Applications

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2016

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- Self-monitoring
- Community Sensing
- Prediction Polls
- Human Computation
- Peer Grading
- Product Reviews

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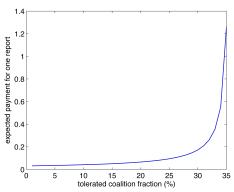
- ISP, mobile phone service, cloud computing services provided to many users under Service Level Agreements (SLA).
- Agreement will stipulate penalties for insufficient quality.
- Monitoring and proving insufficient quality is costly...
- ...and best done by users themselves.
- \Rightarrow self-monitoring with incentives for truthfulness.

R. Jurca et al.: *Reliable QoS Monitoring Based on Client Feedback*, Proceedings of the 16th International World Wide Web Conference (WWW07), pp. 1003-1011, 2007

- Service (e.g. weather forecast) provided to a homogeneous population of users.
- 2 quality parameters:
 - Q_1 : response before deadline (0/1)
 - Q_2 : provided information is correct (0/1)
- cost of service = 1
- benefit for misreport = 0.01
- cost of misreporting (any number of reports): 10

Model for Peer Prediction

- Prior probability of $Q_{1/2} = 0.9$.
- Posterior probability changes 20%: p(1|1) = 0.92, p(1|0) = 0.88 and p(0|0) = 0.12, p(0|1) = 0.08.



Community Sensing



- Pollution is a distributed phenomenon needs many sensors to form a map.
- Government sensors not considered trustworthy.
- \Rightarrow collect data from a community of sensors.

Example: Air Quality Egg

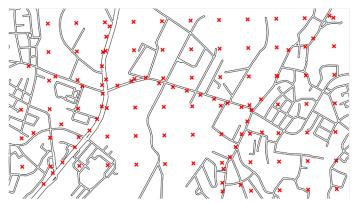


- Open source design developed in a kickstarter project
- Sold over 1000 times at \$185.
- Measurements uploaded to a center controlled by manufacturer.
- Sensor quality insufficient for meaningful measurements.

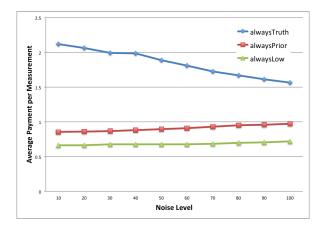
- Mechanism requires peer reports, but no 2 sensors measure at the same location.
- Use prediction of Gaussian process model based on reports close by.
- Only reports in the same time interval are compared; previous aggregate is used as prior distribution *R*.
- B. Faltings et al.: *Incentive Mechanisms for Community Sensing*, IEEE Transaction on Computers, 63(1), 115-128, 2014

Evaluation

• Simulation on an air pollution model of the city of Strasbourg (France):



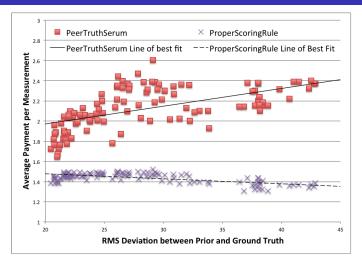
Rewards are Resistant to Noise



• Truthtelling remains the best strategy even with 100% measurement noise.

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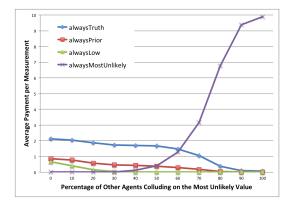
Rewarding Valuable Data



 PTS rewards data collection in uncertain areas, unlike proper scoring rules.

- Incentives are vulnerable to collusion:
 - all agents report the same value x.
 - coordinate on x with smallest R[x].
- Complicated by aggregation into Gaussian model.
- Collusion by many agents will be hard to implement.

Resistance to Collusion



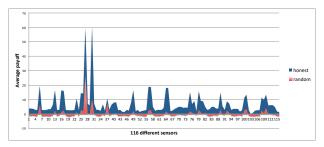
- Collusion unprofitable unless > 60% of agents participate.
- Least likely value is the only promising strategy, but carries no reward unless at least half the agents participate.

- Same simulation model (Strasbourg).
- Use measurement distribution in entire area for R.
- Use correlated neighbouring sensors as peers.
- Consider NO₂ pollution, separated into 4 discrete levels.

G.Radanovic et al.: Incentives for Effort in Crowdsourcing using the Peer Truth Serum, ACM TIST, 2016

Static Sensors

Strategy	mean	min	max	median
honest	6.779	-0.03	59.969	3.658
collude	2.323	-0.146	21.769	1.045
colludeLow	0	0	0	0
random	0.022	-1.974	26.779	-1.076
randomAll	0.071	-2.161	2.137	0.175



Locations with unusual values are more profitable.

