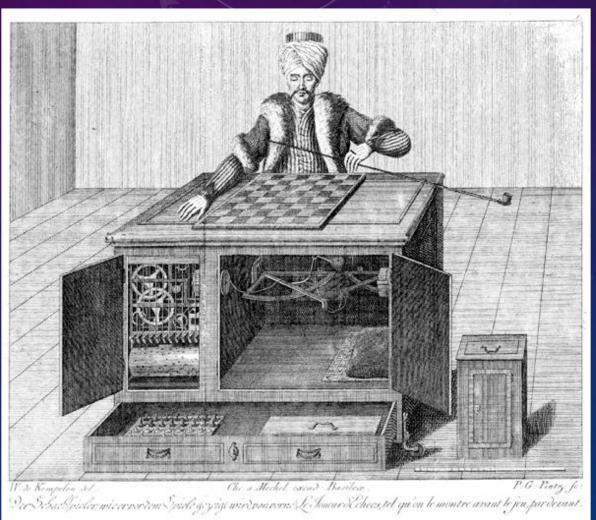


amazonmechanical turk Artificial Artificial Intelligence



Outline

- Background
- Terms
- Demographics
- Workers perspective (incentives)
- Price and time
- Filtering
- Worker-Requester relationship
- Testing Familiar problems
- Special constellations
- Pros and Cons
- Tips
- Future applications
- References

Background

- Created at 2005
- One of many...
- Population
 - Jan, 2007 100,000 from 100 countries
 - Jan, 2011 500,000 from 190 countries
 - Jan, 2015 >1,000,000...?
- Uses
 - Capcha solving
 - Translation / Transcription
 - Picture classification
 - Social science...

