

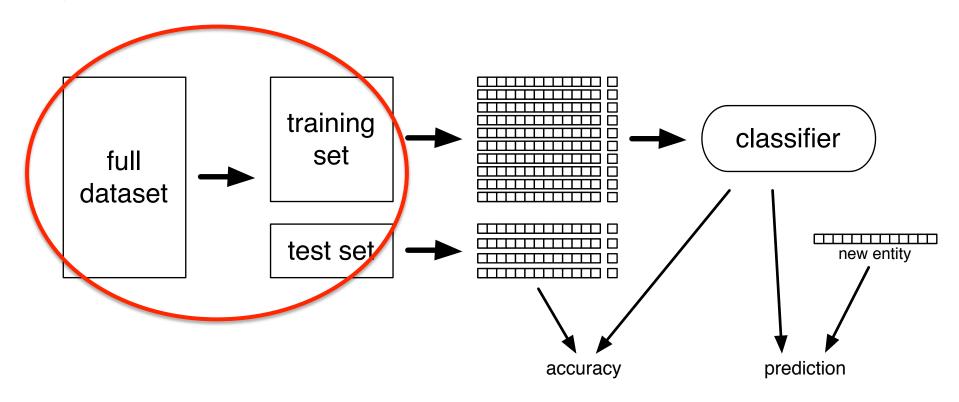
Crowdsourcing for Analytics

Tim Kraska

<kraska@cs.berkeley.edu>

Machines alone are not enough...

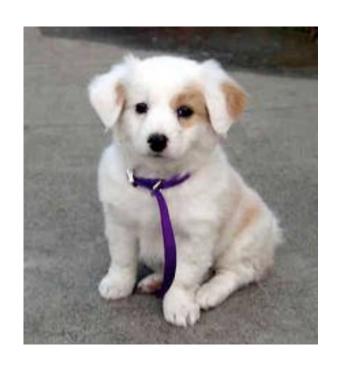
Classification





Machines alone are not enough...

SELECT Image From Pictures Where Image contains "Dog"