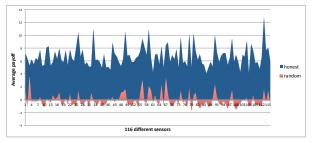
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Mobile Sensors

Strategy	mean	min	max	median
honest	6.779	4.064	12.941	6.456
collude	2.323	1.052	5.141	2.027
colludeLow	0	0	0	0
random	-0.008	-1.781	3.714	-0.294
randomAll	0.03	-1.446	1.792	-0.109



Mobility results in equal opportunities (note different scale).

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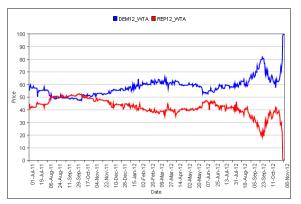
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Classical opinion polls no longer work:

- few people answer surveys \Rightarrow biased results.
- Internet polls have even more bias.
- \Rightarrow need to make polls attractive:
 - incentives
 - gamification
 - wide participation

Predicting Elections

- lowa electronic market: operating since 1988.
- Mainly predicting elections in the US, with real money.
- Actual trading, no automated market maker.



- Prediction can be a sport.
- \Rightarrow rewards = score \Rightarrow position in leaderboard.
 - Examples: Swissnoise, Scicast.
 - Usually research projects with limited duration.

- Public prediction platform operated at EPFL from spring 2013 to summer 2015.
- Users can suggest questions to put up on the platform.
- Initially run as prediction market (with logarithmic scoring rule market maker), later added PTS as alternative.
- Users randomly distributed among prediction market/PTS to compare accuracy and behavior.
- F. Garcin and B. Faltings: *Swissnoise: Online Polls with Game-Theoretic Incentives.* Proceedings of the 26th Conference on Innovative Applications of AI, 2972-2977, 2014

Swissnoise



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Predict the future!



C How it works...



Ask

Start by asking any forecasting questions you might want to know. The crowd's wisdom will help you predict the outcome of your event.



Predict

You make a prediction on the outcome of future events by buying or selling shares on a market with virtual money.



Win!

If your prediction is correct, the market rewards you for every share you hold.

The contest is running!

You start with 5000 π (our virtual money), and you can get up to USD1'000 in prizes! The player with the highest profit of the week wins!

Interested? sign up! More info about the rules of the contest here.



★ Hall of fame

	player	amount
1	sseebb	USD240.00
2	link	USD100.00
3	lina	USD80.00
4	xawill	USD40.00
4	leclem	USD40.00
5	damghani	USD20.00
5	faltings	USD20.00
5	arnaud	USD20.00
5	richard.faltings	USD20.00
5	xhanto	USD20.00
5	noobii	USD20.00

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- 250+ users
- avg. 15 unique users/day
- 230+ questions
- 19'700+ operations
- CHF 20 gift card to user with highest profit of the week.

Example Questions



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Placing a Bet



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- Market maker computes price changes.
- Slider lets user see how many shares are obtained.



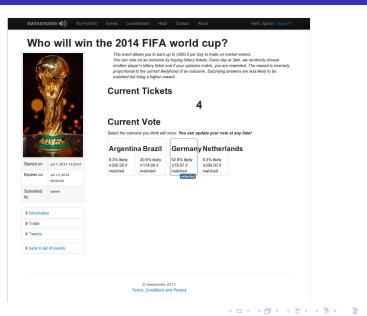
trading stocks



lottery tickets

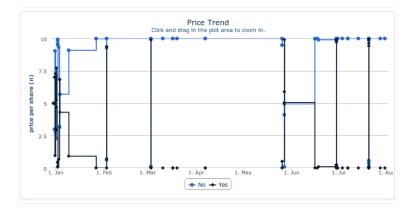
- Lottery compares ticket with randomly chosen peer ticket.
- Reward paid of both agree.
- Lottery run once per day at 3am with all unprocessed tickets.
- After lottery, predictions are integrated.
- Users can buy tickets for multiple consecutive days.

Betting with Lottery



- Risk aversion: unlikely answers need to carry a higher reward.
- Scale according to Weber-Fechner's law: sensitivity to variations should be scale-invariant.
- \Rightarrow would mean increasing rewards *exponentially* with 1/R[x].
 - This proved too extreme: users gambled on least likely values.
 - Solved by averaging between 1/R and $e^{1/R}$.

