

Improving the reliability of
Model-based decision-making estimates in
the two-stage decision task with
reaction-times and drift-diffusion modeling

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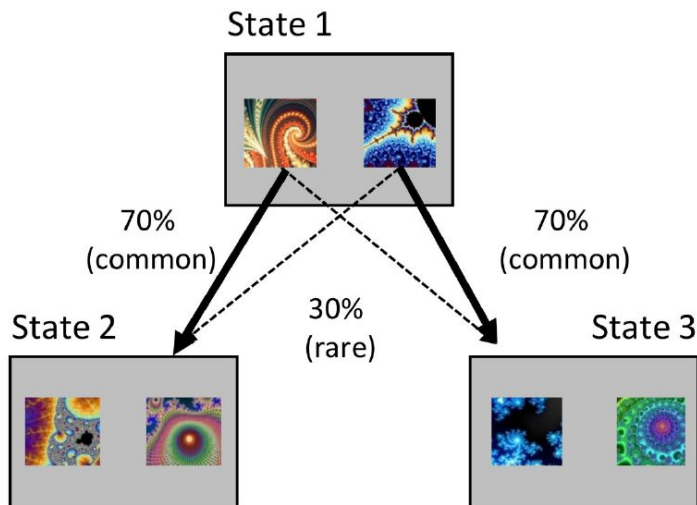
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1. Two-Step Task

- 'Gold standard' for estimating MB, MF contributions to choice
- No task reliability estimates
- Does not use RT

First Stage:

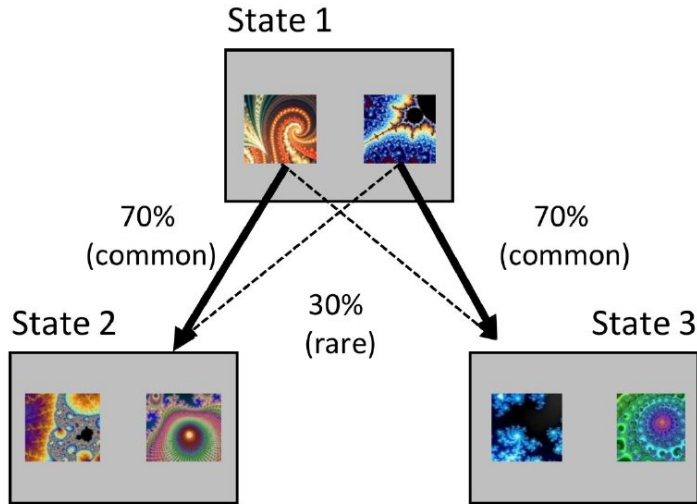


Second Stage:

1. Two-Step Task

First Stage:

Second Stage:



Choice

RT

2. 'Model-Basedness' Estimation

	Process-Based Estimate		Model-Agnostic Estimate	
	RL only	RLDDM	MB-I	MB-II
Data Used	Choice	Choice+RT	Choice	RT
MB Estimate	w	w	RewardXTransition Interaction Effect	Transition Effect

(1) Process-Based Estimates

w-parameter I (RL-choice)

(1) Second stage Q-value update (SARSA-lambda)

$$Q^{\text{MF}}_{(a2,t+1)} = Q^{\text{MF}}_{(a2,t)} + \alpha(r_{(t)} - Q^{\text{MF}}_{(a2,t)})$$

(2) First stage Q-value update (SARSA-lambda)

$$Q^{\text{MF}}_{(a1,t+1)} = Q^{\text{MF}}_{(a1,t)} + \alpha(Q^{\text{MF}}_{(a2,t)} - Q^{\text{MF}}_{(a1,t)}) + \alpha\lambda(r_{(t)} - Q^{\text{MF}}_{(a2,t)})$$

$$Q^{\text{MB}}_{(a1,t)} = P(s_2|a_1) * \max(Q^{\text{MF}}_{(s2,t)}) + P(s_3|a_1) * \max(Q^{\text{MF}}_{(s3,t)})$$

$$Q^{\text{net}}_{(a1,t)} = (1 - w)Q^{\text{MF}}_{(a1,t)} + wQ^{\text{MB}}_{(a1,t)} + p \cdot \text{Stay}_{(a1,t)}$$

w-parameter I (RL-choice)

