# Microtask Crowdsourcing: Fundamentals (Part II)

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# **Quality Control**

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- Extremely important part of the experiment
- Approach as "overall" quality; not just for workers
- Bi-directional channel
  - You may think the worker is doing a bad job.
  - The same worker may think you are a lousy requester.
  - Do check the worker forums!

CrowdFlower has an option to directly chat with a live worker!

# **Quality Control**

- Approval rate: easy to use, & just as easily defeated
- Mechanical Turk Masters (since June 2011)
  - Only for specific tasks
- Qualification test
  - Pre-screen workers' ability to do the task (accurately)
- Assess worker quality as you go
  - Trap questions with known answers ("honey pots")
  - Measure inter-annotator agreement between workers

# Qualification tests: pros and cons

- Advantages
  - Great tool for controlling quality
  - Adjust passing grade
- Disadvantages
  - Extra cost to design and implement the test
  - May turn off workers, hurt completion time
  - Refresh the test on a regular basis
  - Hard to verify subjective tasks like judging relevance
- Try creating task-related questions to get worker familiar with task before starting task in earnest

# Methods for measuring agreement

- What to look for
  - Agreement, reliability, validity
- Inter-agreement level
  - Agreement between judges
  - Agreement between judges and the gold set
- Some statistics
  - Percentage agreement
  - Cohen's kappa (2 raters)
  - Fleiss' kappa (any number of raters)
- With majority vote, what if 2 say relevant, 3 say not?
  - Use expert to break ties
  - Collect more judgments as needed to reduce uncertainty

# Quality Control & Assurance

#### Filtering

- Approval rate (built-in but defeatable)
- Geographic restrictions (e.g. US only, built-in)
- Worker blocking
- Qualification test
  - Con: slows down experiment, difficult to "test" relevance
  - Solution: create questions to let user get familiar before the assessment
- Does not guarantee success
- Identify workers that always disagree with the majority
- Ask workers to rate the difficulty of a task

# Other quality heuristics

- Justification/feedback as quasi-captcha
  - Should be optional
  - Automatically verifying feedback was written by a person may be difficult (classic spam detection task)
- Broken URL/incorrect object
  - Leave an outlier in the data set
  - Workers will tell you
  - If somebody answers "excellent" for a broken URL => probably spammer

# Dealing with bad workers

- Pay for "bad" work instead of rejecting it?
  - Pro: preserve reputation, admit if poor design at fault
  - Con: promote fraud, undermine approval rating system
- Use bonus as incentive
  - Pay the minimum \$0.01 and \$0.01 for bonus
  - Better than rejecting a \$0.02 task
- If spammer "caught", block from future tasks
  - May be easier to always pay, then block as needed

# Answer justification

- Why settle for a label?
- Let workers justify answers
- Has to be optional for good feedback

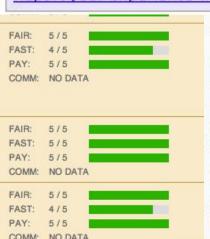
### Build Your Reputation as a Requestor

- Word of mouth effect
  - Workers trust the requester (pay on time, clear explanation if there is a rejection)
  - Experiments tend to go faster
  - Announce forthcoming tasks (e.g. tweet)

#### **Crowd Worker Communities**



Turkopticon.com
Mturkforum.com
Turkernation.com



Small batch and mega bubbles. Not sure if I'm going in....

Title: Which is the most appropriate type?

Requester: Philippe Cudre-Mauroux [A28PIN9Y6KHR3H] (TO)

Description: Please read the text and select the most appropriate description

for each of the proposed entities.

Reward: \$0.10

Qualifications: HIT abandonment rate (%) is less than 51, HIT approval rate

(%) is greater than 25, Location is US

Link: <a href="https://www.mturk.com/mturk/preview?">https://www.mturk.com/mturk/preview?</a> groupId=2ZSQUQIHPCGJ2FZIT6N51H1LQYU60M

Powered by non-amazonian script monkeys ��

To many bubbles but YMMV with your patience level.





7 MYTHS ABOUT HUMAN ANNOTATION

One truth: knowledge acquisition for the semantic web assumes one correct interpretation for every example

All examples are created equal: triples are triples, one is not more important than another, they are all either true or false

**Disagreement bad:** when people disagree, they don't understand the problem

**Experts rule:** knowledge is captured from domain experts

One is enough: knowledge by a single expert is sufficient

**Detailed explanations help:** if examples cause disagreement - add instructions

Once done, forever valid: knowledge is not updated; new data not aligned with old

"Truth is a Lie: 7 Myths about Human Annotation", AI Magazine 2014, L. Aroyo, C. Welty











# disagreement is signal for the natural ambiguity of language and diversity & perspectives of human interpretation

Lora Aroyo, Chris Welty: Truth is a Lie: 7 Myths about Human Annotation, Al Magazine 2014.