Will Scotland be independent (Pred. Market)



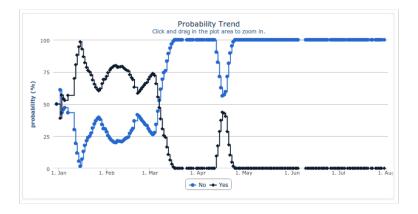
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Will Scotland be independent (Peer Prediction)



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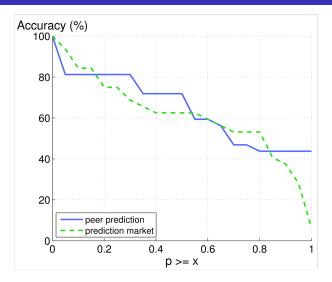
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- Problem with prediction market: cashing rewards changes predictions.
- Very tricky to adjust liquidity parameter (how much price influenced by share demand).
- PTS is more stable: no perturbation from participants who cash in their winnings.
- PTS more "fun": rewards can be gained every day.
- Accuracy is about the same.
- But Peer Prediction does not require ground truth ⇒ more widely applicable.

Accuracy Comparison



• Both schemes have very similar accuracy.

- Outsource tasks to human computation platforms such as Amazon Mechnical Turk.
- Workers are tempted to cut corners and guess answers without solving the task.
- Many tasks have similar answers \Rightarrow worker errors have bias.
- Bias is impossible to correct by increasing number of workers.
- Can we do better by incentivizing workers?
- B. Faltings et al.: *Incentives to Counter Bias in Human Computation* Proceedings of HCOMP 2014, pages 59-66. AAAI, 2014

- Assume bias is known as prior answer distribution R.
- \Rightarrow derive payment using PTS (1/R) principle.
 - Example task: counting binoculars in an image.



- Count imaging devices.
- Correct answer = 34.
- Priming to 34 and 60.
- No bonus, vague bonus, peer consistency and PTS.

Influence of Priming

Bonus	Priming	Average Error	t-test
no bonus	none	1.0667	
	60	5.6316	p = 0.0266
	34	2.9434	p = 0.2092
vague	none	2.2500	
	60	6.6563	p = 0.0810
	34	9.0984	p = 0.0032
peer conf.	none	0.3492	
	60	3.3429	p = 0.0554
	34	2.4194	p = 0.1496
peer conf.	none	0.3492	
PTS	60	0.8000	p = 0.4036
	34	2.1667	p = 0.2145

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Bonus Scheme	Priming	Average Error	t-test
none	60	5.6316	
vague	60	6.6563	p = 0.3782
peer conf.	60	3.3429	p = 0.1306
PTS	60	0.8000	p = 0.0088
none	34	2.9434	
vague	34	9.0984	p = 0.0110
peer conf.	34	2.4194	p = 0.4020
PTS	34	2.1667	p = 0.3731

PTS corrects bias.

• Little influence if priming is to correct value.

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- Each student grades solutions of 5 others.
- 2 types of questions:
 - fill in a piece of code.
 - correct incorrect code.
- Answers categorized into 4 classes (correct and 3 different kinds of wrong answers).
- Incentives = bonus points for the course.
- Ground truth = expert grader (Ph.D. student).

G.Radanovic et al.: Incentives for Effort in Crowdsourcing using the Peer Truth Serum, ACM TIST, 2016

Error rates (measured against expert grader):

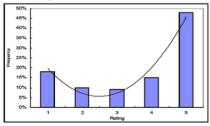
Mechanism	Num. student	Error rate (%)
PTSC	16	6.88
peer consistency	16	10.48
constant	14	11.98

Decrease is significant (p-values):

Mechanism	PTSC	peer consistency	constant
PTSC	-	0.0255	0.0497
peer consistency	0.0255	-	0.5566
constant	0.0497	0.5566	-

- Writing reviews is work \Rightarrow needs a reason:
- Being extremely satisfied (or paid by the seller).
- Being extremely dissatisfied (or paid by competitor).
- Distribution of reviews can have a "U" shape.

Figure 3: Distribution of the Ratings on Amazon.com (fitted with a U-shaped curve) for a Music CD (Mr. A-Z)



- Solution: Reviewers need to be paid for their data!
- Best criterion: reviews should be *suprising* and *confirmed*.
- \Rightarrow Peer Truth Serum is a good scheme.
 - So far no practical experiment...

- Rewarding reviews: difficult to evaluate since ground truth is not known.
- Applying PTSC to crowdwork: can no longer do experiments on AMT.
- Actual crowdsensing implementation: better sensor platform in development, hope we can do this in 2017.