(3) Choice Rule: Softmax

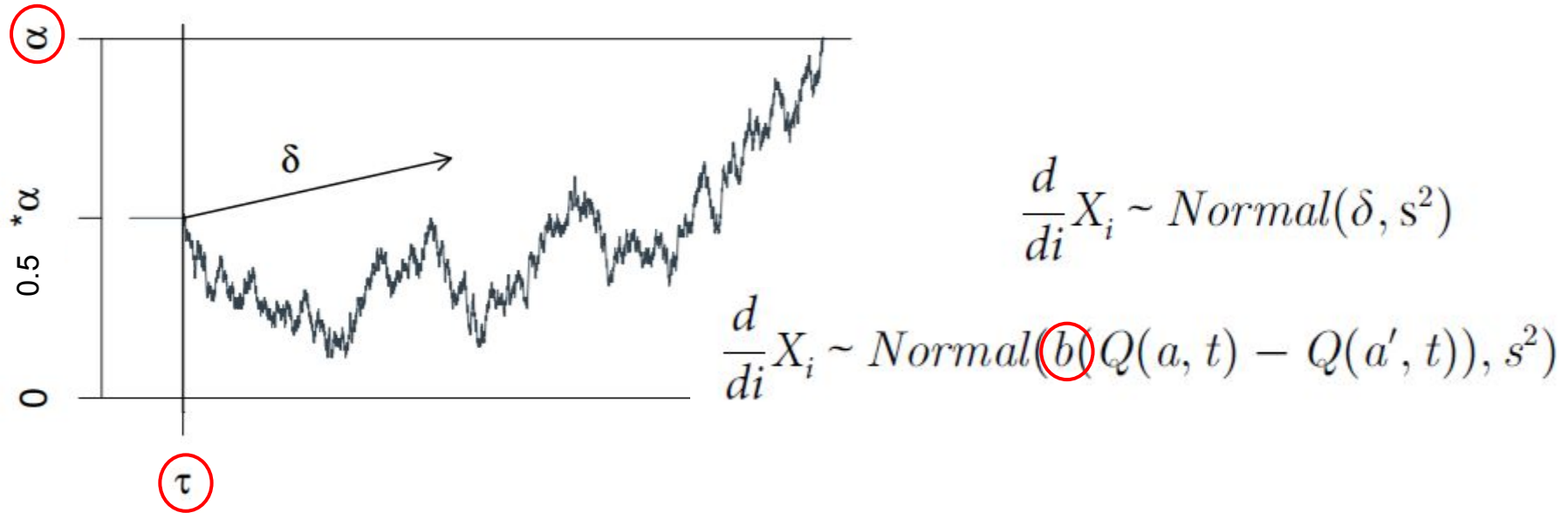
$$P(a1, t) = \frac{\exp(\beta Q^{net}(a1, t))}{\sum_{a'} \exp(\beta Q^{net}(a', t))}$$

$$P(a2, t) = \frac{\exp(\beta Q^{MF}(a2, t))}{\sum_{a'} \exp(\beta Q^{MF}(a', t))}$$

: 5 or 7 parameters

w-parameter II (RLDDM-choice & RT)

(3)' Choice Rule: (Wiener) drift diffusion model

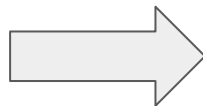
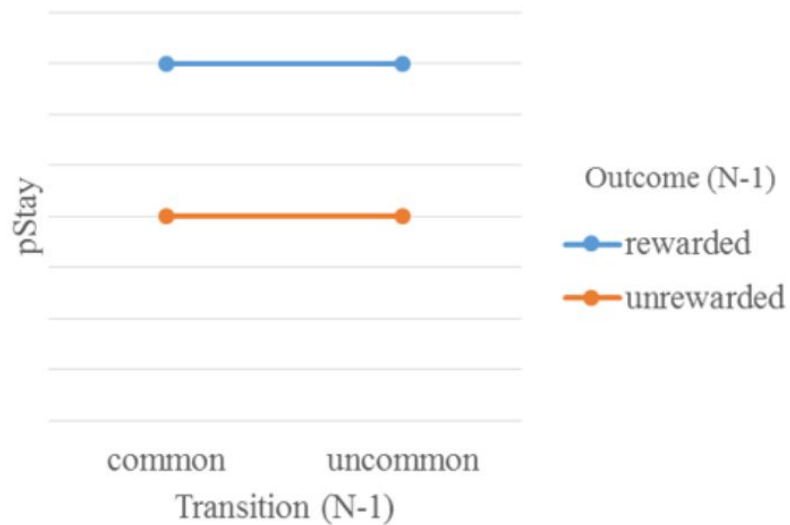


(Image from: Wabersich & Vanderkerckhove (2014))

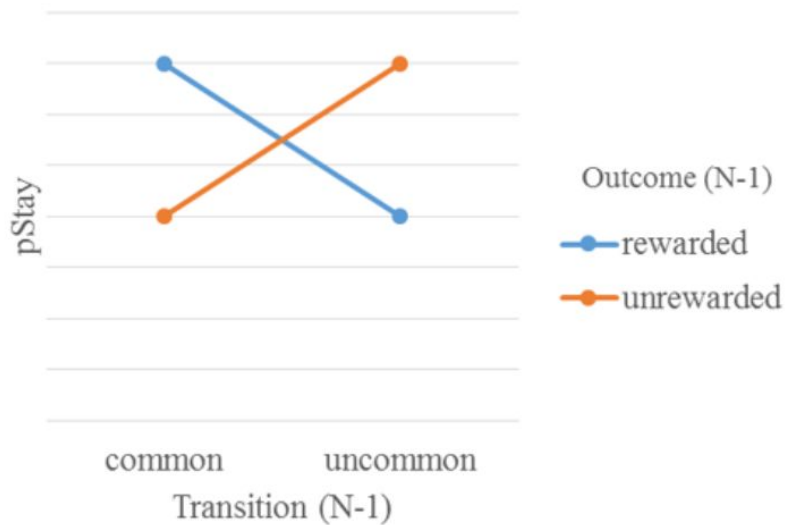
(2) Model-Agnostic Estimates

MB-I (choice)