Lora Aroyo, Chris Welty: The Three Sides of CrowdTruth. J. Human Computation. 1(1). 2014.

Oana Inel, Khalid Khamkham, Tatiana Cristea, Arne Rutjes, Jelle van der Ploeg, Lora Aroyo, Robert-Jan Sips, Anca Dumitrache and Lukasz Romaszko: Crowd Truth: Machine-Human Computation Framework for Harnessing Disagreement in Gathering Annotated Data. ISWC-RBDS 2014.





Can we a-priori estimate optimal workers and tasks' assignment to obtain maximum accuracy on all tasks?

a two-staged statistical Crowdsourcing Experimental Design

#### Stage 1:

- Train all workers
- On a proportion of tasks
- Identify best workers & 🃤
- Hard tasks

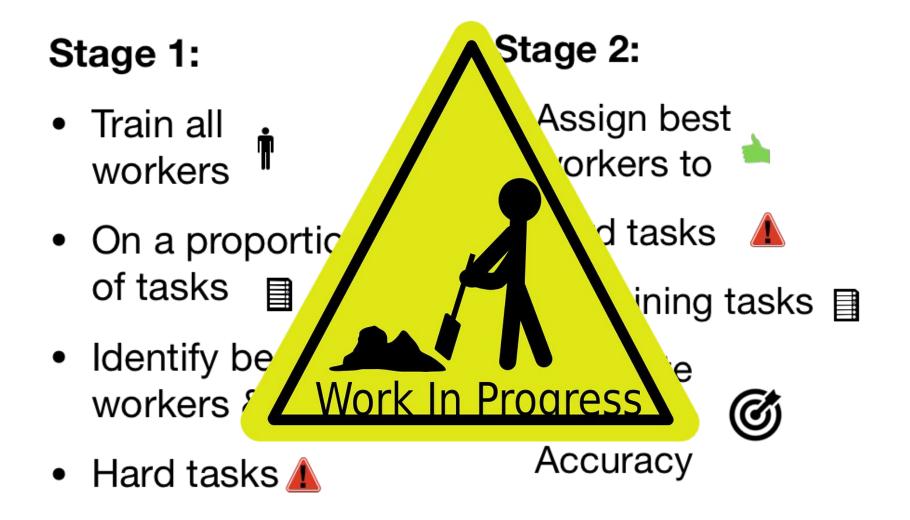
#### Stage 2:

- Assign best workers to
- Hard tasks



- Remaining tasks
- Calculate Overall Accuracy





#### Recommendations

- Reputation system for workers
- More than financial incentives
- Recognize worker potential (badges)
  - Paid for their expertise
- Train less skilled workers (tutoring system)

#### Recommendations

- Promote workers to management roles
  - Create gold labels
  - Manage other workers
  - Make task design suggestions (first-pass validation)
- Career trajectory (based on reputation):
  - Untrusted worker
  - 2. Trusted worker
  - 3. Hourly contractor
  - 4. Employee
- Platforms logs
  - Which kind of tasks attract skilled workers

# Summary

- Things that work
  - Qualification tests
  - Honey-pots
  - Good content and good presentation
  - Economy of attention
- Things to improve
  - Manage workers in different levels of expertise including spammers and potential cases.
  - Mix different pools of workers based on different profile and expertise levels.

# Summary

- Enforce Quality:
  - Task design
  - Iterate
  - Crowd incentives
  - Know your crowd: Model workers

# Hands-on II: Executing a task on CrowdFlower (Figure Eight)

https://www.figure-eight.com/

#### Crowdsourcing Settings - 4 people per group

Group	Task	# Workers (Redundancy)	Contributor Level
1	Data categorization	2	1
2	Data categorization	3	2
3	Data categorization	3	3
4	Data categorization	Dynamic	2
5	Image annotation	2	1
6	Image annotation	3	2
7	Image annotation	3	3
8	Image annotation	Dynamic	2

# Acknowledgements

Slides adapted from the tutorial "Microtask Crowdsourcing to Solve Semantic Web Problems" by Gianluca Demartini, Elena Simperl, and Maribel Acosta at ISWC 2013.

Source: <a href="https://github.com/maribelacosta/crowdsourcing-tutorial">https://github.com/maribelacosta/crowdsourcing-tutorial</a>