Adding People to Analytics

Data collection

America's top 10 NASDAQ companies with female CEOs

Transcription





Data cleaning



Creativity/Design/Taste

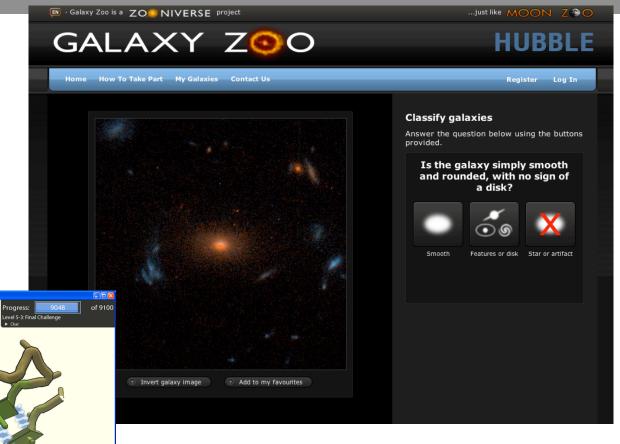




and more



Science

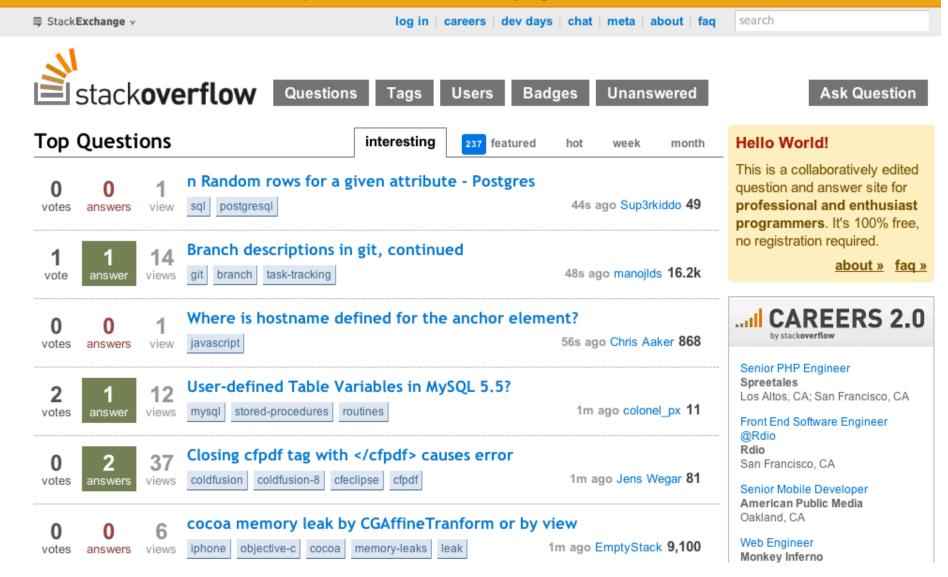




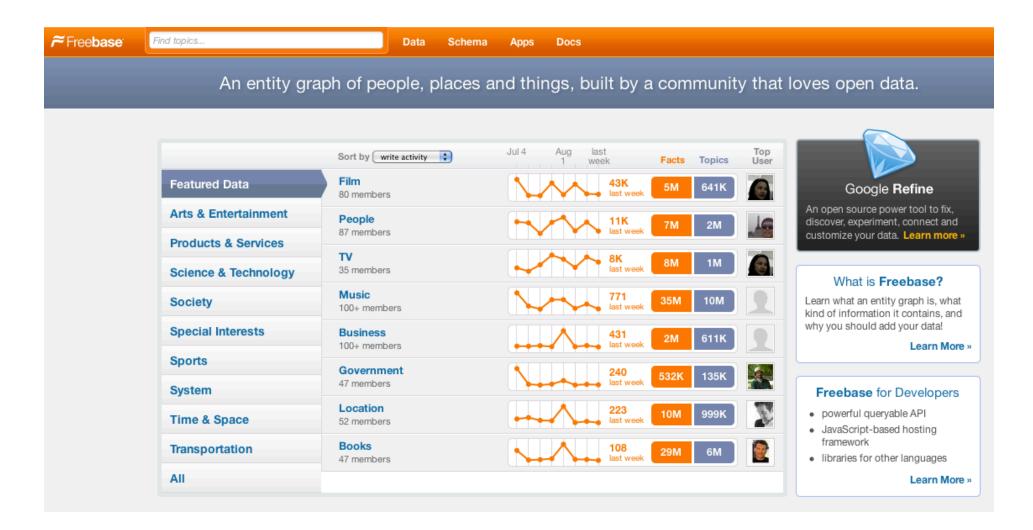


Knowledge Basis

Welcome to Q&A for professional and enthusiast programmers — check out the FAQ!

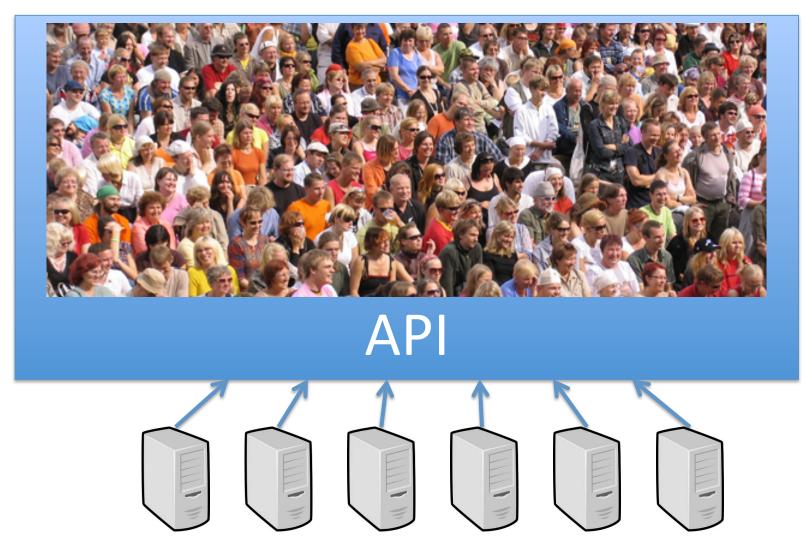


Structured Data





Crowdsourcing for Developers 101



Micro-Task CrowdSourcing



Make Money by working on HITs

HITs - Human Intelligence Tasks - are individual tasks that you work on. Find HITs now.

As a Mechanical Turk Worker you:

- Can work from home
- Choose your own work hours
- Get paid for doing good work



Get Results

from Mechanical Turk Workers

Ask workers to complete HITs - Human Intelligence Tasks - and get results using Mechanical Turk. Get started.