MF



MB



MB-I (choice)

(1) Individual Estimate

	common transition effect on rewarded trials	common transition effect on unrewarded trials	MB-I
MB agent	$\Delta p_{\text{Stay}} > 0$ ↑	$\Delta p_{\text{Stay}} < 0$ ↑	> 0 ↑
MF agent	$\Delta p_{\text{Stay}} = 0$	$\Delta p_{\text{Stay}} = 0$	$= 0$

$$\text{Rewarded} = P(Y_{\text{stay}(t+1)} | X_{\text{transition}(t)} = 0, X_{\text{reward}(t)} = 1) - P(Y_{\text{stay}(t+1)} | X_{\text{transition}(t)} = 1, X_{\text{reward}(t)} = 1)$$

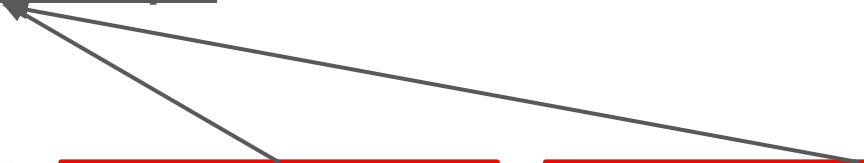
$$\text{Unrewarded} = P(Y_{\text{stay}(t+1)} | X_{\text{transition}(t)} = 0, X_{\text{reward}(t)} = 0) - P(Y_{\text{stay}(t+1)} | X_{\text{transition}(t)} = 1, X_{\text{reward}(t)} = 0)$$

$$\text{MB} - I_{(\text{choice})} = \text{Rewarded} - \text{Unrewarded}$$

MB-I (choice)

(2) Hierarchical estimate

- Mixed-effect logistic regression
- MB-I: 'individual slope' of transition X reward interaction


$$P(Y_{\text{stay}(t+1)}) \sim \boxed{\text{Transition}_{(t)} * \text{Reward}_{(t)}} + \boxed{(\text{Transition}_{(t)} * \text{Reward}_{(t)} | \text{Subject})}$$

Fixed effect
(group prior constraint)

Random effect
(different for each subject)

$$= \text{Transition}(t) + \text{Reward}(t) + \underline{\text{Transition}(t) \cdot \text{Reward}(t)} + (\sim | \text{Subject})$$

MB-II (RT)

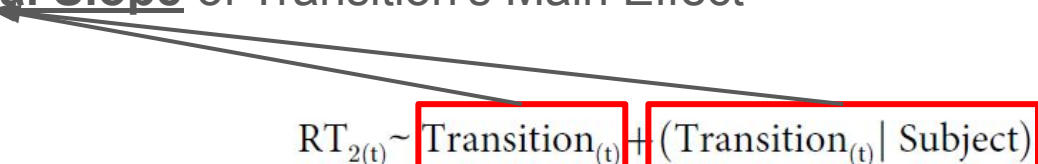
(1) Individual Estimate

- MB-II > 0 if model-based

$$MB - II_{(RT)} = \text{mean}(RT_2 | X_{\text{transition}(t)} = 1) - \text{mean}(RT_2 | X_{\text{transition}(t)} = 0)$$

(2) Hierarchical Estimate

- Individual Slope of Transition's Main Effect

$$RT_{2(t)} \sim \text{Transition}_{(t)} + (\text{Transition}_{(t)} | \text{Subject})$$
A diagram consisting of two arrows. One arrow originates from the boxed term 'Transition_{(t)}' and points to the underlined text 'Individual Slope'. The other arrow originates from the boxed term '(Transition_{(t)} | Subject)' and also points to the underlined text 'Individual Slope'.