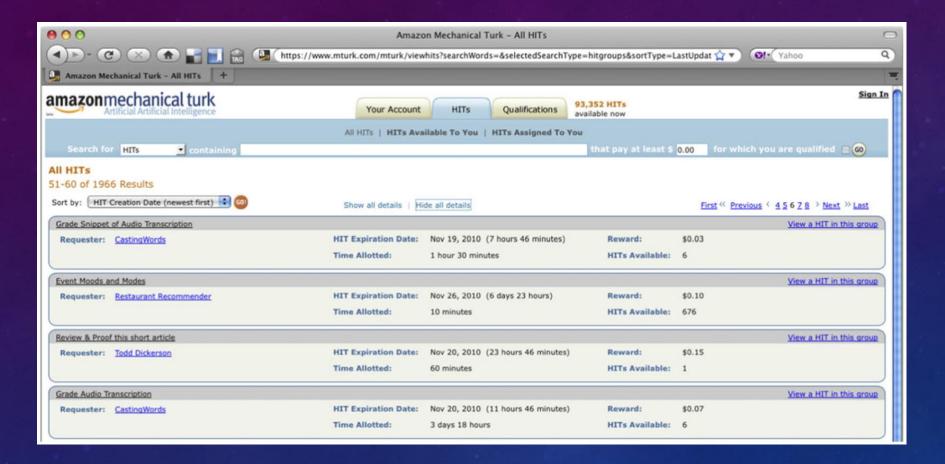




Terms

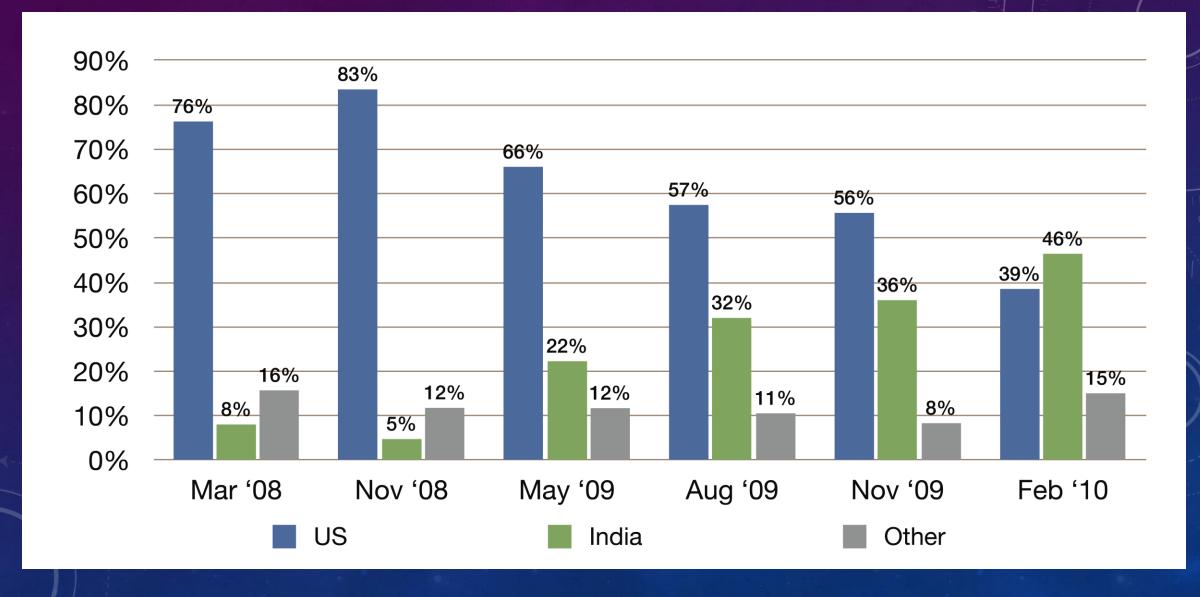
- Requester, Worker
- HIT human intelligence task
- Bonus
- Qualification
- Rejection / Success rate
- ACQ attention check question
- API application programming interface
 - ('artificial artificial intelligence')
- Bot
 - ('artificial artificial artificial intelligence')
- Sweatshops

