

Go- no-go decisions based on gradually revealed visual information

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Control of manual movements...

... is well described as a decision making process.

Gallivan et al 2019, Nat Neurosci Reviews

... gradually changes with the decision variable.

Selen et al 2012, J Neurosci

... rapidly adjusts to new information even after movements are launched.

Nashed et al 2014, J Neurosci

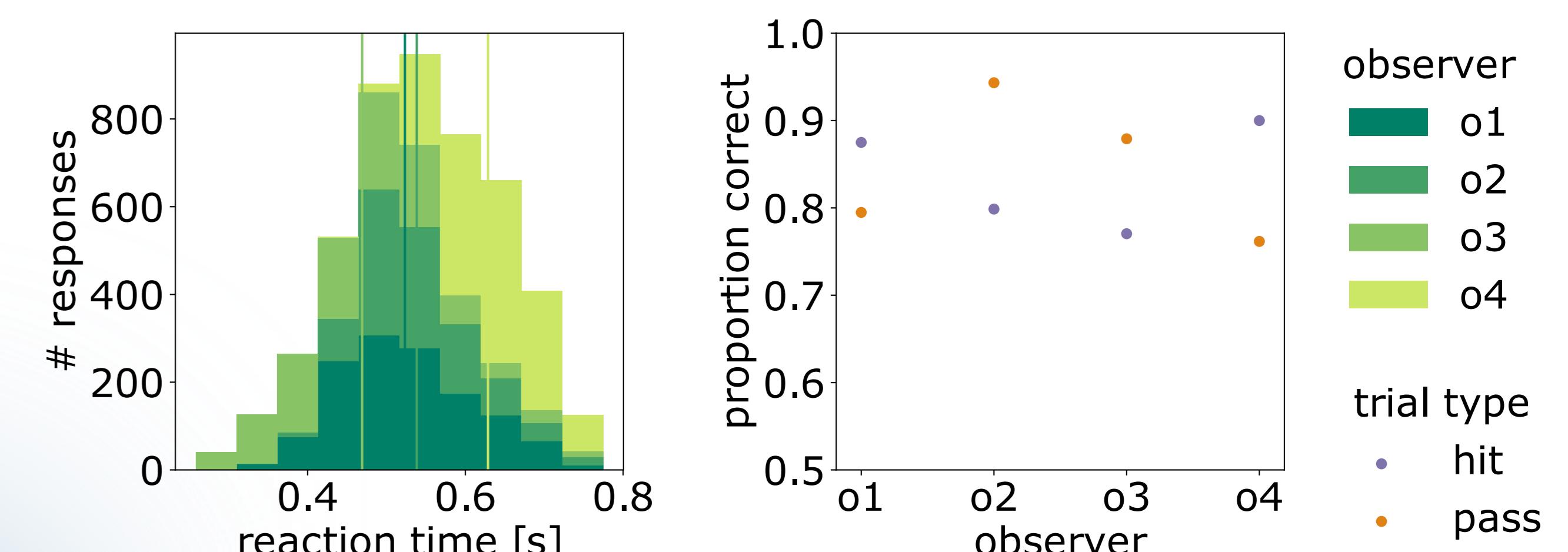
But what if new evidence requires a different response?

Does this also happen during response preparation?

How is new visual evidence used for the preparation of manual movements?

We collected data in a pilot study.

