

# An Experiment on Interpersonal Projection Bias

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# Motivation

- Forecasting others' preferences widely relevant (e.g., auctions, negotiations, coordination, managing workers, social learning, etc.)
  - **Question:** How well do we predict others' preferences?
  - Real-effort setting where people have transparently different preferences
- **Finding:** people project own preference state (fatigue) onto others when predicting their willingness to work
  1. Subjects have familiarity with various states (advances psych lit on social projection; e.g., false-consensus effect)
  2. Measure intra-personal projection in same domain (builds on Conlin et al 2007; Busse et al 2015; Augenblick & Rabin 2019)

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# Real-Effort Task

- All participants do multiple rounds of counting task



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Symbol to count: ?

How many "?" are in the picture?

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# Overview of Experiment

Subjects (MTurk) assigned to one of two roles: **worker** or **predictor**

1. **Workers** do  $s \in \{5, 20\}$  tasks then state willingness to continue working

⇒ Elicit WTW when either Fresh ( $s = 5$ ) or Tired ( $s = 20$ )

2. **Predictors** guesses average WTW in each state

- Induce variation in predictors' preferences:
  - Work on same task as workers
  - Make predictions when Fresh and Tired

⇒ How does predictor's own state influence predictions?

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## Design: Workers

- Two groups of workers ( $N \approx 600$ ):
  1. **Fresh:** Do **5** rounds, then state # additional tasks willing to do for \$ $m$  (BDM)
    - $W(5)$  denotes average response
  2. **Tired:** Do **20** rounds, then state # additional tasks willing to do for \$ $m$  (BDM)
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## Design: Predictors

- Predictors ( $N \approx 670$ ) also work on the task
  - Incentivized predictions about workers
- 3 groups of Predictors

## Design: Predictors

### 1. In-Group: guesses about workers in same state

After 5 rounds:

Predict  $W(5)$

After 20 rounds:

Predict  $W(20)$  and  $W(5)$



### 2. Out-Group A: guesses about workers in opposite state

After 5 rounds:

Predict  $W(20)$

After 20 rounds:

Predict  $W(5)$  and  $W(20)$



### 3. Out-Group B: no guess while fresh

After 5 rounds:

Reminder

After 20 rounds:

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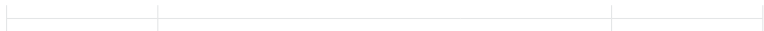
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# Hypotheses

- **Model:** when I'm in state  $s'$ , I think person in state  $s$  has utility function

$$\hat{u}(w|s) = \alpha u(w|s') + (1 - \alpha)u(w|s)$$

- **Hypotheses for  $\alpha > 0$ :**
  1. Same state as target  $\Rightarrow$  unbiased
  2. Less tired than target  $\Rightarrow$  overestimate WTW
  3. More tired than target  $\Rightarrow$  underestimate WTW

## Workers: Lower WTW when Tired

### AVERAGE WILLINGNESS TO WORK

	<i>Workers' State</i>		
	Fresh	Tired	Difference
WTW (Tasks/\$)	10.64 (0.692)	6.81 (0.408)	<b>3.83***</b> (0.803)
Observations	300	299	

## Predictors: Accurate Guesses About Own State

PREDICTIONS OF WTW, SAME STATE (TASKS/\$)

	Prediction	True WTW	Difference
Fresh Predicting Fresh	10.81 (0.605)	10.64 (0.692)	0.17 (0.957)
	$n = 223$	$n = 300$	
Tired Predicting Tired	6.65 (0.230)	6.81 (0.408)	-0.17 (0.439)
	$n = 666$	$n = 299$	

## Predictors: Biased Guesses About Other State

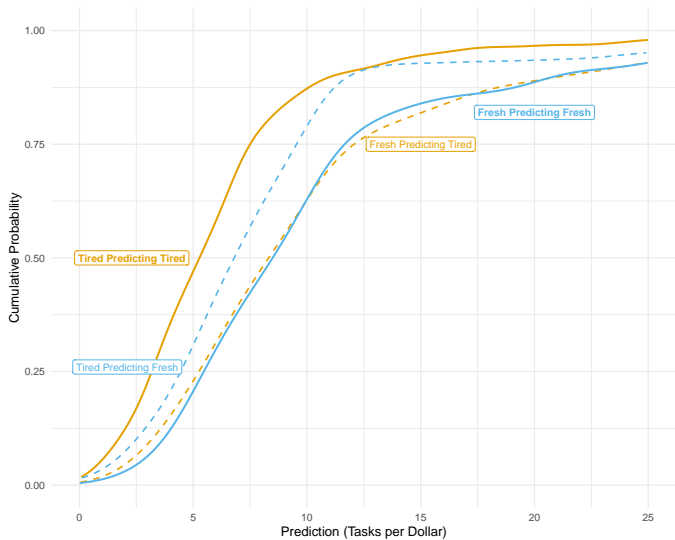
FIRST PREDICTIONS VS WORKERS' WTW (TASKS/\$)

	Prediction	True WTW	Difference
Fresh Predicting Tired	10.22 (0.491)  $n = 221$	6.81 (0.408)  $n = 299$	<b>3.40***</b> (0.639)
Tired Predicting Fresh	8.44 (0.532)  $n = 222$	10.64 (0.692)  $n = 300$	<b>-2.20**</b> (0.873)



## Quantifying Interpersonal Projection Bias

- From previous table:
  1. Fresh predictors overestimate WTW of tired workers by 50%
  2. Tired predictors underestimate WTW of fresh workers by 21%
- From parametric estimation in paper:  $\alpha \geq .23$



## Within-Subject Revisions in Predictions

- Within-subject revisions in guesses:
  - In-Group guessed about fresh workers both when fresh and tired
    - When fresh, guesses were accurate
    - Once tired, guesses reduced by 19%
  - ⇒ Changing this guess significantly reduced expected earnings
- Out-Group A also exhibits significant revisions in guesses about tired workers

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## Intra-Personal Projection Bias

- New group of workers (while fresh) predicted own WTW once tired

PREDICTING WORKERS' GUESSES AND WTW

	Prediction	Actual	Difference
WTW (Tasks/\$)	7.37 (0.393)	5.67 (0.387)	<b>1.70***</b> (0.348)
Observations	298	298	298

*Notes:* Difference significant at  $p < .001$

- Overestimate own WTW by 30%
  - Recall: fresh predictors overestimate WTW of tired workers by 50%
- ⇒ Interpersonal bias more severe than *intrapersonal* bias

# Conclusion

- **Take away: Difficult to empathize with others' states**
  - Even when others' states are known
  - Even when others' states are familiar
- Additional Points:
  - Neither accuracy nor confidence increases over time  $\Rightarrow$  errors in predictions likely not due to limited info
  - Magnitude of *intrapersonal* projection similar to other studies (Conlin et al 2007; Busse et al 2015; Augenblick & Rabin 2019)
- In Progress:
  - How does projection distort learning from others' behavior?
  - How does it influence strategic behavior (e.g., bidding in auction)?

Thanks!