3. Data Properties

1. Participants

- Kiddle et al. (2018): NSPN study

	Baseline	Follow-up1	Follow-up2
N	819	63	571
Mean age	18.85 yrs old	18.94 yrs old	20.33 yrs old
Mean time since baseline	0	6.53 months	17.75 months
# of trials in task	121	121	201

3. Data Properties

2. Model-based estimates

(1) **Process-based** fit estimate

- RL model : 7 parameters (separate α and β for first and second stage)
- DDM-RL model : 11 parameters (separate a , b , τ , q for first and second stage)

	RL		DDM-RL	
# of parameters	5	7	8	11
BIC(baseline)	134738.6	134194.7	113648.1	104220.4
BIC(follow-up)	216901.9	215646	174546	162964.3

3. Data Properties

2. Model-based estimates

(2) **Model-agnostic** effect size

- $\eta^2 = SS_{(\text{effect})} / SS_{(\text{Total})}$
- MB-I (choice): $\eta^2 = SS_{(\text{reward} \times \text{transition})} / SS_{(\text{Total})} = .251$
- MB-II (RT): $\eta^2 = SS_{(\text{transition})} / SS_{(\text{Total})} = .673$

4. Results - (1) Relationship between parameters

Table 1. Correlation estimates describing the relationship between the different model-based estimates.

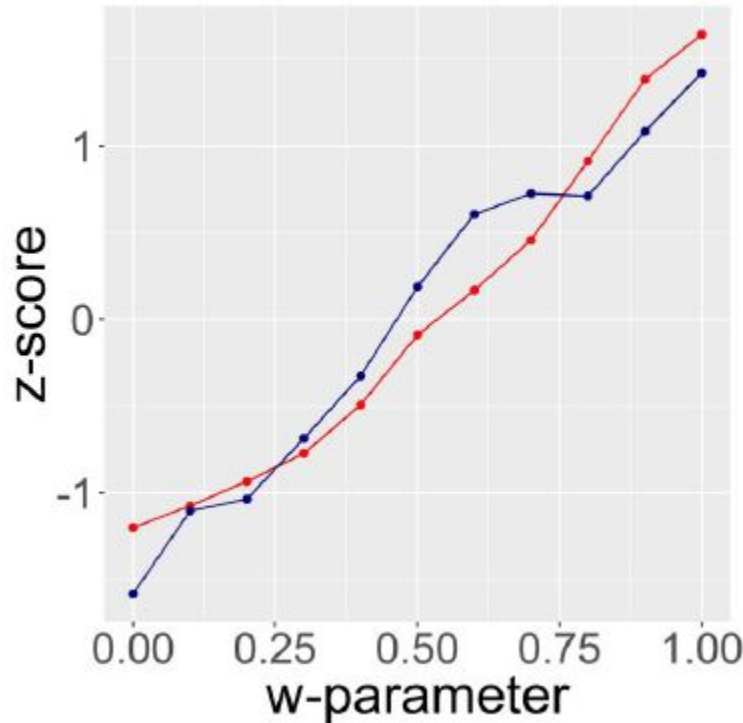
	MB-I (choice)	MB-II (RT)
<u>Individual scores:</u>		
MB-II (RT)	.53 ^a (.47-.59)	.
w-parameter (RL)	.38 (.30-.45)	.26 (.18-.34)
w-parameter (DDM-RL)	.41 (.33-.47)	.24 (.16-.32)
<u>Hierarchical scores:</u>		
MB-II (RT)	.61 ^a (.56-.66)	.
w-parameter (RL)	.31 ^b (.23-.38)	.33 ^b (.26-.41)
w-parameter (DDM-RL)	.37 ^b (.30-.44)	.36 ^b (.29-.43)

Note.

^aPearson correlation estimate.

^bSpearman rank estimate

4. Results - (1) Relationship between parameters



— MB-I (stage 1, choice)
— MB-II (stage 2, RT)

