

## Introduction:

- Roving is when two different stimulus types are randomly interleaved during training.
- Roving can impair learning.
- Prior studies have speculated that roving is related to disruption of memory-traces (Yu et al., 2004), diminishment of stimulus predictability (Adini et al., 2004), or prevention of conceptual tagging (Zhang et al., 2008).
- Here we show that roving has a much more parsimonious, mathematical explanation.

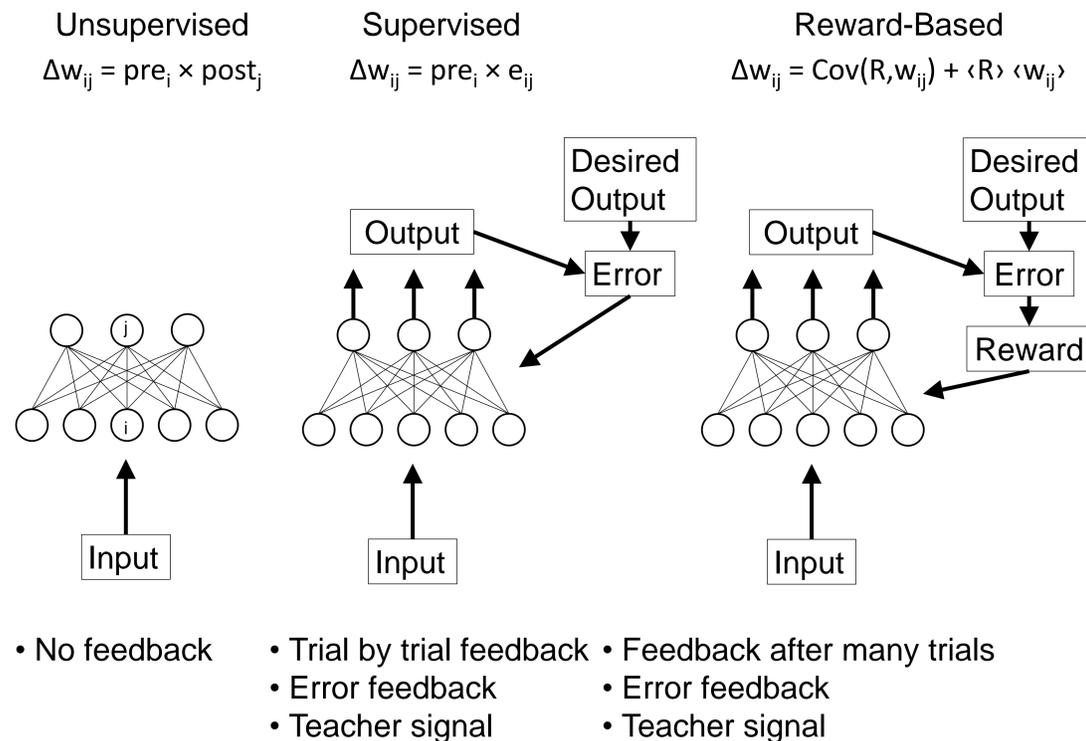


Figure 1. Unsupervised models fail because feedback facilitates learning. Supervised models fail because learning is possible without feedback. **Hence, only reward-based learning can explain perceptual learning.**

## Unsupervised Bias Hypothesis:

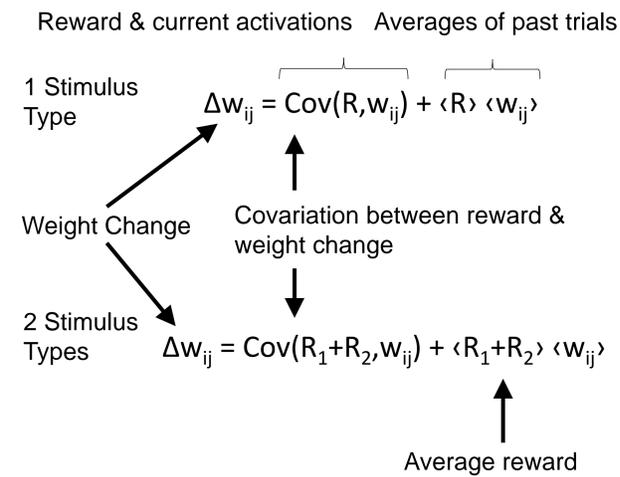


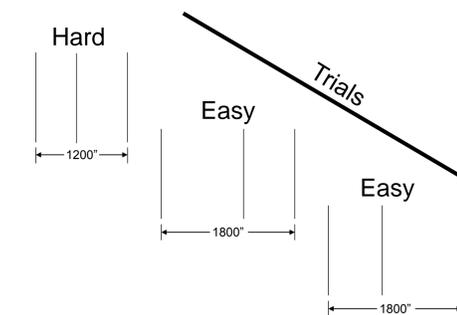
Figure 2. Perceptual learning can occur only if  $\langle R \rangle = 0$  or  $\langle R_1 + R_2 \rangle = 0$ , respectively otherwise synaptic drift occurs.

## Predictions:

- The unsupervised bias hypothesis predicts that perceptual learning fails when an easy and a hard task are roved because of their different rewards.

## Experiment:

- Bisection task
- We first measure pre-training thresholds for a 1200" and an 1800" bisection stimulus.
- During training, participants are presented with both the 1200" and the 1800" stimuli randomly interleaved.



## Results:

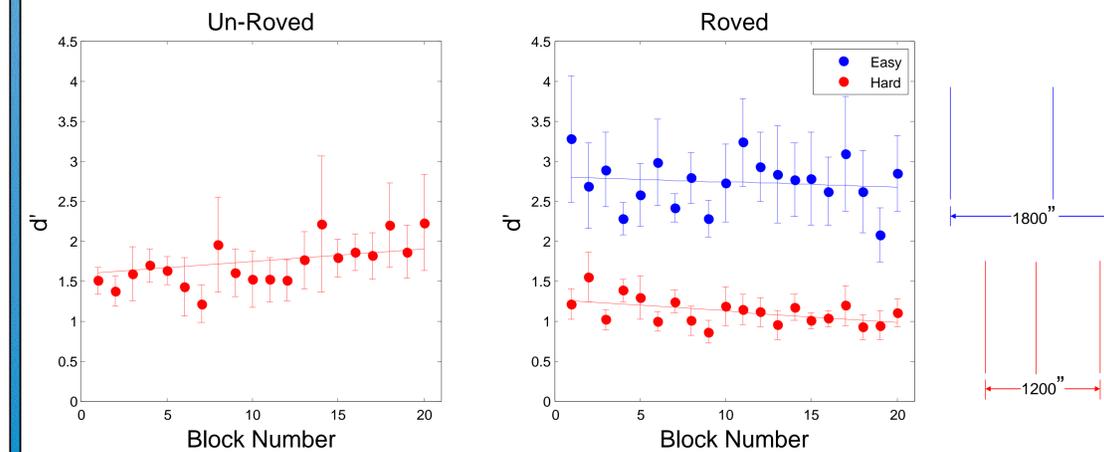


Figure 3. While learning is possible for a difficult, un-roved task (left), no learning occurs when the same task is roved with an easy task (right).

## Summary:

- A parsimonious mathematical analysis shows why roving can impair learning.
- Our experiment provides evidence for the unsupervised bias hypothesis..