How does it look for workers



Recency Money vs time Reputation

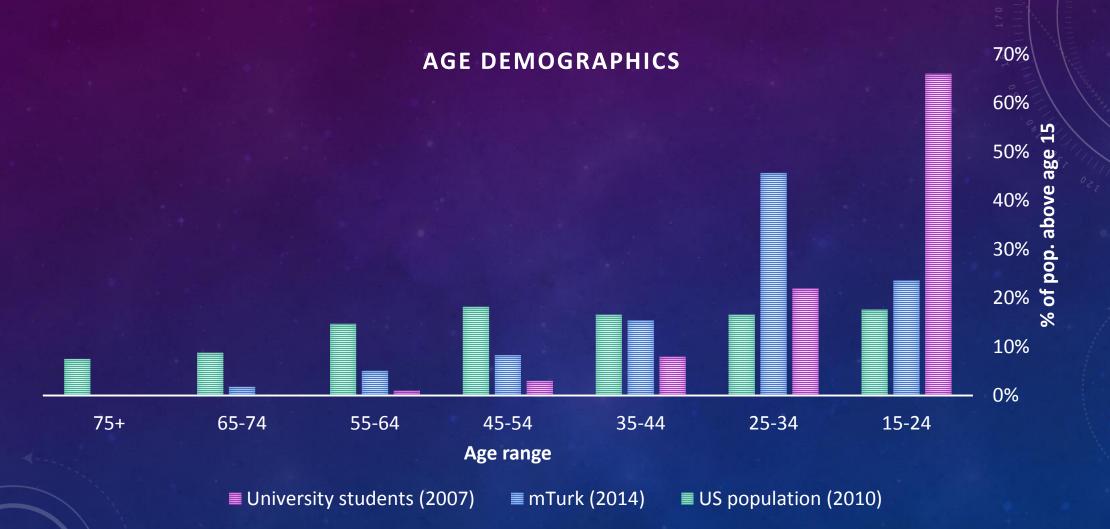
Demographics – workers country



Demographics – age, gender, education, income

		Nov '08	Feb '09	May '09	Aug '09	Nov '09	Feb '10
	Overall	32.9	33.0	32.5	31.0	31.7	30.4
Age (average)		33.6		34.3	33.2	35.4	34.3
11] Buhr	mester, K	Wana	d Gosling	27.6	26.4	27.8
	I In st	ort, MTu	rk name and	d Gosling	(2014)	48% 52%	55% / 45%
Gender (male/female	divers	e than sta	'' ^N Particip	d Gosling Dants were ternet sar America	(2011):		/ 59%
	more (diverse th	an to	ternet sar I America 174%	nnia	emograph	oicau. H
 Education	Overall	47%	чт туріса	l America	ripies and	significat	nth.
(Bachelors or higher)	US	43%		4070	College	Samples "	, 'ciy
or riigher)	India	96%		74%	69%	17.05.	$\longrightarrow \sqcup$
	Overall	10%	22%		27%	32%	39%
Income (<\$10k/yr.)	US	8%		No data	12%	10%	15%
. ,	India	35%			54%	65%	61%

Demographics – age



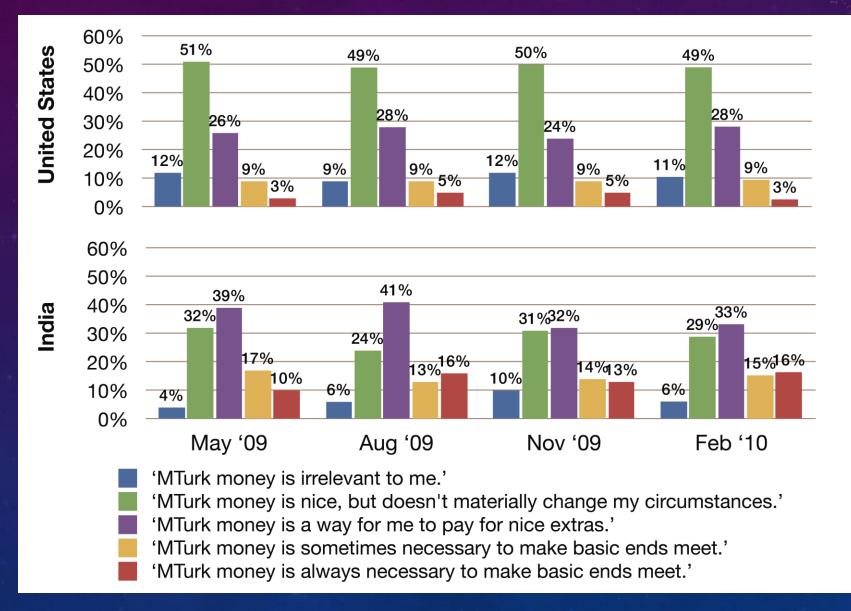
Can we trust demographic data?

- Mason, Suri (2011) Conducting behavioral research on Amazon's Mechanical Turk
 - Only 0.4% (1/207) changed their demographic data
- Rand (2012) The promise of Mechanical Turk: How online labor markets can help theorists run behavioral experiments
 - I.P. address check
 - Country of residence truthfulness 97%
 - Same participants in 2 separate studies (N~100)
 - Gender 96%
 - Age -93% (within 1 year)
 - Country 98%
 - Education level 81%
 - Yearly income 82% (within 1 bracket out of 10)
 - Belief in god 84% (within 1 point out of 10)