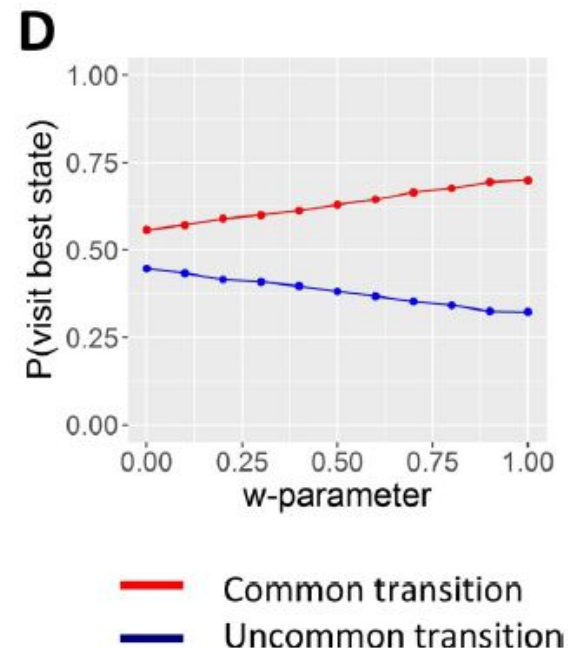
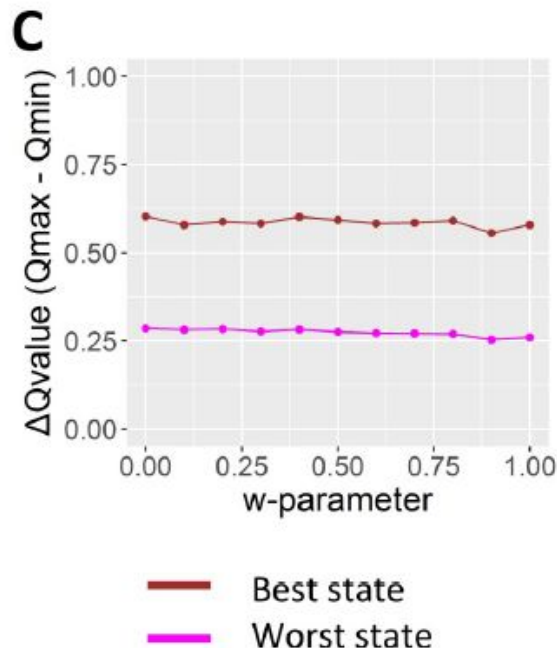
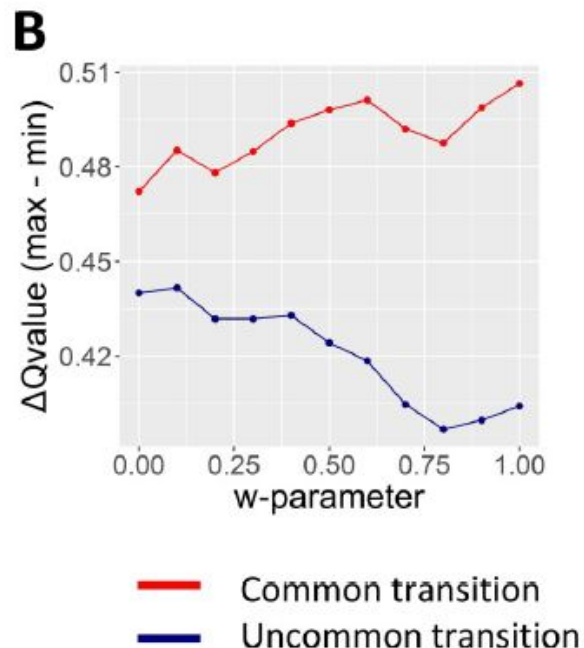
Why?

