



Incentivizing Truthful Feedback on Crowdsourcing Platforms

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Display Size	Memory	Color	Brand
6.8 inches	256 GB	Cosmic Emerald	Motorola
- 

Moto G stylus | 2021 | 2-Day battery | Unlocked | Made for US by Motorola | 4/128GB | 48MP Camera | Black

Introduction

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 - Goal is to rate the quality of a product.
- Challenge is that users who report their reviews may not be truthful.
- This paper proposes a rewarding mechanism to the users so that they are incentivized to provide truthful feedback.

A Simple Example

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- For each type x of plumber, the distribution of observations is $p_Y(x)$.
- Each agent the arrival time of one or more plumbers, and they review the plumbers accordingly.
- We want to estimate $p_Y(x)$ for every plumber type from the users' feedback.
 - Therefore, we want to incentivize workers to be truthful.

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- Y_{wt} denotes the response provided by the w^{th} worker for the t^{th} task
- $v_w(y)$ is rewarded to worker w for a response y .

Problem Statement

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- One way to ensure that agents are truthful is to induce a Bayes-Nash equilibrium.
- A rewarding mechanism induces a Bayes-Nash equilibrium if the strategy profile (response of workers) $\{Y_{wt} : w \in W, t \in T_w\}$ satisfies the inequality:

$$E[v_w(\{Y_{wt} : w \in W, t \in T_w\})] \geq E[v_w(\{Y'_{wt} : w \in W, t \in T_w\} \cup \{Y_{w't} : w' \in W, w' \neq w, t \in T_{w'}\})]$$

for each alternative strategy $\{Y'_{wt} : t \in T_w\} \neq \{Y_{wt} : t \in T_w\}$ and all workers w .

Intuitive Explanation

- The Bayes-Nash Equilibrium says that if all the other workers adhere to the strategy profile $\{ Y_{wt} : w \in W, t \in T_w \}$ then it is beneficial to worker w to also adhere to $\{ Y_{wt} : w \in W, t \in T_w \}$.
- In our case, we would want the truthful strategy to be a Bayes-Nash Equilibrium.
- Further, if the inequality is strict, then it is called a strict Bayes-Nash Equilibrium.

Informational Requirements and Assumptions

- Assumptions and Public Knowledge

- Generating model (P_X, p) is **separated**: For every $y \neq y'$,

$$E_{X \sim P_X} [p_y(X)p_{y'}(X)] < E_{X \sim P_X} [p_y^2(X)] E_{X \sim P_X} [p_{y'}^2(X)].$$

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- **α -separation** quantifies the gap:

$$\alpha + E_{X \sim P_X} [p_y(X) p_{y'}(X)] < E_{X \sim P_X} [p_y^2(X)] E_{X \sim P_X} [p_{y'}^2(X)].$$

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- Workers and host do not know the generating model (P_X, p)
 - Know that the model is separated.
- Workers and host assume that everyone is statistically identical.

Square Root Agreement Rule

- W_t = set of workers that respond to task t .
- T_w = set of tasks that worker w submitted a response to.
 - T = All tasks; $|T_w| \leq T \forall w$.
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Square Root Agreement Rule

- Repeat the following for every worker $w \in [n]$:
 - Let T_w be the set of tasks answered by worker w .
 - For every task $t \notin T_w$ **not** answered by this worker, select two different workers i, j who replied to task t .
 - Compute their #agreements on every possible observation y :

$$A(y) = \sum_{t \notin T_w} 1\{Y_{it} = Y_{jt} = y\}$$

- Scale appropriately to define a popularity index $I_w(y)$ for every observation y :

$$I_w(y) = \frac{1}{T - |T_w|} \left(1 + \sum_{t \notin T_w} 1\{Y_{it} = Y_{jt} = y\} \right).$$

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- For a task t that worker w answered (say y), reward the worker only if she agrees with an arbitrarily selected co-worker w' :

$$v_w(Y_w = s) = \frac{1\{Y_{w'} = s\}}{\sqrt{I_w(y)}}, s \in \mathcal{Y}$$

inversely proportional to the popularity of the agreement.

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- Agents are statistically identical; If I observe y , it is likely that others observe y .
 - This incentivizes truthful behavior, based on an agent's observation.
- Issue: Workers may find it more likely for others to submit a **highly popular observation**.