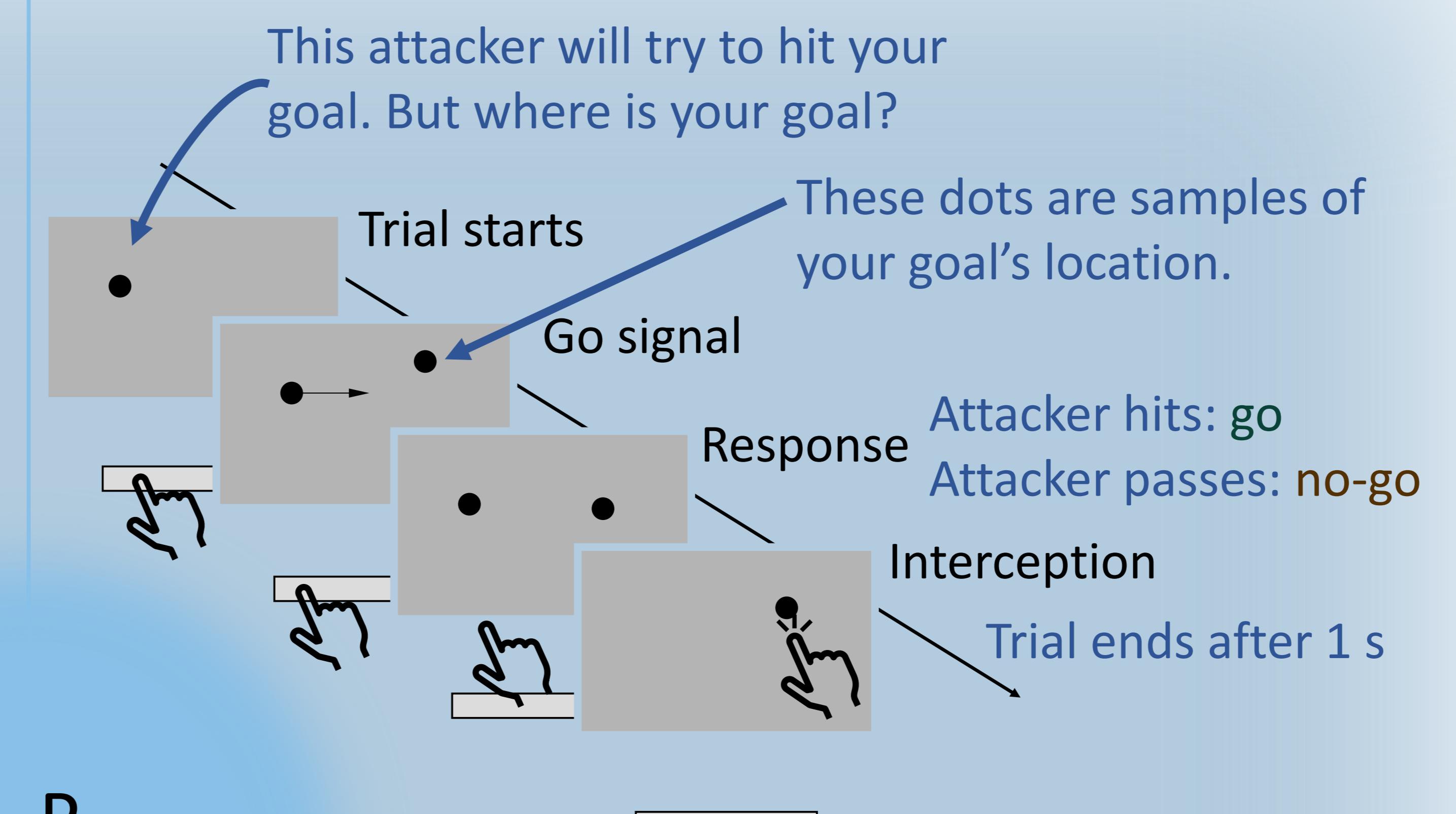
Observers: 4
Sessions: 3
Trials/session: 800
Final Datapoints: 9589



The hit probability $p[H]$ for each sample is given by the cumulative density function of a uniform distribution, with parameters that could be known by the observer:

$$cdf(x) = \begin{cases} 0 & \text{for } x > a \\ \frac{x-a}{b-a} & \text{for } x \in [a, b] \\ 1 & \text{for } x <= b \end{cases}$$

x: absolute distance between attacker and sample
a: width of the goal
b: sample position, normalized to zero.



Question P
Behaviour M
Paradigm Mechanism

Human observers adjust to early new evidence, but not to late evidence.