: Model-based strategy

⇒ Higher discriminability in 2nd stage

⇒ Lower RT

4. Results - (1) Relationship between parameters



4. Results - (2) Parameter Recovery

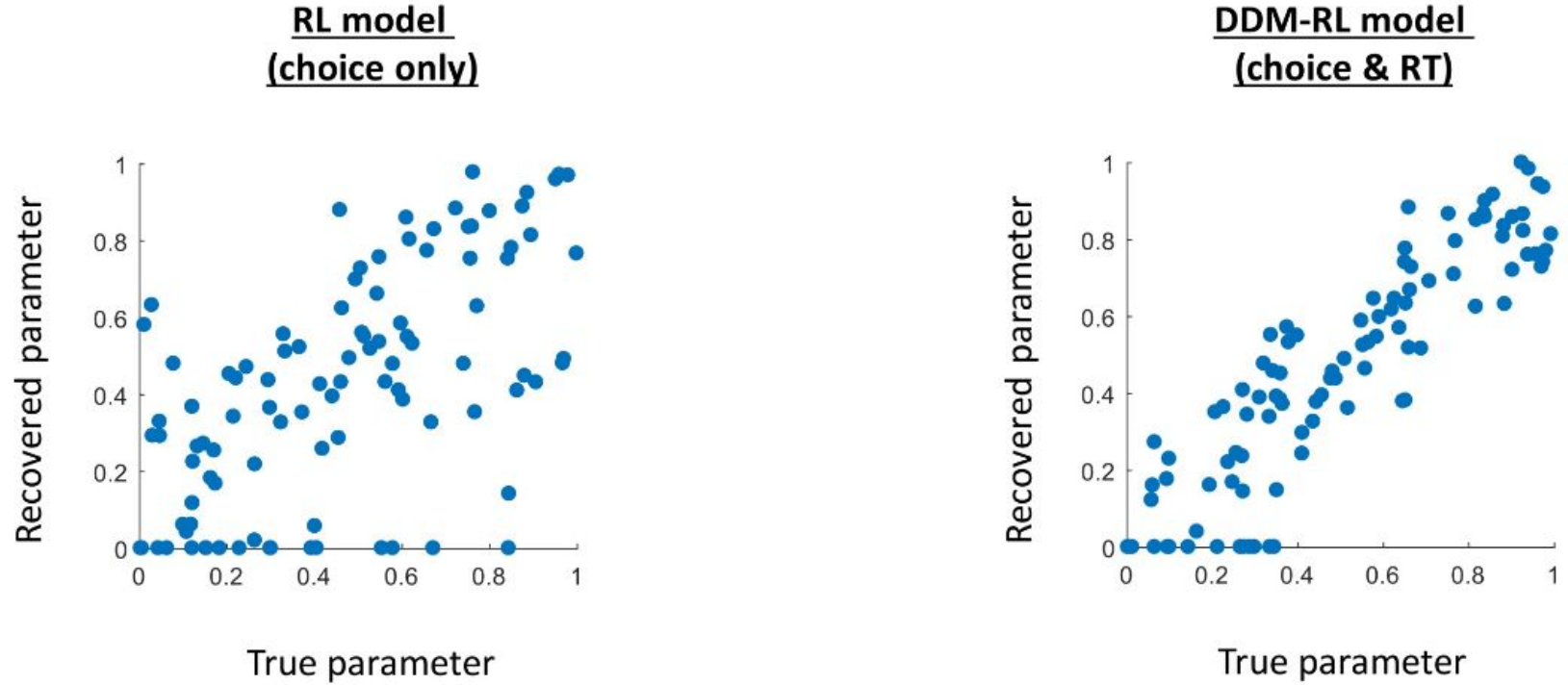


Fig 4. (A/B) Scatter plots for true compared to recovered w -parameter (estimating model-based/free trade off). Results show a better correlation for DDM-RL (panel B; modeling choice & RT, $r = .9$) compared with an RL (choice only) model previously reported in the literature (panel A, $r = .62$).

4. Results - (2) Parameter recovery

Table 2. Spearman's correlation estimating the relationship between the true and recovered parameters.

		Trials in the analysis			
		200	500	1000	5000
RL model (choice)	α_1	.54	.62	.68	.92
	α_2	.95	.98	.99	.99
	λ	.53	.71	.71	.88
	w	.61	.69	.82	.97
	p	.82	.90	.91	.97
	β_1	.82	.90	.93	.98
	β_2	.89	.96	.98	.99
	α_1	.68	.72	.84	.94
DDM-RL model (choice & RT)	α_2	.99	.99	.99	.99
	λ	.58	.75	.83	.92
	w	.90	.95	.96	.99
	p	.91	.94	.97	.99
	b_1	.93	.93	.99	.99
	a_1	.93	.98	.99	.99
	τ_1	.99	.99	.99	.99
	b_2	.99	.99	.99	.99
	a_2	.97	.99	.99	.99
	τ_2	.99	.99	.99	.99

4. Results - (2) Internal consistency (agnostic)

Table 3. Psychometric properties for model-based estimates.

		<u>Internal consistency (201 trials)</u>
MB-I _(choice)	Individual scores	.52 ^c (.45-.58)
	Hierarchical scores	.81 ^c (.78-.84)
MB-II _(RT)	Individual scores	.87 ^c (.85-.90)
	Hierarchical scores	.87 ^c (.85-.89)
Latent score (choice & RT)		.
w-parameter (RL model)	Individual scores	.
	Hierarchical scores	.
w-parameter (DDM-RL model)	Individual scores	.
	Hierarchical scores	.

