ASHI636: Advanced Topics in Neuroscience Dr. Olav E. Krigolson

> Lecture 1 Learning



## Our Course

- Jan 28<sup>th</sup> Learning
- Feb 4<sup>th</sup> Decision-Making
- Feb 11<sup>th</sup> Executive Control
- Feb 18<sup>th</sup> NO CLASS (READING BREAK)
- Feb 25<sup>th</sup> Emotion, Good, and Evil
- Mar 3<sup>rd</sup> Consciousness
- Mar 10<sup>th</sup> Computational Neuroscience OR Neuroscience Laboratory Experience

## Disclaimers

 Scope of course Eric Kandel Principles of Neuroscience 67 Chapters, 7 Appendices, 1720 pages

- 2. Level of detail Background Time
- 3. Medicine

4. What having a PhD really means

## Quick Review

What is Neuroscience? Cellular Level Neuron Level Systems Level

What did we cover in Introductory Neuroscience?







### Animals ranked by total number of neurons in the brain Number of neurons

Human		86.06 billion
Baboon	10.95	
Pig	2.23	
Raven	2.17	
Grey parrot	1.57	
Owl monkey	1.48	
Emu	1.34	
Barn owl	0.69	
Starling	0.483	
Squirrel	0.479	
Blackbird	0.38	
Goldcrest	0.16	
Mouse	0.07	

### So what happens when we learn?







Increased neurotransmitter release Increase receptors Structural changes



In practice, we talk about the strength of a connection in terms of a "weight" or a "value".





#### 1. Before conditioning



Food

response

Unconditioned stimulus



Salivation

Unconditioned response







## Rescorla - Wagner

# Expectancy







Expectancy



Expectancy

## **Prediction Error**

When the actual outcome is different than the expected outcome

# Value





# Vpunishment





## States

### Current State







### **Previous State**

### Current State



## **Prediction Error**

## The difference in VALUE between the current state and preceding state

$$\mathsf{PE} = (\mathsf{V}_{\mathsf{reward}} - \mathsf{V}_{\mathsf{cue}})$$

 $PE = (V_{current state} - V_{preceding state})$ 

## Learning IS ALWAYS a two step process.

### At each point in time we:

Calculate a prediction error
Update the previous value

# $V_{\text{cue new}} = V_{\text{cue old}} + PE$


## $V_{tone} = 0$ $V_{reward} = 100$

PE = (100 - 0)PE = 100

# $V_{cue new} = V_{cue old} + PE$ $V_{cue new} = 0 + 100$



### $V_{tone} = 100$ $V_{reward} = 100$

PE = (100 - 100)PE = 0

### $V_{cue new} = V_{cue old} + PE$ $V_{cue new} = 100 + 0$

Learning Rates

 $V_{\text{cue new}} = V_{\text{cue old}} + PE * LR$   $V_{\text{cue new}} = 0 + 100 * 0.2$   $V_{\text{cue new}} = 0 + 20$ 



Trial	V <sub>cue</sub>	V <sub>reward</sub>	PE	PE x LR
1	0	100	100	20
2	20	100	80	16
3	36	100	64	12.8
4	48.8	100	51.2	10.24
5	59.04	100	40.96	8.192

## V<sub>reward</sub>



V<sub>cue</sub>



#### Prediction Error x Learning Rate













Another Example...

























#### This also applies to motor skills!

Recall that a skill is a collection of neurons being activated.

PE's can be used in principle to strengthen the connections between these neurons to "learn" the correct movement pattern.











$$PE = V_{outcome} - V_{action}$$
$$PE = +$$





$$V_{action} = V_{action} + PE * LR$$
$$V_{action} \uparrow$$







## Thus, we choose the top set of neurons because this choice has a higher VALUE





## This is a basic principle of decision making – always choose the highest value option



### Part II


Olds & Milner, 1954 Routtenberg & Malsbury, 1969 Crow, 1972





#### Correct trials (reward)









# **Classical Conditioning**







# **Classical Conditioning**





















# Thorndike's Law of Effect

Actions that are followed by feelings of satisfaction have a greater likelihood of being generated again in the future, whereas actions that are followed by feelings of dissatisfaction have a lesser likelihood of being generated again in the future.



Early in learning,  $V_S = 0$ Thus, when juice is given:  $\delta(t) = r(t) + \gamma V(t) - V(t - 1)$  $\delta(t) = 1 + 0 - 0$ 



After learning,  $V_S > 0$ Thus, when juice is given:  $\delta(t) = r(t) + \gamma V(t) - V(t - 1)$  $\delta(t) = 1 + 0 - 1$  $\delta(t) = 0.0$ 

# However, when the<br/>stimulus is encounteredSR $V_{S-1} = 0$



 $\delta(t) = r(t) + \gamma V(t) - V(t - 1)$ 

$$\delta(t) = 0 + 1 - 0$$

 $\delta(t) = 1.0$ 

# And when a predicted reward does not occur:



$$\delta(t) = r(t) + \gamma V(t) - V(t - 1)$$
  
$$\delta(t) = 0 + 0 - 1.0$$
  
$$\delta(t) = -1.0$$

# Part III

+

#### $\mathbf{H} \, \mathbf{H} \, \mathbf{H} \, \mathbf{H} \, \mathbf{H}$

+

#### HHSHH

# The Error-Related Negativity



+

#### Beep!

+

#### **Correct!**

### The Feedback Related Negativity





Miltner et al., 1997; Holroyd et al., 1998



#### Anterior Cingulate Cortex is the probable source of the ERN

EEG: Debener et al., 2005; Holroyd et al., 2004 Monkey: Ito et al., 2003



#### (Holroyd & Coles, 2002)



(Holroyd & Coles, 2002)


(Holroyd & Coles, 2002)









Holroyd & Krigolson, 2007

# The Role of Medial-Frontal System in the Acquisition of Perceptual Expertise

Krigolson, Pierce, Tanaka, & Holroyd, 2009

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#### **Correct!**







## **Correct!**



















## Propagation of the Prediction Error



Before Learning



Learned Stimulus/ Reward Mapping

Krigolson et al., In press

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Krigolson et al., 2014













### **Behavioral Accuracy**





#### **Choice Presentation**

#### **Reward Delivery**



DATA



MODEL


+





+

## Win!

