# Choice reflexes in the rodent (and human) sensorimotor striatum

#### A new mechanism to promote exploration?



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# Rapid response adaptation following poor reinforcements





Neural basis:

- change in value?
- high stochasticity?
- specific mechanism?

# Rapid response adaptation following poor reinforcements



Lose-switch:



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'Sensorimotor habit'
S-R associations
slow to change
devaluation insensitive

 'deliberative'
A-O associations
te) rapid change devaluation sensitive

#### Competitive Task: 2 x 2 matrix game ('Matching Pennies')

- $C_n^{p_1} = C_n^{p_2}$ : Player 1 wins
- $C_n^{p_1}!=C_n^{p_2}$ : Player 2 wins



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- Optimal strategy against strong player is a 'mixed strategy' (random responding)
- Expected win probability is 0.5 against an optimal opponent

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Prediction: 'habits' in DLS will produce patterned responses & poor performance



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- Counter to devaluation experiments
- Led us to hypothesize that:
  - 1) Loss information in DLS is short lived
  - 2) Involves negative reward prediction error signal by dopamine
  - 3) General feature of DLS processing; same features in humans

### Lose-Switch decays and is independent of Win-Stay

#### Lose-Switch and Win-Stay are:

- Prevalent
- Uncorrelated
- Change with inter-trial interval (ITI)



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# Summary

- LS and WS are predominant, persistent
- LS and WS are **uncorrelated**
- LS depends on **negative RPE** in sensorimotor striatum
- LS emerges in humans under cognitive load
- LS decays over ~7-8 seconds in rats, and >10 s in humans
  - Much shorter than devaluation

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# **Implications & Speculations:**

- Confound in experiments, could improve RL model fits to behaviour
- LS may solve ethologically-relevant tasks
- LS will immediately and briefly promote exploration independent of changes in value or stochasticity
- LS may pause the current response policy so that other computationallyexpensive systems can determine a new policy



# Wild Speculations:

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• Ramifications on network dynamics:





# Ongoing:

• Cellular mechanisms:



• Ramifications on network dynamics:





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• Ramifications on network dynamics:



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