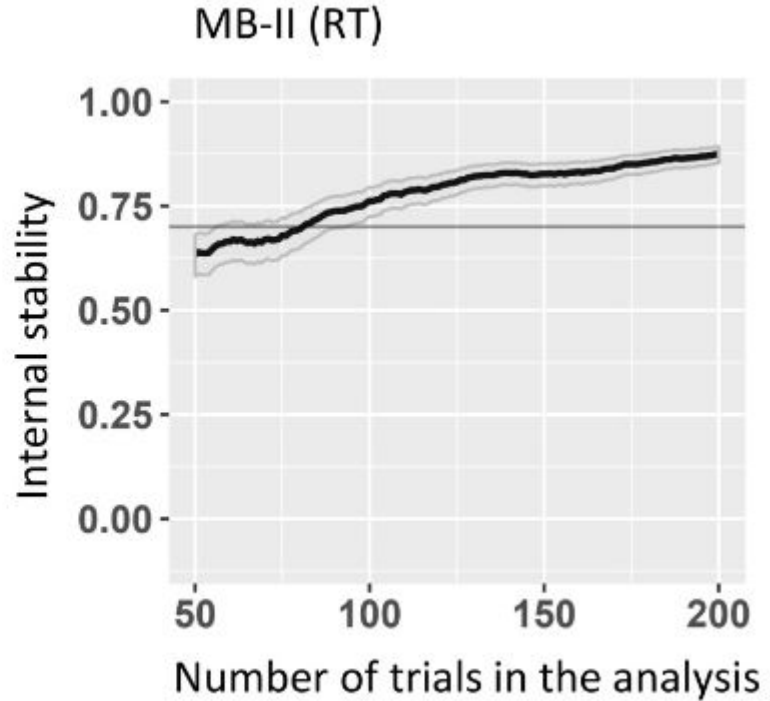
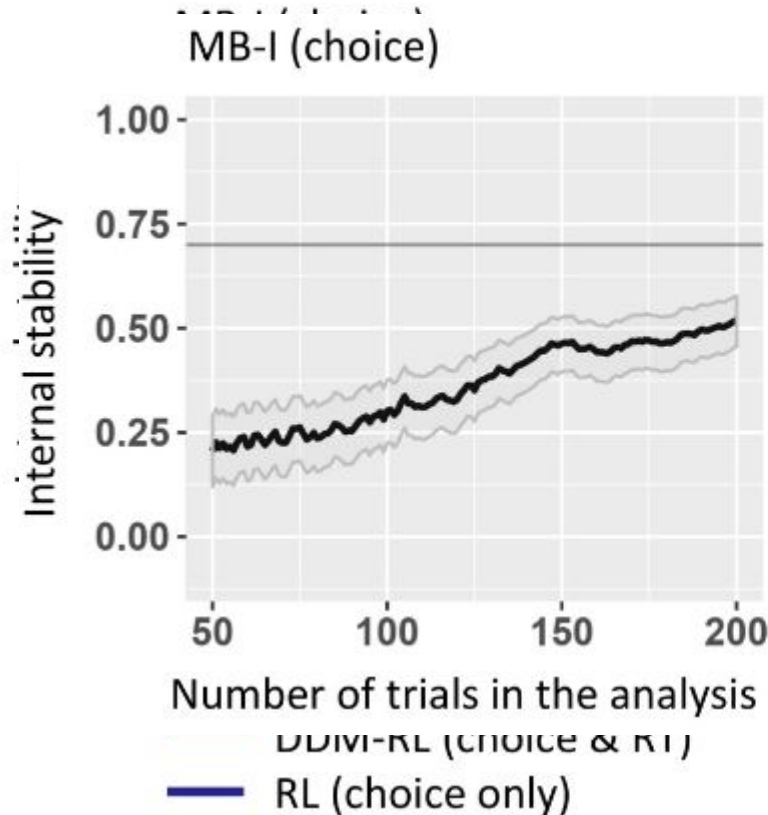
^aPearson correlation estimate.

^bSpearman rank correlation estimate.

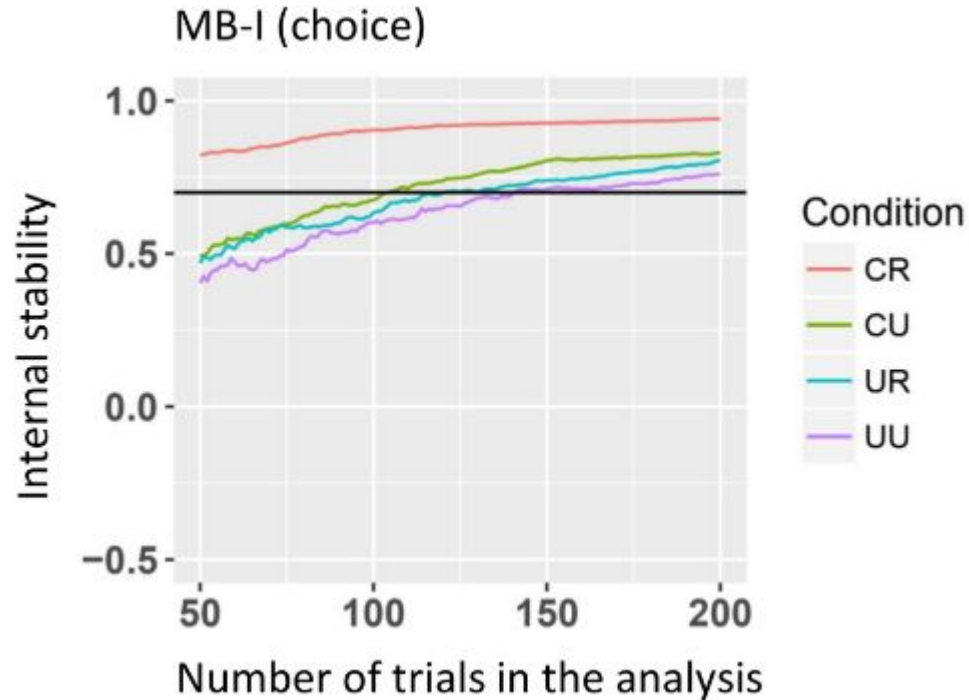
^cSpearman-Brown corrected Pearson correlation estimate.

Estimates in brackets represent 95% confidence intervals.

4. Results - (2) Internal consistency (agnostic)



4. Results - (2) Internal consistency (agnostic)



4. Results - (3) Temporal stability

Table 3. Psychometric properties for model-based estimates.