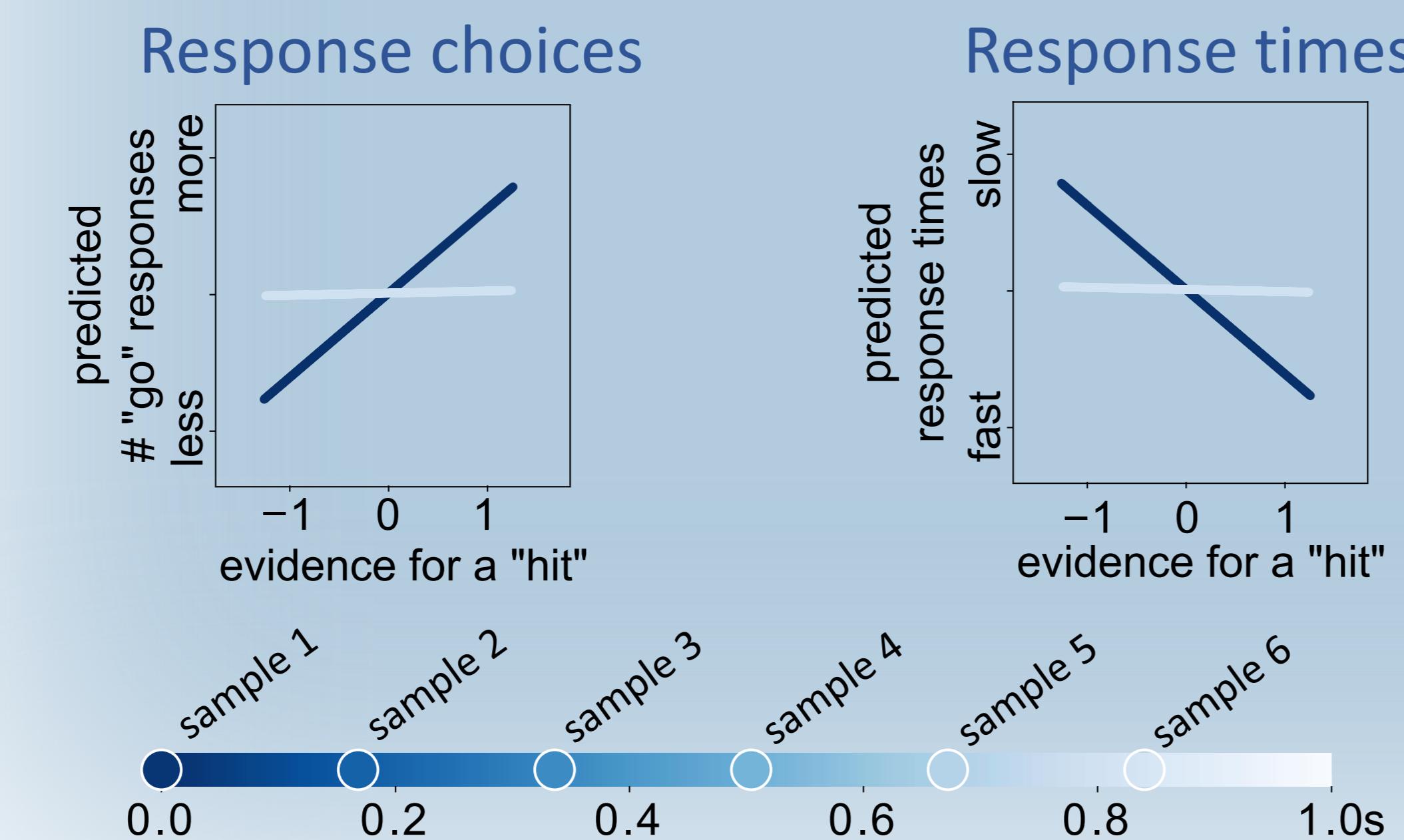
Full models included:

$p[H]$ of each sample (sample 1-6)