Intuition 2: Importance of Popularity Index

- Popularity index, asymptotic:

$$I_w(y) = \frac{1}{T - |T_w|} \left(1 + \sum_{t \notin T_w} 1\{Y_i = Y_j = y\} \right) \\ \xrightarrow{T_w, T \rightarrow \infty} P(Y_i = Y_j = y).$$

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- Repels workers from submitting an answer based on its popularity.

Result: SRA induces a Bayes-Nash Eq.

- Suppose all other agents $\neq w$ are truthful.

$$E[v_w(s)] = \begin{cases} \frac{P(Y_{w'} = y | Y_w = y)}{I_w(y)} : s = y \text{ (truthful)} \\ \frac{P(Y_{w'} = y' | Y_w = y)}{I_w(y')} : s = y' \neq y \end{cases} .$$

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- Truthful reward \geq False reward if

$$\begin{aligned} & P(Y_w = Y_{w'} = y)P(Y_w = Y_{w'} = y') \geq P(Y_{w'} = y', Y_w = y) \\ \Leftrightarrow & \sum_x P_X(x)p_y^2(x) \sum_x P_X(x)p_{y'}^2(x) \geq \left(\sum_x P_X(x)p_y(x)p_{y'}(x) \right)^2 \end{aligned}$$

for every $y' \neq y$.

Result Cont'd: Strict Bayes Nash

$$P(Y_w = Y_{w'} = y)P(Y_w = Y_{w'} = y') > P(Y_{w'} = y', Y_w = y)$$

- SRA is **strictly Bayes-Nash incentive compatible** *in expectation* if model is separable for some $\alpha > 0$.
 - Difference in reward for submitting true vs. false response is lower bounded by a value

$$\omega = \omega(T, \alpha, |\mathcal{Y}|, n)$$

when T is sufficiently large.

Extensions Overview

- Ex-ante collaboration: Workers may agree to submit a strategy $\sigma_w(s|y) \in \Delta(\mathcal{Y})$

before the game starts.

- Symmetric strategy: $\sigma_w = \sigma \ \forall w$.
 - Everyone submits $s = 1$ regardless of observation y .
- [Theorem] Truthful responses strictly maximize rewards over all symmetric strategies when $\#tasks \rightarrow \infty$.

Collusion and Equilibrium

- Workers are not allowed to communicate throughout the game.
- It may be possible that workers communicate/collaborate beforehand to reduce effort while maintaining high reward.
- One such implication is a symmetric strategy profile, where all workers agree on a fixed modification $y \mapsto q(y)$.
 - For example, we want to get rid of cases when submitting trivial answers can achieve a high payoff. Otherwise, workers can receive high reward with low effort.

Collusion and Equilibrium

- Uninformativeness of a symmetric strategy profile q is define as

$$\Omega(q) = \frac{1}{|Y|(|Y| - 1)} \sum_{y, y'} \sum_{y'' : y'' \neq y'} \sqrt{q_y(y')q_y(y'')}$$

and say that a strategy q is ω -uninformative if $\Omega(q) \geq \omega$.

- $\Omega(q) = 0 \Leftrightarrow q$ is fully-informative: $q.(y)$ has disjoint supports.
- $\Omega(q)$ is maximized if the report q is chosen independently of the true answer.
- Given enough tasks, a fully-informative strategy profile maximizes reward. In other words, SRA is strongly truthful across symmetric equilibria, asymptotically.

Strong Truthfulness Over All Equilibria

- Crowdsourcing host can choose how to assign tasks.
- Suppose $\frac{wN}{n} < M$ (fix to our notation).
- Asymmetric strategies that may arise (Sec. 5.3).
- State theorem 4.

Limitations and Our Extension

- Because workers are assumed to be statistically identical, a majority vote is the most accurate estimate of the underlying truth.
- When workers are heterogeneous, e.g. follow the Dawid-Skene model

$$p_w = P(Y_{wt} = y^*) \forall t,$$

then SRA fails to incentivize truthful behavior.

- Most algorithms that efficiently aggregate worker responses are designed for the Dawid-Skene model, and it is of interest how to incentivize truthful behavior under such settings.
- Mechanisms designed for heterogeneous workers require extraneous reports, i.e. is not minimal.
- We are currently designing a mechanism that incentivizes truthful response when workers are heterogeneous.

Summary

- SRA incentivizes truthful responses without requiring extraneous reports.
- Under SRA, an honest response maximizes workers' rewards over symmetric equilibria.