

Early Results: Pavlov and his Dog



- <u>Classical</u> (<u>Pavlovian</u>) conditioning experiments
- **↑** Training: Bell → Food
- ◆ After: Bell → Salivate
- Conditioned stimulus (bell) predicts future reward (food)

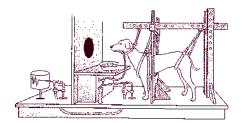


Image: Wikimedia Commons; Animation: Tom Creed, SJU

Predicting Delayed Rewards

- → How do we predict rewards delivered some time after a stimulus is presented?
- → Given: Many trials, each of length T time steps
- → Time within a trial: $0 \le \underline{t} \le T$ with stimulus $\underline{u(t)}$ and reward r(t) at each time step t (Note: r(t) can be zero for some t)
- ♦ We would like a neuron whose output v(t) predicts the expected total future reward starting from time t

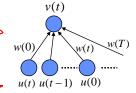
$$v(t) pprox \left\langle \sum_{\tau=0}^{T-t} r(t+\tau) \right\rangle_{trials}$$

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Learning to Predict Future Rewards

◆ Use a set of synaptic weights w(t) and predict based on all past stimuli u(t):

$$v(t) = \sum_{\tau=0}^{t} w(\tau) \underline{u(t-\tau)}$$
(Linear filter!)



♦ Learn weights $w(\tau)$ that minimize error:

$$\left(\sum_{\tau=0}^{T-t} r(t+\tau) - \underline{v(t)}\right)^2$$

(Can we minimize this using gradient descent and delta rule?)

Yes, BUT future rewards are not yet available!



Temporal Difference (TD) Learning

★ Key Idea: Rewrite error function to get rid of future terms:

$$\left(\sum_{\tau=0}^{T-t} r(t+\tau) - v(t)\right)^{2} = \left(\underline{r(t)} + \sum_{\tau=0}^{T-t-1} r(t+1+\tau) - v(t)\right)^{2}$$

$$\approx \left(\underline{r(t)} + v(t+1) - v(t)\right)^{2} \quad \text{Minimize this using gradient descent!}$$

→ Temporal Difference (TD) Learning:

$$\Delta w(\tau) = \underline{\varepsilon} \left[r(t) + v(t+1) - v(t) \right] u(t-\tau)$$
Expected future reward Prediction

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