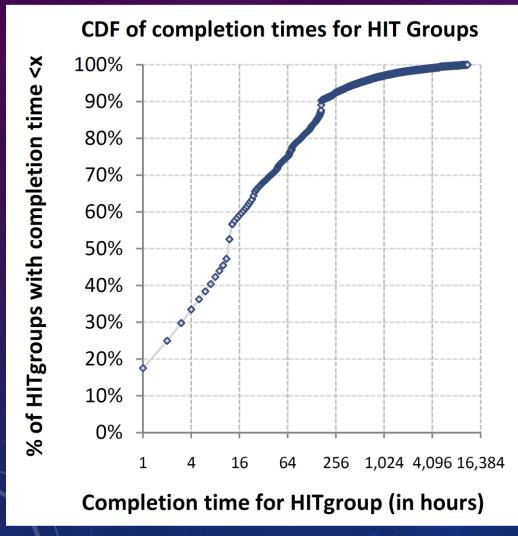
Workers' perspective

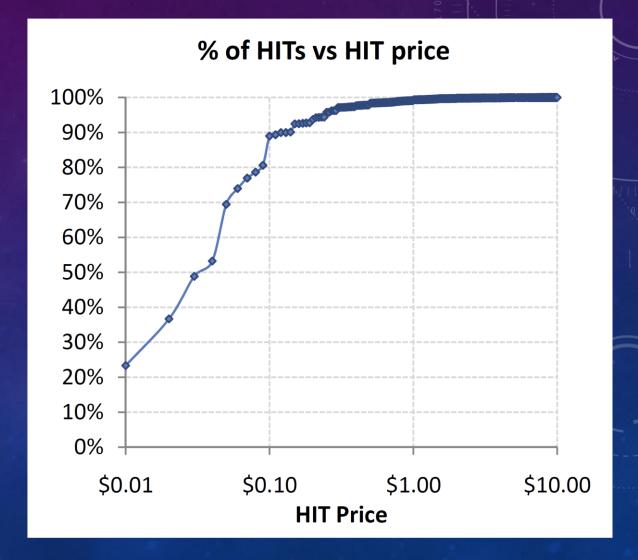
- 59% of Indian workers and 69% of U.S. workers agreed that "Mechanical Turk is a fruitful way to spend free time and get some cash" (Ipeirotis, 2010).
- "Most workers are not motivated primarily by the financial returns and genuinely care about the quality of their work" (Paolacci, Chandler, Ipeirotis, 2010).

Demographics – MTurk as a source of income



Price and Time



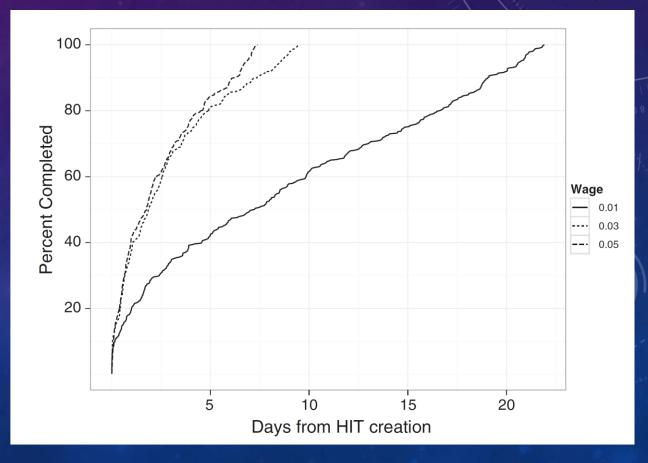


Price and Time Buhrmester, Kwang and Gosling (2011)

HIT completed per hour

Lengh Compensation	5 min	10 min	30 min
2¢	5.6	5.6	5.3
10¢	25.0	14.3	6.3
50¢	40.5	31.6	16.7

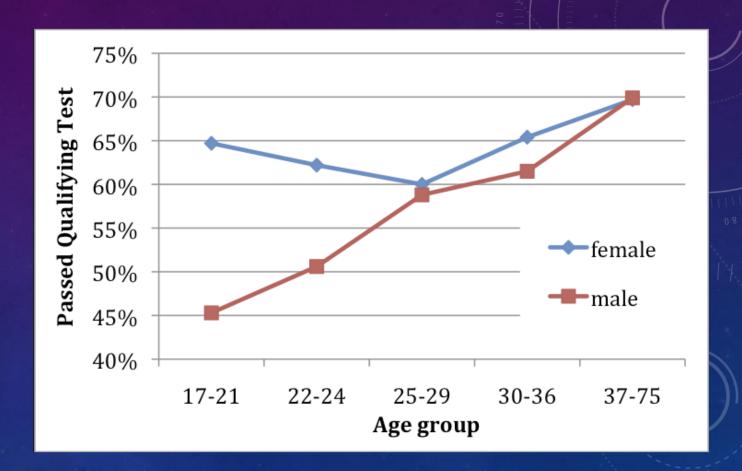
Mason, Suri (2011)



