(De)Noise: Moderating the Inconsistency Between Human Decision-Makers

Nina Grgić-Hlača, Junaid Ali, Krishna P. Gummadi and Jennifer Wortman Vaughan





Microsoft Research



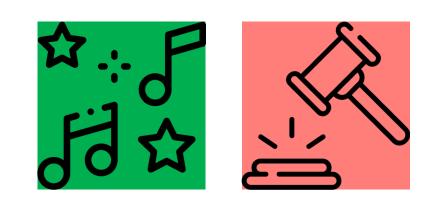














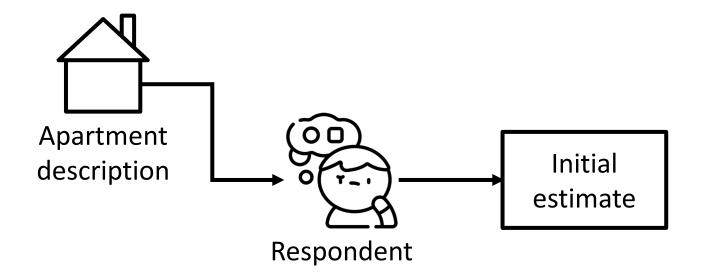


How to reduce inconsistency between decision-makers?

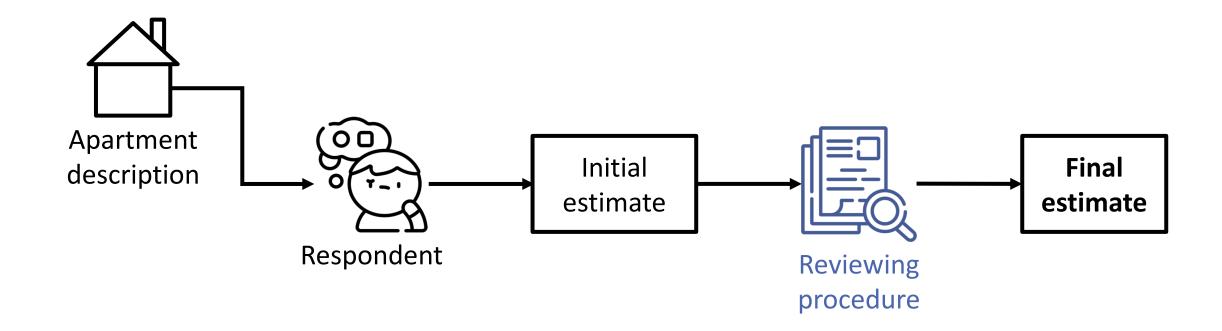
With the help of algorithmic decision aids!

Human-subject experiment: 660 participants recruited via Prolific

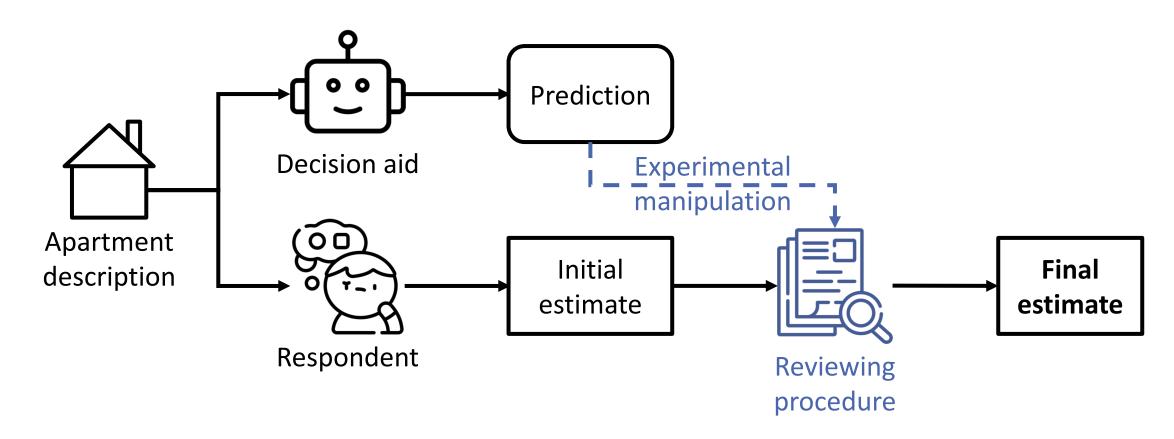
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	Reviewing Procedure	Algorithmic Assistance	Algorithm Predicts
T1	one-by-one	/	/
T2	one-by-one	advice	true price

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T2	one-by-one	advice	true price

For the apartment below, the computer program estimated its price to be: **\$1,800,000**

You initially estimated its price to be: \$1,300,000

Please provide your responses again below. If you wish to change your initial response, please feel free to do so.

