

# RL model fitting

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# Outline

- Fitted RL models to the behavioural data maximizing log likelihood to find parameters
- Collapsed all data across sessions (performance was similar on all sessions) but not across subjects
- Q learning based models fit very well (predict choice with 98% accuracy)
- Mice do not appear to exhibit significant side biases, or Q learning decay
- But do find substantial differences between free and forced choice and some choice stickiness

# Fitting procedure

- Computed Log likelihood for RL model on the behavioural choice data across all sessions
- Fitted parameters of model to maximize log likelihood, picked best of 5 random initializations
- Computed log likelihood only on free choices but still updated model on forced
- Since only 2 ports available in a trial, set Q value of other to large negative value

## Q learning accuracy

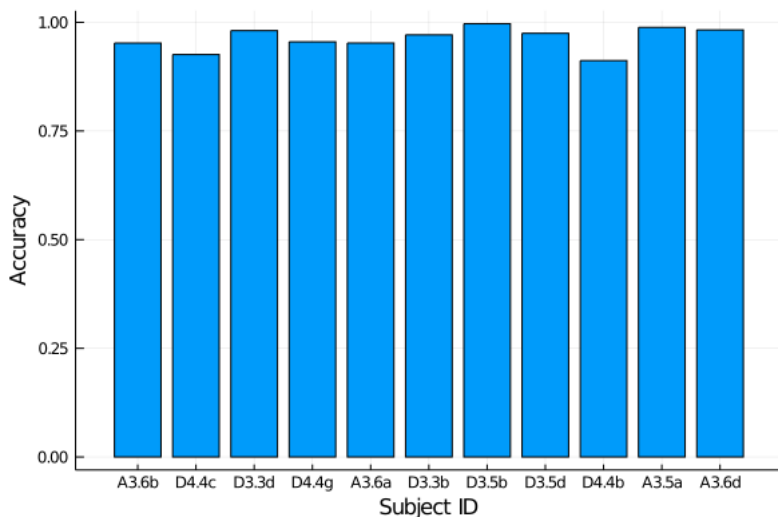
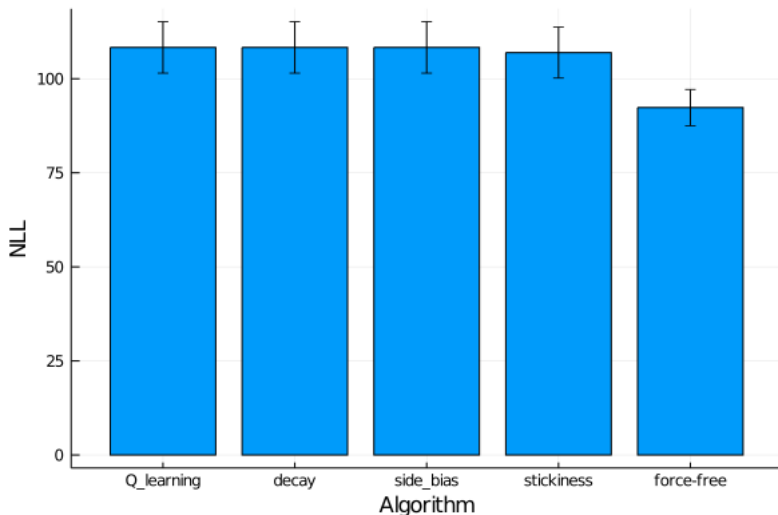


Figure: Q learning models for all subjects reach a high accuracy

# Models

- Standard Q learning updates a Q function of 3 values (one for each port) with standard Q learning so Q values are updated  $Q(x)+ = \alpha\delta(x)$  and actions probabilities are computed by  $p(a|x) = \sigma(\beta Q(x))$
- Q learning decay introduces an additional decay parameter  $Q(x)+ = \alpha\delta(x) - \gamma Q(x)$  where  $\gamma$  is the decay parameter
- Q learning with port bias introduces 3 bias terms (one for each port) so that action probabilities are  $p(a|x) = \sigma(\beta Q(x) + b)$
- Q learning with stickiness adds a bias in action selection towards previous choice  $p(a|x) = \sigma(\beta Q(x) + \eta p(a_{t-1}|x_{t-1}))$
- Q learning with forced and free learning rates updates the Q values with different learning rates  $\alpha_{force}$  and  $\alpha_{free}$  for forced and free choice trials

## Log likelihoods



**Figure:** Most models reach the same log likelihood and are not significant. There is small evidence of choice stickiness and large evidence of differing learning rates.

# Next steps

- Fit Go/No-go models (I think this will also have very high accuracy)
- Regress Q values and Go and No Go values against the neural photometry data
- Comparison of fits in D1 and D2 to try to find evidence for the non-monotonicity predicted