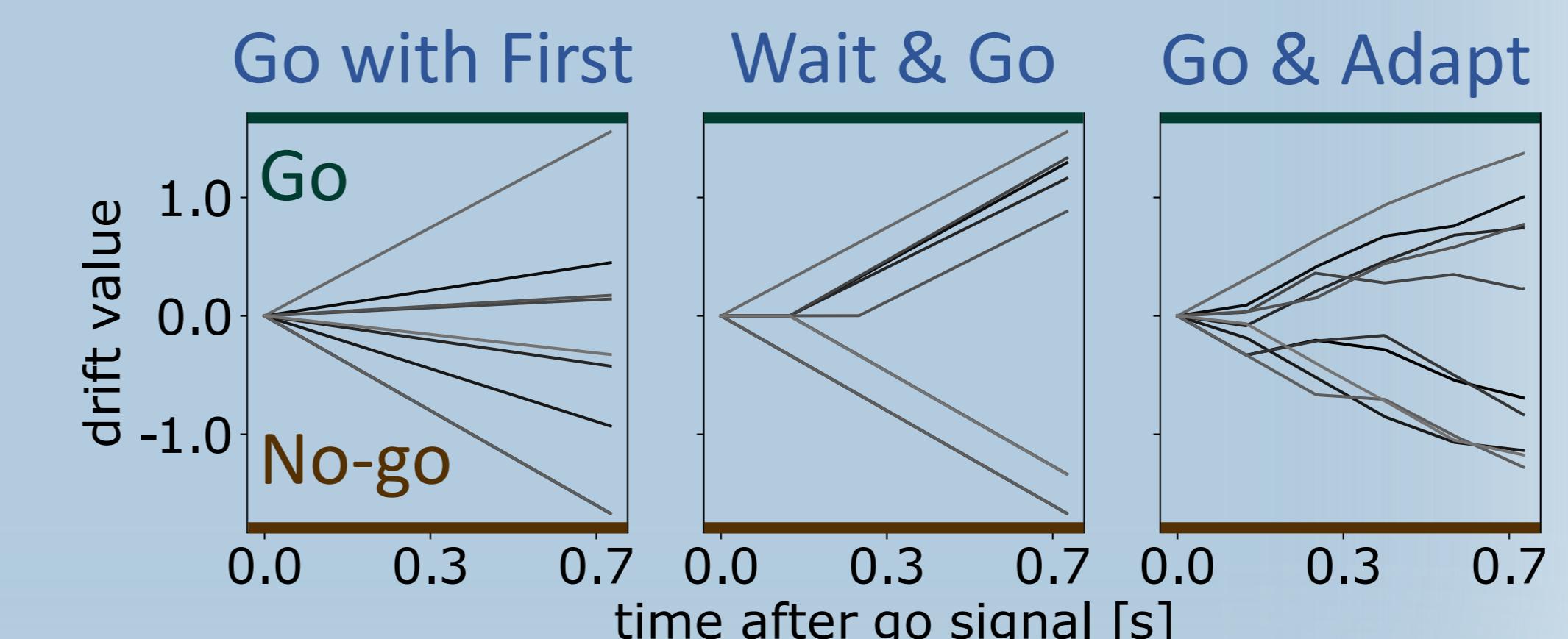
Trial condition (hit or pass trial)