Bedrooms	2.0	
Bathrooms	2	
Square footage	1320	
Total rooms	4.5	
Monthly maintenance fee	\$1,330	
Days on the market	80	
Subway distance (miles)	0.168	
School distance (miles)	0.225	
Your initial estimate	\$1,300,000	

Please provide your response below:

What do you think the apartment was sold for?



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T2	one-by-one	advice	true price

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Please provide your response below:

What do you think the apartment was sold for?



Prior work in social psychology: pairwise decisions are easier than absolute [Miller et al.; Stewart et al.]

Absolute identification by relative judgment

Neil Stewart 1, Gordon D A Brown, Nick Chater

Affiliations + expand

PMID: 16262472 DOI: 10.1037/0033-295X.112.4.881

Abstract

In unidimensional absolute identification tasks, participants identify stimuli that vary along a single dimension. Performance is surprisingly poor compared with discrimination of the same stimuli. Existing models assume that identification is achieved using long-term representations of absolute magnitudes. The authors propose an alternative relative judgment model (RJM) in which the elemental perceptual units are representations of the differences between current and previous stimuli. These differences are used, together with the previous feedback, to respond. Without using long-term representations of absolute magnitudes, the RJM accounts for (a) information transmission limits, (b) bowed serial position effects, and (c) sequential effects, where responses are biased toward immediately preceding stimuli but away from more distant stimuli (assimilation and contrast).

The Magical Number Seven, Plus or Minus Two Some Limits on Our Capacity for Processing Information

George A. Miller Harvard University

First, the span of absolute judgment and the span of immediate memory impose severe limitations on the amount of information that we are able to receive, process, and remember. By organizing the stimulus input simultaneously into several dimensions and successively into a sequence of chunks, we manage to break (or at least stretch) this informational bottleneck.

Second, the process of recoding is a very important one in human psychology and deserves much more explicit attention than it has received. In particular, the kind of linguistic recoding that people do seems

	Reviewing Procedure	Algorithmic Assistance	Algorithm Predicts
T1	one-by-one	/	/
T2	one-by-one	advice	true price
T3	in pairs	/	/

Please provide your responses again below. If you wish to change your initial response, please feel free to do so.

	Apartment A	Apartment B
Bedrooms	1.0	2.0
Bathrooms	1	2
Square footage	1240	1160
Total rooms	3	4
Monthly maintenance fee	\$1,170	\$1,330
Days on the market	119	71
Subway distance (miles)	0.149	0.026
School distance (miles)	0.323	0.323
Your initial estimate	\$1,200,000	\$1,200,000

Please provide your response below:

What do you think apartment A was sold for?



Please provide your response below:

What do you think apartment B was sold for?



Please provide your responses again below. If you wish to change your initial response, please feel free to do so.

	Reviewing Procedure	Algorithmic Assistance	Algorithm Predicts
T1	one-by-one	/	/
T2	one-by-one	advice	true price
T3	in pairs	/	/
T4	in pairs	pair selection	people's pairwise comparisons

	Apartment A	Apartment B
Bedrooms	1.0	2.0
Bathrooms	1	2
Square footage	1240	1160
Total rooms	3	4
Monthly maintenance fee	\$1,170	\$1,330
Days on the market	119	71
Subway distance (miles)	0.149	0.026
School distance (miles)	0.323	0.323
Your initial estimate	\$1,200,000	\$1,200,000

Please provide your response below:

What do you think apartment A was sold for?



Please provide your response below:

What do you think apartment B was sold for?



Poviowing

Algorithmic

Algorithm

For the two apartments shown below, our computer program estimated that **Apartment A is <u>less expensive than</u> Apartment B**.

However, you estimated that **Apartment A is <u>equally expensive as</u> Apartment B** .

Please provide your responses again below. If you wish to change your initial response, please feel free to do so.

	Procedure	Algorithmic Assistance	Algorithm Predicts
T1	one-by-one	/	/
T2	one-by-one	advice	true price
T3	in pairs	/	/
T4	in pairs	pair selection	people's pairwise comparisons
\$>\$ \frac{1}{3}	in pairs	pair selection + advice	people's pairwise comparisons

	Apartment A	Apartment B
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Your initial estimate	\$1,200,000	\$1,200,000

Please provide your response below:

What do you think apartment A was sold for?



Please provide your response below:

What do you think apartment B was sold for?





