Black, reward without laser stimulation. Responses were averaged over a 600 ms window after reward delivery.
**a**

**M.S.E. subtraction - M.S.E. division**

---

**SUBTRACT**

- **OUTPUT**
  - **Dopamine response**
  - **Reward size**

---

**DIVIDE**

- **INPUT**
  - **Dopamine response**
  - **Reward size**

---

**b**

**Putative DA neurons**

- **Response (spikes/s)**
  - **Reward size (μL)**
  - **n = 84**

---

**c**

**# resamples**

- **M.S.E. subtraction - M.S.E. division**
  - **P < 0.001**

---

**d**

**Identified DA neurons**

- **Response (spikes/s)**
  - **Reward size (μL)**
  - **n = 40**

---

**e**

**# resamples**

- **M.S.E. subtraction - M.S.E. division**
  - **P < 0.001**

---

**f**

**Putative DA neurons**

- **Response (spikes/s)**
  - **Reward size (μL)**
  - **n = 45**

---

**h**

**Putative DA neurons, subtracting immediate baseline**

- **Response (spikes/s)**
  - **Reward size (μL)**
  - **n = 45**

---

**g**

**# resamples**

- **M.S.E. subtraction - M.S.E. division**
  - **P = 0.015**

---

**i**

**# resamples**

- **M.S.E. subtraction - M.S.E. division**
  - **P = 0.041**

---
Extended Data Figure 6 | Statistical test for subtraction versus division.

a, To understand how dopamine neurons compute reward prediction error, we first determined how dopamine neurons respond to various sizes of unexpected reward (schematized as orange curves). We then taught the mice to expect reward and observed how expectation shifted this dose–response (black curves). We modelled four types of shift: output subtraction (top left), input subtraction (bottom left), output division (top right), and input division (bottom right). Output subtraction was consistently the best fit. For equations, see Methods. Analysis adapted from a previous study39.

b–e, Results from dopamine identification experiment. b, c, Results from all putative dopamine neurons (n = 84). ***P < 0.001, bootstrap.

d, e, Results from light-identified dopamine neurons (n = 40). ***P < 0.001, bootstrap.

f, g, Results from putative dopamine neurons in the GABA stimulation experiment (n = 45). *P < 0.05, bootstrap.

h, i, Results from putative dopamine neurons in the GABA stimulation experiment, subtracting the 500 ms period immediately before reward delivery. This takes into account the laser-induced baseline shift in dopamine responses. *P < 0.05, bootstrap.

f, g, Results from putative dopamine neurons in the GABA stimulation experiment, subtracting the 500 ms period immediately before reward delivery. This takes into account the laser-induced baseline shift in dopamine responses. *P < 0.05, bootstrap.

h, i, Results from putative dopamine neurons in the GABA stimulation experiment, subtracting the 500 ms period immediately before reward delivery. This takes into account the laser-induced baseline shift in dopamine responses. *P < 0.05, bootstrap.

Average responses (mean ± s.e.m. across neurons) to different sizes of reward, with fits for output subtraction (solid line) and output division (dotted line). c, e, g, i, Results of bootstrapping analysis. For each resample, we compared the mean squared error for the subtractive fit with the mean squared error for the divisive fit. Negative numbers favour subtraction. P values were calculated as the proportion of resamples in which division was a better fit than subtraction.
Extended Data Figure 7 | Laser effect is more than a baseline shift.

a–d, Results from GABA stimulation experiment. e–h, Results from GABA inhibition experiment.

a, Firing rate (mean ± s.e.m.) of putative dopamine neurons that did not show a significant baseline shift. ***P < 0.001, t-test. b, To visualize whether GABA stimulation preferentially affected phasic dopamine responses in addition to baseline firing rates, we took the activity in Fig. 2c and subtracted the trials when laser was delivered alone. Any remaining change at the time of reward could not be due to a baseline shift. **P < 0.01, t-test.

c, Firing rate (mean ± s.e.m.) of putative dopamine (left) and GABA (right) neurons in trials where laser was delivered in the absence of reward. This dopamine response was subtracted to calculate the firing rates in b. d, Histogram of the phasic effect of GABA stimulation. The values were calculated by subtracting the black line from the blue line in b. Empty bars, slope not different from zero (P > 0.05, Wilcoxon rank-sum test). Filled bars, slope different from zero (P < 0.05). Triangle, mean (P < 0.001, t-test).

e–h, Same conventions as a–d, but for the GABA inhibition experiment. ***P < 0.001, t-test.
**Extended Data Figure 8 | Behavioural performance on all four experiments.**

a, In the dopamine identification task (Fig. 1), lick rates (mean ± s.e.m. across sessions) for odours predicting reward (black) or nothing (grey). b, In the GABA stimulation task (Fig. 2), lick rates (mean ± s.e.m. across sessions) for reward alone (black), reward + GABA stimulation (blue), or GABA stimulation alone (orange). c, In the GABA inhibition task (Fig. 3), lick rates (mean ± s.e.m. across sessions) for the odours predicting reward with 90% probability (black) and 10% probability (grey). Green laser was delivered to inhibit VTA GABA neurons in 25% of reward (green) and nothing (orange) trials. d, e, In the bilateral stimulation experiment (Fig. 4), anticipatory licks (mean ± s.e.m. across mice) for mice injected with ChR2 (d) and GFP (e). Grey bars, odour B; blue or orange bars, odour D. Left, last three training sessions before odour D was paired with laser; middle, last three sessions with laser delivery (excluding probe trials); right, last three sessions after laser was turned off. **P < 0.01; ***P < 0.001; paired t-test.
Extended Data Figure 9 | Neuron classification for GABA inhibition experiment. a–c, Mice in which laser was delivered with continuous intensity. d–f, Mice in which laser was delivered with ramping intensity. a, d, Responses of all VTA neurons recorded in the tasks. Each row reflects the area under the ROC values for a single neuron in the second before and after delivery of expected reward. Baseline is taken as one second before odour onset. Yellow, increase from baseline; cyan, decrease from baseline. b, e, The first three principal components of the area under the ROC curves. These values were used for unsupervised hierarchical clustering, as shown in the dendrogram on the right. c, f, Average firing rates for the three clusters of neurons in each task. Odour was delivered for 1 s, followed by a 0.5 s delay and then reward delivery.
Extended Data Figure 10 | Ramping laser stimulation eliminates baseline shift. a, Firing rate (mean ± s.e.m.) of putative VTA GABA neurons during odour B trials with (green) or without (black) ramping laser delivery. ***P < 0.001, t-test. b, Histogram of putative GABA neuron responses to laser delivery. Responses were averaged over the entire duration of the laser. Filled bars, significant effect of laser (P < 0.05, Wilcoxon rank-sum test); empty bars, P > 0.05. Triangle, mean (P < 0.001, t-test). c, Firing rate (mean ± s.e.m.) of putative dopamine neurons with (green) or without (black) ramping GABA inhibition. ***P < 0.001, t-test. d, Histogram of putative dopamine neuron responses to laser delivery. Responses were averaged over the 0.5 s window after reward delivery. Filled bars, significant effect of laser (P < 0.05, Wilcoxon rank-sum test); empty bars, P > 0.05. Triangle, mean (P < 0.001, t-test).
Erratum: Arithmetic and local circuitry underlying dopamine prediction errors
Neir Eshel, Michael Bukwich, Vinod Rao, Vivian Hemmelder, Ju Tian & Naoshige Uchida

In this Letter, the x-axis labels of Fig. 3b and d were incorrect; for both panels, the labels should read: “Time from odour onset (s)”. Figure 3 has been corrected in the HTML and PDF versions online.