As a Mechanical Turk Requester you:

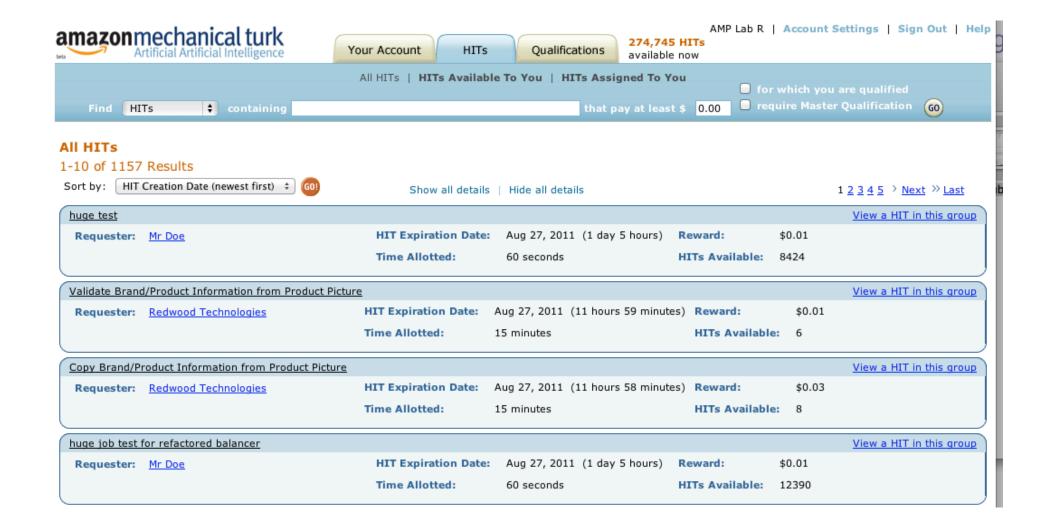
- Have access to a global, on-demand, 24 x 7 workforce
- Get thousands of HITs completed in minutes
- Pay only when you're satisfied with the results