Manual reviewing

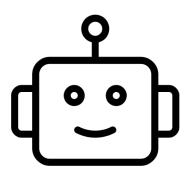
(T1, T3)

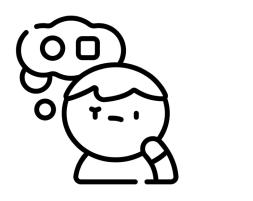


VS

Reviewing with algorithmic assistance

(T2, T4, T5)

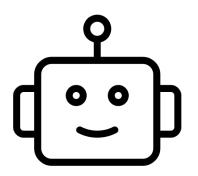




VS

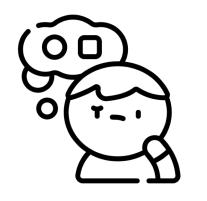
Reviewing with algorithmic assistance

(T2, T4, T5)



Which reviewing procedure will lead to

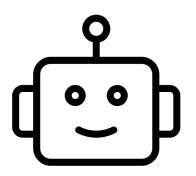
• H1: More updates in initial estimates?



VS

Reviewing with algorithmic assistance

(T2, T4, T5)



Which reviewing procedure will lead to

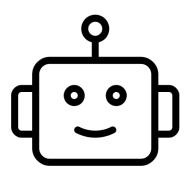
- H1: More updates in initial estimates?
- H2: More accurate final estimates?



VS

Reviewing with algorithmic assistance

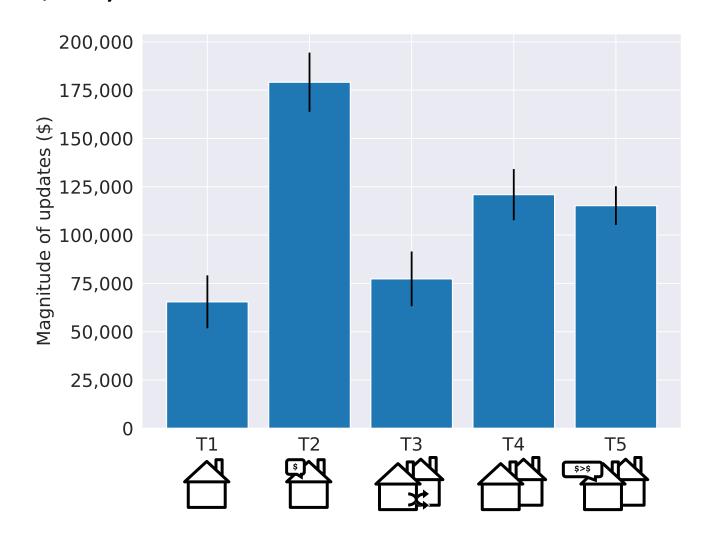
(T2, T4, T5)



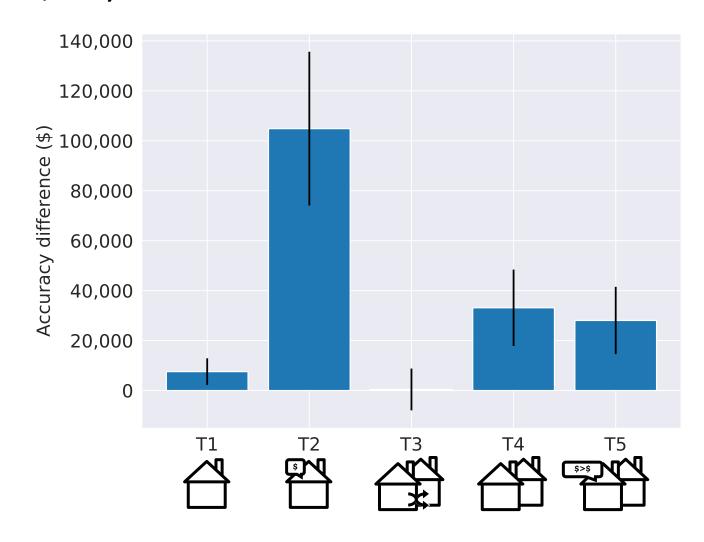
Which reviewing procedure will lead to

- H1: More updates in initial estimates?
- H2: More accurate final estimates?
- H3: More consistency between participants' estimates?