Fully specified random effects per subject

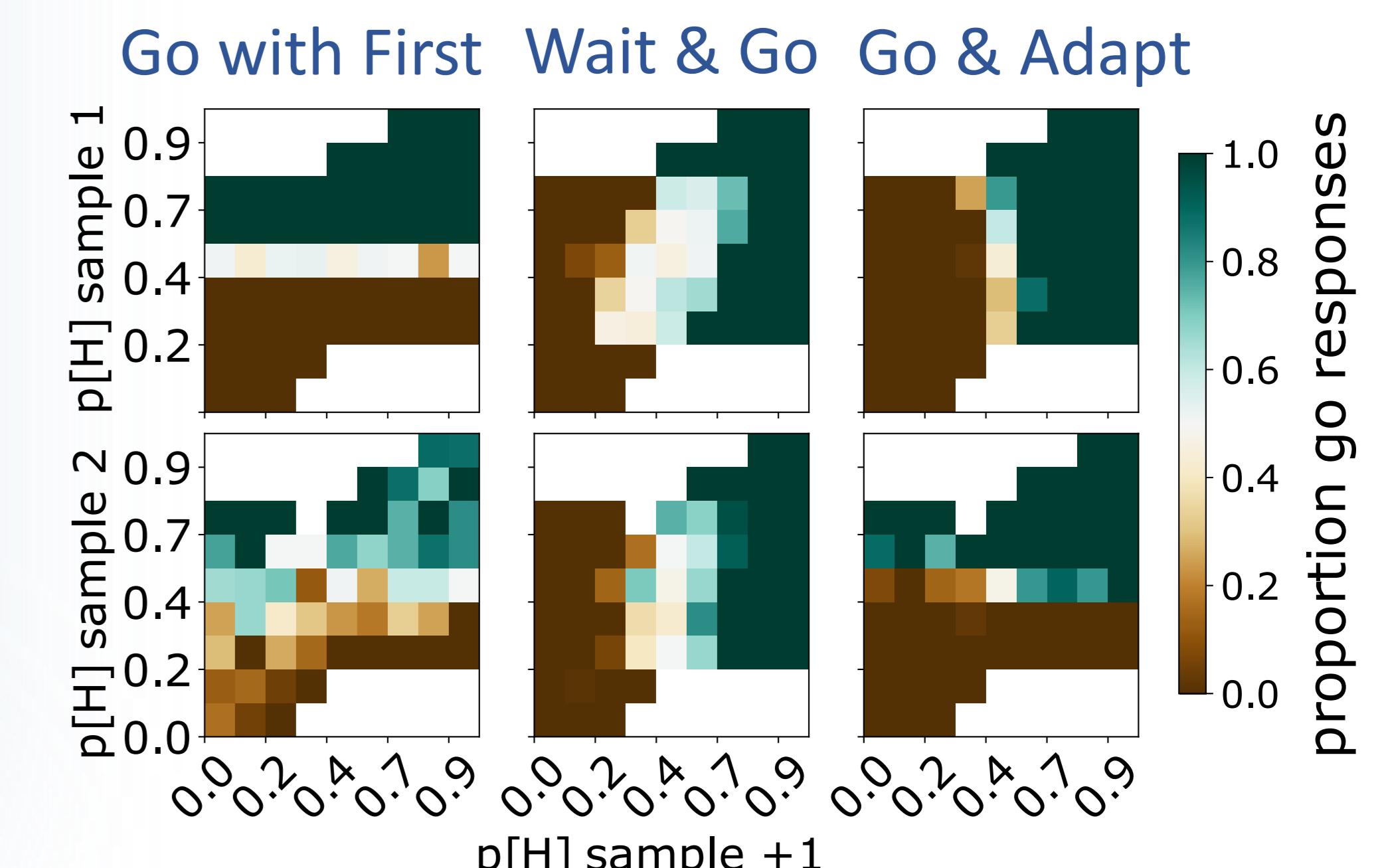
Linear regression models describe response behaviour.



Drift diffusion models describe how motor plans evolve in time.



Not all models adjust well to new evidence.



Response update patterns of human observers relative to the response onset ...