		<u>Temporal stability</u>
MB-I _(choice)	Individual scores	.28 ^a (.20-.36)
	Hierarchical scores	.40 ^a (.32-.46)
MB-II _(RT)	Individual scores	.33 ^a (.25-.40)
	Hierarchical scores	.33 ^a (.25-.40)
Latent score (choice & RT)		.75 ^a (.71-.78)
w-parameter (RL model)	Individual scores	.16 ^b (.07-.24)
	Hierarchical scores	.21 ^b (.13-.29)
w-parameter (DDM-RL model)	Individual scores	.20 ^b (.12-.28)
	Hierarchical scores	.14 ^b (.05-.22)

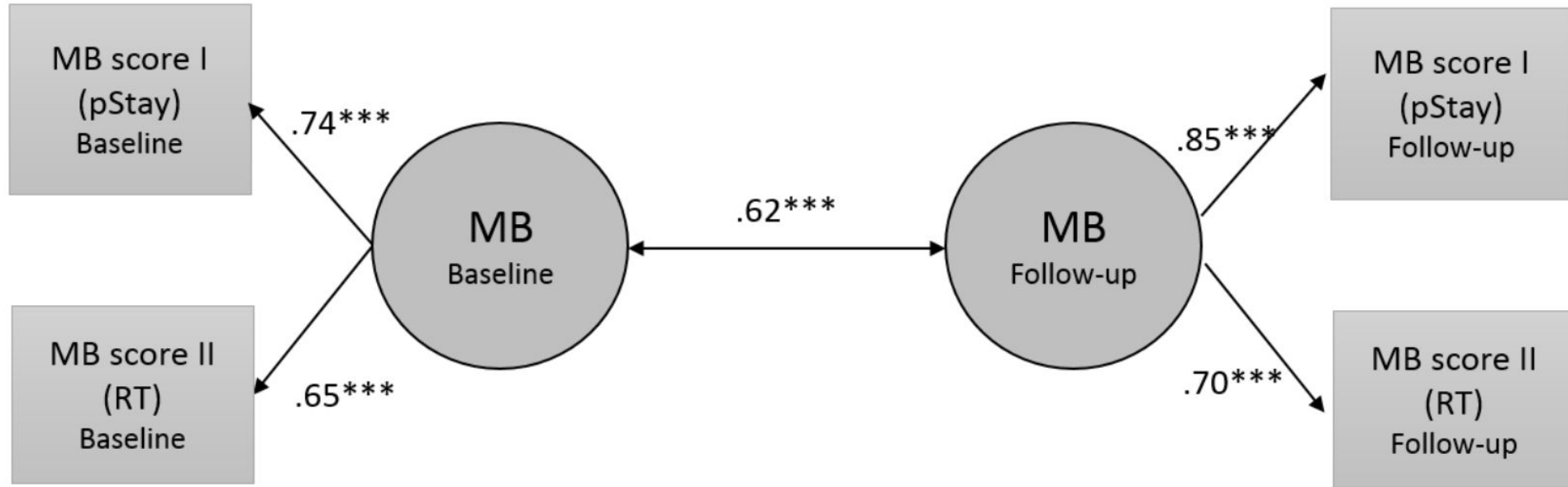
^aPearson correlation estimate.

^bSpearman rank correlation estimate.

^cSpearman-Brown corrected Pearson correlation estimate.

Estimates in brackets represent 95% confidence intervals.

4. Results - (3) Temporal stability



4. Results - (4) Group-level effect

<Power Analysis>

- Power = $P(\text{reject NH} \mid \text{NH is false})$
- effect size, sample size, experiment length \Rightarrow simulate, MANOVA

Table 4. Statistical power (percent of studies that rejected the null hypothesis, given an effect exists) for a between group design (control vs. experiment).
Table values show the chance of finding a statistically significant between group effect as a function of true effect-size, sample-size and number of trials in the experiment.

		<u>200 trials</u>	<u>500 trials</u>	<u>1000 trials</u>
Small effect (Cohen's $d = .2$)	30 participants	4.5%	5.8%	7.5%
	100 participants	6.2%	9%	8.2%
	500 participants	14.1%	24%	28.2%
Medium effect (Cohen's $d = .5$)	30 participants	7%	7%	7.9%
	100 participants	11%	15%	15.1%
	500 participants	52.8%	64%	67%
Large effect (Cohen's $d = .8$)	30 participants	11.5%	10.9%	13%
	100 participants	31.5%	30.3%	37.8%
	500 participants	94.7%	90.7%	96.9%

5. Discussion

1. RLDDM's characteristics

- (1) RLDDM's w -parameter shows close relationships with model-agnostic measures
- (2) Improved parameter recovery, stable model-agnostic estimates
- (3) Provides link between MB-I and MB-II
- (4) Low temporal stability

5. Discussion

2. Other factors affecting stability

(1) Model-agnostic estimate for second-stage RT

(2) Hierarchical estimation

(3) Latent factor analysis

3. What to do?

(1) Use both choice and RT

(2) More trials

Thank you :)