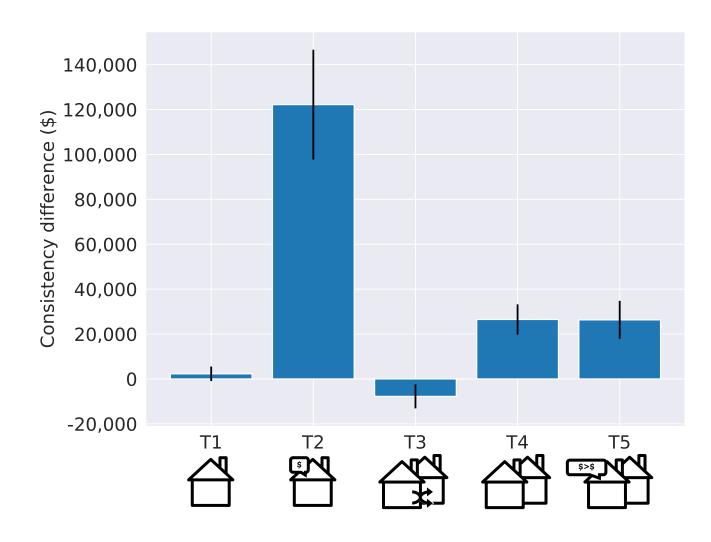
 H1: A larger magnitude of updates of estimates



- H1: A larger magnitude of updates of estimates
- H2: a larger increase in accuracy of estimates



- H1: A larger magnitude of updates of estimates
- H2: a larger increase in accuracy of estimates
- H3: a larger increase in consistency with others' estimates

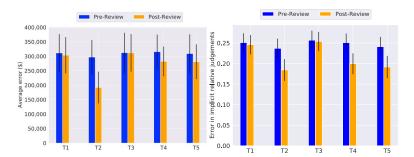


- H1: A larger magnitude of updates of estimates
- H2: a larger increase in accuracy of estimates
- H3: a larger increase in consistency with others' estimates

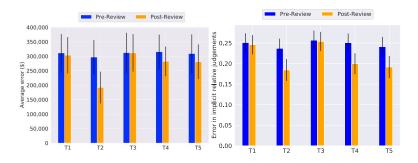
- H1: A larger magnitude of updates of estimates
- H2: a larger increase in accuracy of estimates
- H3: a larger increase in consistency with others' estimates

Are these findings robust?

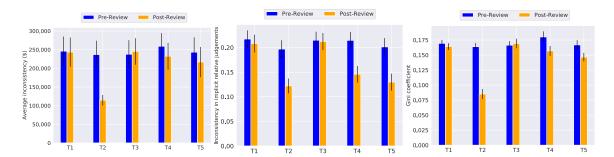
Accuracy



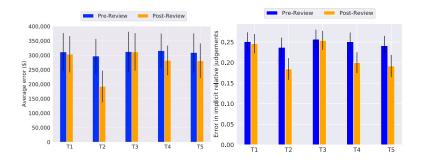
Accuracy



Between-respondent consistency



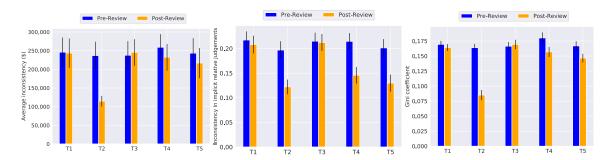
Accuracy



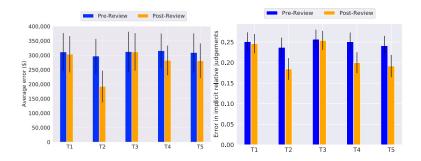
Future work:

- Robustness
 - Decision-making tasks
 - Respondent samples
 - Types of algorithmic assistance

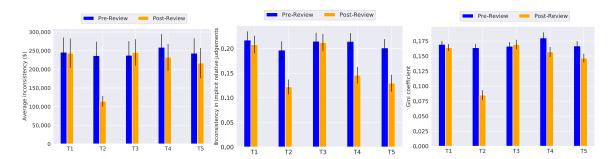
Between-respondent consistency



Accuracy



Between-respondent consistency



Future work:

- Robustness
 - Decision-making tasks
 - Respondent samples
 - Types of algorithmic assistance
- Different notions of consistency
 - Consistency of an individual decision-maker across time & inputs

Algorithmic decision aids increase respondents' accuracy and consistency

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With access to ground truth data

• T2 is most effective



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Without access to ground truth data

• T4 and T5 are still applicable and effective!



Algorithmic decision aids increase respondents' accuracy and consistency

With access to ground truth data

T2 is most effective



• T4 and T5 are still applicable and effective!



T4 is still applicable and effective!







Algorithmic decision aids increase respondents' accuracy and consistency

With access to ground truth data

T2 is most effective



Without access to ground truth data

• T4 and T5 are still applicable and effective!



Without providing explicit advice

T4 is still applicable and effective!



