SUPPLEMENTARY INFORMATION

Controllability Governs the Balance Between Pavlovian and Instrumental Action Selection

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Supplementary Methods

The Bayesian model used in this study was derived as follows. The generative model of an uncontrollable environment assumed by the Pavlovian system is given by:

$$\theta_{s} \sim Beta\left(\theta_{0}\frac{\eta_{0}}{2}, (1-\theta_{0})\frac{\eta_{0}}{2}\right)$$
(1)
$$r|s \sim Bernouli(\theta_{s})$$

where θ_s is the reward probability for stimulus s, θ_0 is the mean of the prior, η_0 is a parameter controlling the dispersion of the prior, and r is the reward outcome on a particular trial (all reward outcomes are assumed to be independently and identically distributed). The generative model of a controllable environment assumed by the instrumental system is essentially the same, except conditioned on action, a:

$$\theta_{sa} \sim Beta\left(\theta_0 \frac{\eta_0}{2}, (1-\theta_0) \frac{\eta_0}{2}\right)$$
(2)

$$r|s, a \sim Bernouli(\theta_{sa})$$

For both systems, the posterior conditional on the stimulus-action-reward history \mathcal{D} is a Beta distribution with the same functional form as the prior, shown here for the Pavlovian system:

$$P(\theta_{s}|\mathcal{D}, m = uncontrollable) = Beta(\hat{\theta}_{s}\frac{\eta_{s}}{2}, (1 - \hat{\theta}_{s})\frac{\eta_{s}}{2})$$
(3)
$$\hat{\theta}_{s} = \frac{\eta_{0} + N_{s}}{\eta_{s}}$$

$$\eta_{s} = \eta_{0} + T_{s}$$

where *m* indexes the assumed environment (uncontrollable for the Pavlovian system, controllable for the instrumental system), N_s is the number of times stimulus *s* was paired with reward, and T_s is the number of times stimulus *s* was presented. The equations for the instrumental system are identical except for conditioning on both stimuli and actions. Through simple algebraic manipulation, the update for the posterior mean $\hat{\theta}_s$ can be expressed as a recursive learning rule (Eq. 2 in the main text).

The posterior over generative models is given by:

$$P(m = uncontrollable | \mathcal{D}) \propto P(\mathcal{D} | m = uncontrollable) P(m = uncontrollable).$$
(4)

The marginal likelihood is an integral over the latent parameters:

$$P(\mathcal{D}|m = uncontrollable) = \int P(\mathcal{D}|m = uncontrollable, \theta_s) P(\theta_s) d\theta_s \qquad (5)$$
$$= \frac{B(\hat{\theta}_s \frac{\eta_s}{2}, (1 - \hat{\theta}_s) \frac{\eta_s}{2})}{B(\theta_0 \frac{\eta_0}{2}, (1 - \theta_0) \frac{\eta_0}{2})}$$

where *B* denotes the beta function. The beta function can be expressed recursively:

$$B(x+1,y) = B(x,y)\frac{x}{x+y}$$
(6)
$$B(x,y+1) = B(x,y)\frac{y}{x+y}$$

Applying these recursions to the log posterior odds over m, one obtains the update rule for L in the main text (Eq. 6).



Supplementary Figures

Supplementary Figure 1. Probability of Go response across trials for each experimental condition and stimulus, smoothed with a 5-trial moving average. Error bars show standard error of the mean.



Supplementary Figure 2. Go bias for Experiment 1 (N = 271) and Experiment 2 (N = 183) when no subjects are excluded. For both experiments, there is a strong Go bias but no effect of controllability. Error bars show standard error of the mean.



Supplementary Figure 3. Bias-variance analysis for purely Pavlovian (w = 1) and purely instrumental (w = 0) agents. Error bars show standard error of the mean.

Supplementary Table

	Inverse	Prior mean	Prior	Prior mean	Prior
	temperature	(instrumental)	confidence	(Pavlovian)	confidence
	-		(instrumental)		(Pavlovian)
Experiment 1	11.3949	0.3774	33.3939	0.4848	34.4666
Experiment 2	10.4635	0.2189	31.5312	0.4628	29.4448

Supplementary Table 1. Average parameter estimates for the adaptive model.