Filtering – un-rejected rate

My experience without prescreening: ~ 50%

With pre-screening: ~ 80-90%



Filtering

- Post
 - ACQ

- Median Invalid Duration duration < 1 minute Comment Responses Exp 1 48.6% 1:30 30.5% Exp 2 2.5% 4:06 6.5%
- Table 1. Improvement in response quality in Experiment 2 upon the introduction of verifiable questions.

- Comprehension question (manipulation check)
- "How would you answer... if..."
- "please select b"
- Time measurement (whole assignment/per page)
- Pre
 - Success rate, min. number of HITs
 - **Country** (multiple choice available now)

Kittur, Chi, Suh (2008)

Filtering

		High Reputation		Low Reputation			
		Passed ACQs	No ACQs	Passed ACQs	Failed ACQs	No ACQs	
N		302	156	117	60	59	
Cronbach's alpha	SDS	.629 _a	.698 _a	.471 _b	.242 _b	.557 _{ab}	
	RSES	.936 _a	.934 _{ad}	.912 _{ad}	.825 _{bc}	.889 _{cd}	
	NFC	.952 _a	.947 _{ad}	.891 _{ad}	$.759_{bc}$.863 _{ed}	
SDS mean percent (SD)		44.87 (21.5)	45.71 (23.7)	48.63 (18.9)	53.0 (17.3)	49.83 (21.2)	
Anchoring effect size (r)		.198 _a *	.183 _a *	.280a*	046_{b}	.049 _b	
Average percent of midpoint marked on scale items (SD)		19.28 (14.1)	20.78 (14.06)	25.12 (17.04)	34.21 (26.56)	27.61 (21.08)	

- Peer, Vosgerau, Acquisti (working paper) Reputation as a sufficient condition for data quality on Amazon Mechanical Turk
 - ACQ increase data quality
 - But only for low reputation workers...
 - No significant effect in high reputation (>95%, >1000 HIT)
 - Conclusion
 - Use only high reputation workers (not 'master workers')

Reputation	Passed (All ACQs)	Failed (at Least One ACQ)	iled (at Least One ACQ) # of ACQs Fa		Failed	
			1	2	3	
High	294 (97.4 %)	8 (2.6 %)	8 (2.6 %)	0	0	
Low	117 (66.1 %)	60 (33.9 %)	35 (19.8 %)	14 (7.9 %)	11 (6.2 %)	

Testing – Behavioral decision making

- An Assessment of Experiments run on Amazon's Mechanical Turk, Wolfson & Bartkus, 2013
 - Endowment Effect
 - Kahneman, Knetsch and Thaler (1986) car dealer, wages



- Prospect Theory
 - Kahneman and Tversky (1979) risk averse risk seeking



- Anchoring
 - Tversky and Kahneman (1974) wine price* (social sec. num.)
 - Stewart (2009) credit card payment (minimum payment)



Testing – Behavioral decision making

	Mechanical Turk	Midwestern university	Internet boards
Asian Disease			
% Risky Positive Frame	17.6%	28.1%	23.7%
% Risky Negative Frame	55.3%	67.7%	63.0%
χ^2	10.833	20.230	13.013
p	< 0.001	< 0.001	< 0.001
Effect size (w)	0.39	0.39	0.39
Linda problem			
% Conjunction Fallacy	72.2%	78.3%	64.4%
Physician problem			
Avg. Quality Success (SD)	5.93 (0.81)	5.63 (0.75)	5.73 (0.98)
Avg. Quality Failure (SD)	5.13 (1.24)	4.86 (1.29)	4.93 (1.41)
t	3.70	4.14	2.547
p	< 0.001	< 0.001	0.007
Effect size (d)	0.76	0.73	0.66