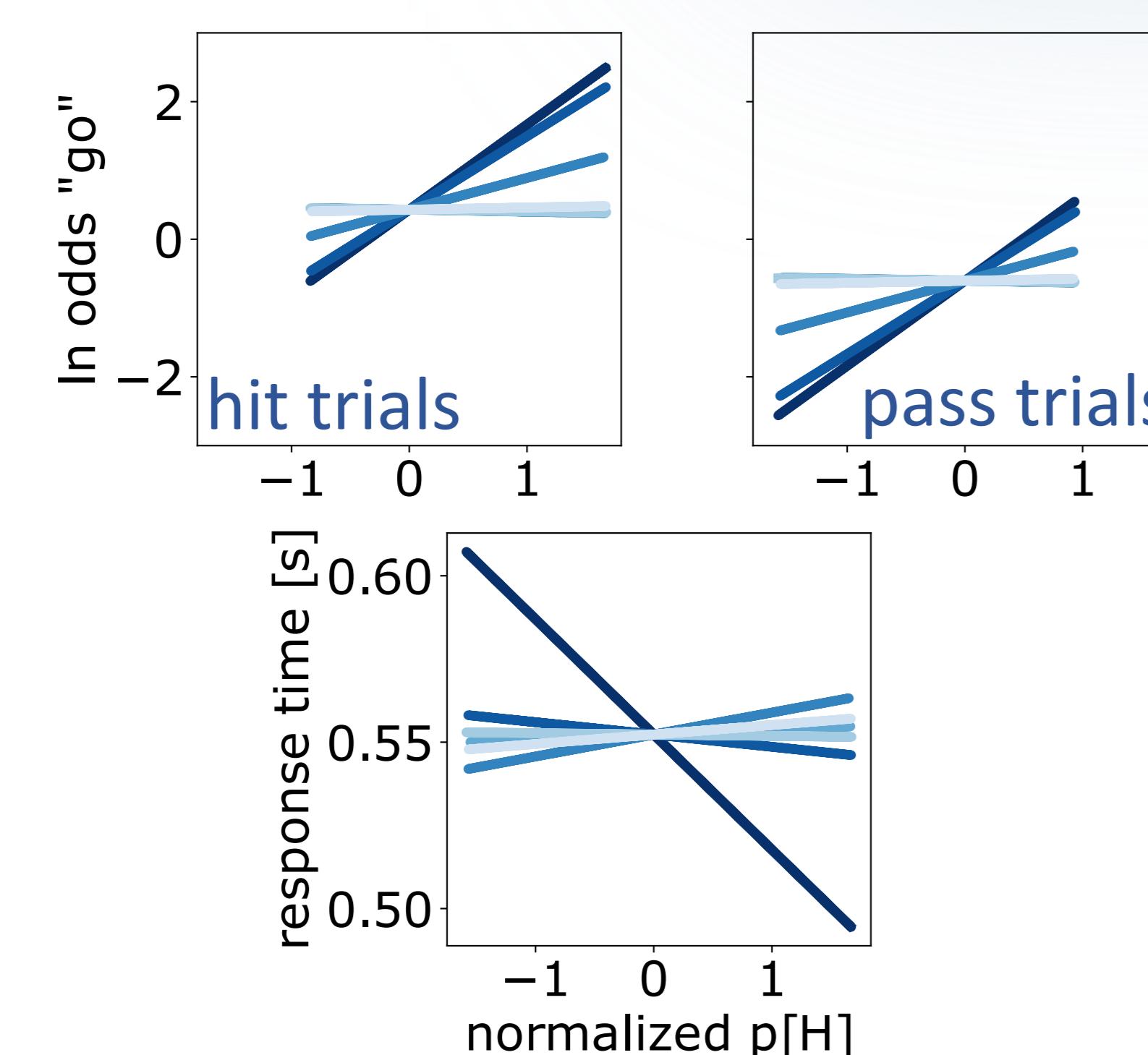
These patterns most closely match the predictions from Go & Adapt model.

... and relative to the go signal suggest that earlier samples can update the response choice, while later samples can't.

The best fitting linear models were:

response choice ~ sample 1 + sample 2 + sample 3 + trial condition + (1|subject)

response time ~ sample 1 + sample 3 + sample 6 + (1|subject)



	Estimate	Z-stat	P-val
Intercept	-0.589	-1.952	0.05
sample 1	1.234	28.261	< 0.01
sample 2	1.067	24.900	< 0.01
sample 3	0.461	11.154	< 0.01
trialCondition	1.006	11.408	< 0.01

	Estimate	DF	T-stat	P-val
Intercept	0.56	3.00	17.23	< 0.01
sample 1	-0.034	4743.14	-23.53	< 0.01
sample 3	0.007	4743.05	5.57	< 0.01
sample 6	0.003	4743.01	2.69	< 0.01