Paolacci, Chandler, Ipeirotis (2010)

Testing – Cognitive psychology

- Crump, McDonnell, Gureckis (2013) Evaluating Amazon's Mechanical Turk as a **Tool for Experimental Behavioral Research**
 - Reaction time
 - Stroop

"All of the reaction time tasks chosen for validation purposes were replicated. In addition, error rates were low overall suggesting that participants took the task seriously... Overall,

"In conclusion, AMT is a promising development for experimental cognitive science research. On balance, our investigations suggest that the data quality is reasonably high and compares well to laboratory studies... If we as scientists mes." respect the participants and contribute to a positive experience on AMT it could turn into an invaluable tool for accelerating empirical research."

t used relatively constraints for

- Subliminal priming
- Learning
 - Category learning

"online participants generally learned more slowly [compared to lab - se magnitude of payment does not have a strong effect on the quality of data obtained from online, crowd-sourced systems... building in checks for understanding the instructions is critical for ensuring high quality data."

Pros and Cons

- Cons
 - Reliability?
 - Only internet
 - Random attribution (dropout)
- Pros
 - Simple
 - Fast
 - Cheap (by some studies in the US by a factor of 6 compared to students in lab)
 - More external validity than using students
 - Diversity (age, income...)
 - International (Culture studies)
 - Big pool (>1,000,000?)

Unresolved issues

- Personality (what kind or a person is a Turker?)
- Workload (ego depletion)
- Speciality (familiarity with known manipulations)
- Environment (Alone vs. in a lab: pro or con?)
- Trust (Turker gambles, Requester details)

Special constellations

Longitudinal studies:

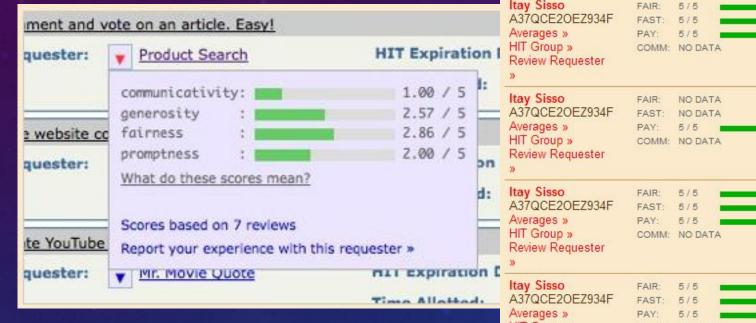
- Holden, Dennie and Hicks (2013)
 - M5-120 personality scale (120 items)
 - 3 weeks test-retest: 280 1st wave, 67 2nd wave (46 relevant)
 - 0.1\$ 1st, 0.15\$ 2nd (41% india)
- Buhrmester, Kwang and Gosling (2011)
 - Big-5 personality scale (44 items)
 - 3 weeks test-retest: 60% 2nd wave
 - 0.2\$ 1st, 0.5\$ 2nd

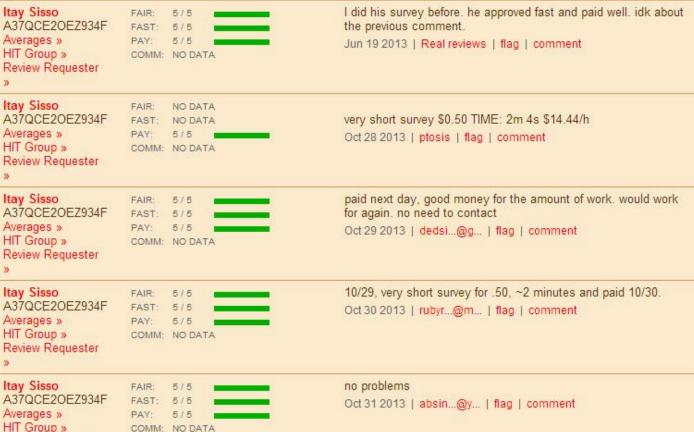
Interaction tasks:

- Mason, Suri (2012) Conducting behavioral research on Amazon's Mechanical Turk
- Suri, Watts (2011) Cooperation and Contagion in Web-Based, Networked Public Goods Experiments
- Pseudo-dyadic Summerville and Chartier (2012)

Tips - Requestres reviews

Turkopticon (http://microwork-dev.ucsd.edu/)

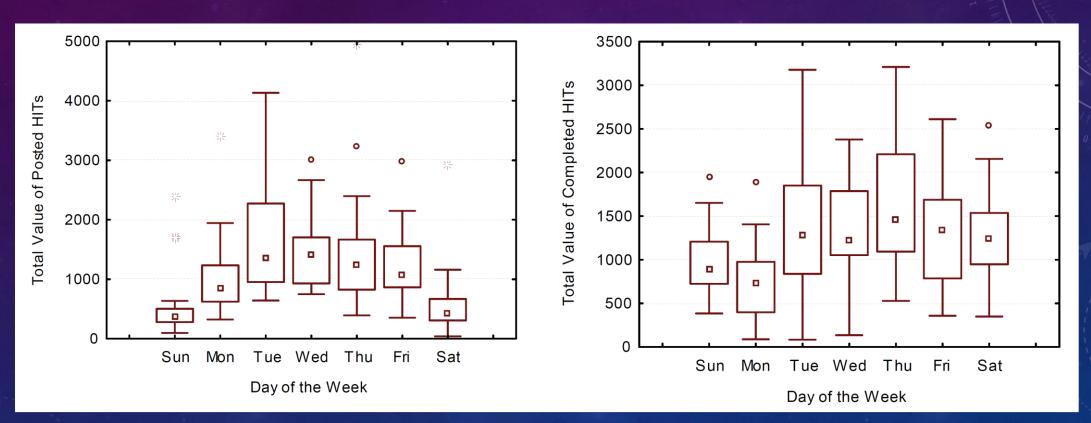




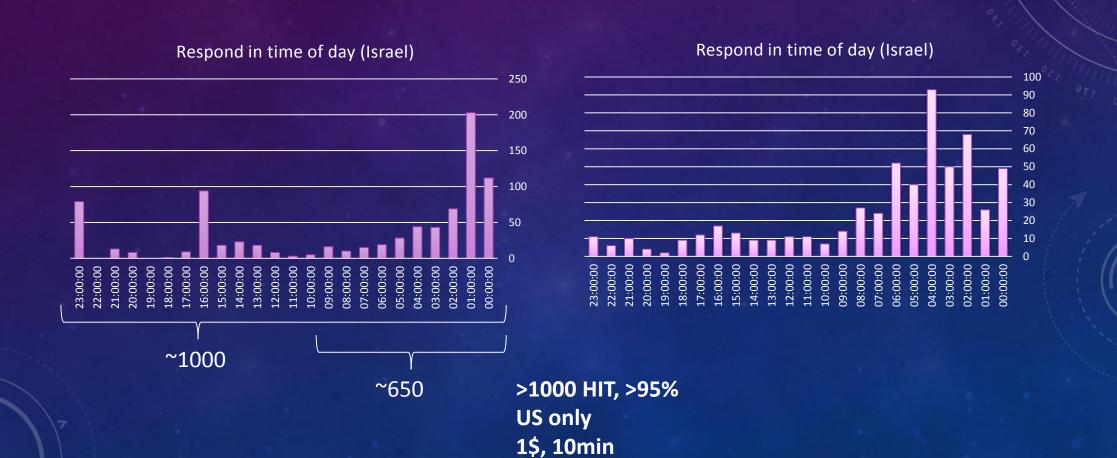
COMM: NO DATA

Review Requester

Tips – when to post?



Tips – When to post?



Other tips

- Filtering by MturkID (Qualification / script+Qualtrics)
- Differential payement (Bonus)
- Test Retest (e-mail)