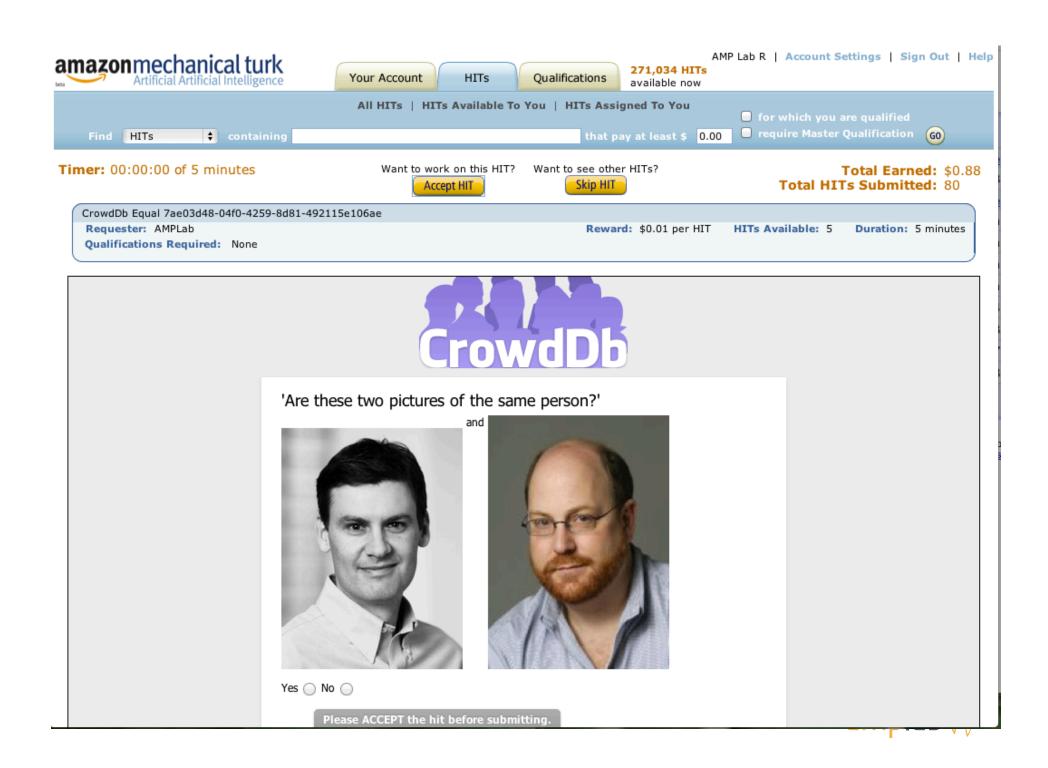


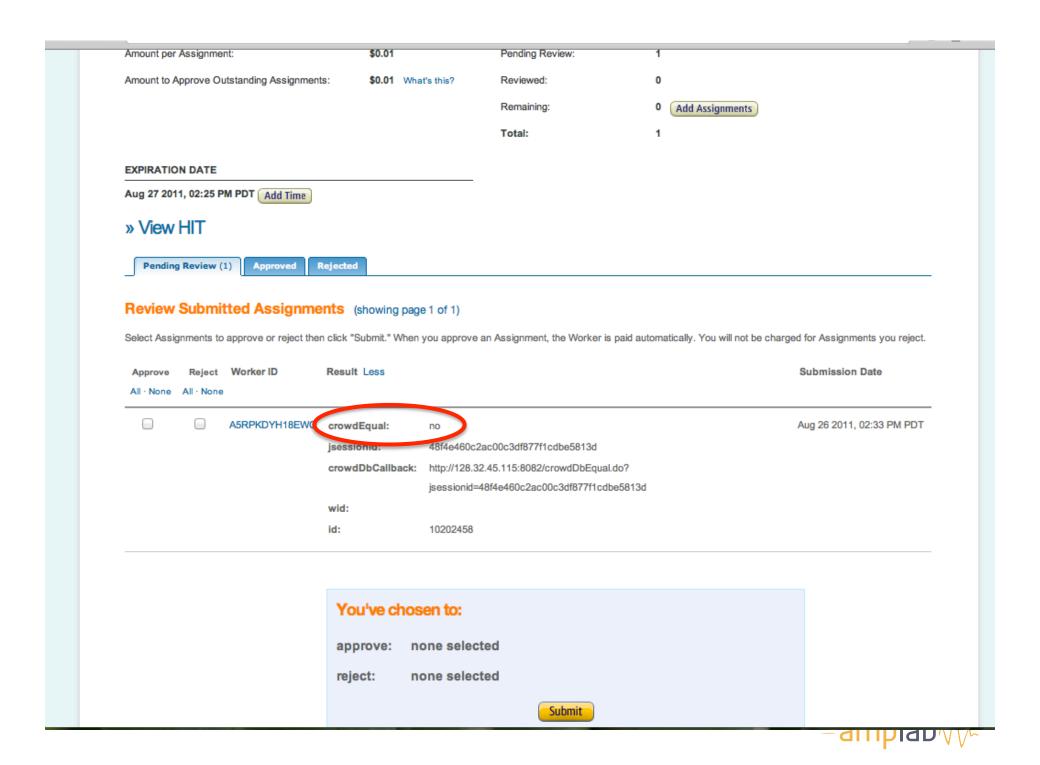
Microtasking – Virtualized Humans

- Current leader: Amazon Mechanical Turk
- Requestors place Human Intelligence Tasks (HITs)
 - Minimum price: \$0.01
 - + #of replicas (assignments), expiration, <u>User Interface</u>
 - API-based: "createHit()", "getAssignments()", "approveAssignments()", "forceExpire()"
 - Requestors approve jobs and payment
- Workers (a.k.a. "turkers") choose jobs, do them, get paid









Samasource.org



Challenges

- Quality
- User Interface Design
- Worker motivation
- Task decomposition
- Leverage worker knowledge/capabilities
- Optimization (time/cost)

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How Can You Trust the Crowd?





Quality Techniques

- Approval Rate / Demographic Restrictions
- Gold Sets/Honey Pots
- Redundancy
- Qualification Test
- Verification/Review
- Justification/Automatic Verification

• ...



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Approval Rate & Demographic Restrictions

```
Classify text about consumer electronics

Requester: Buzz Evaluation

HIT Expiration Date: Sep 7, 2011 (1 week 6 days) Reward: $0.02

Time Allotted: 20 minutes

HITs Available: 3966

Description: Classify text for positive, negative, mixed or neutral tone

Keywords: buzz, classify, coding, tag, sentiment, text, analysis, twitter, blog, social

Qualifications Required:

HIT approval rate (%) is not less than 95

Location is US
```

- + Easy to setup
- + Transparent
- Easy to defeat
- Causes a lot of trouble



Approval Rate



HIT Group » I recently did **299 HITs for this requester**.... Of the 299 HITs I completed, **11 of them were rejected** without any reason being given. **Prior to this I only had 14 rejections, a .2% rejection rate.** I currently have **8522 submitted HITs, with a**

0.3% rejection rate after the rejections from this requester (25 total rejections). I have attempted to contact the requester and will update if I receive a response. Until then be very wary of doing any work for this requester, as it appears that they **are rejecting about 1 in every 27 HITs being submitted.** posted by ...

fair:2/5 fast:4/5 pay:2/5 comm:0/5

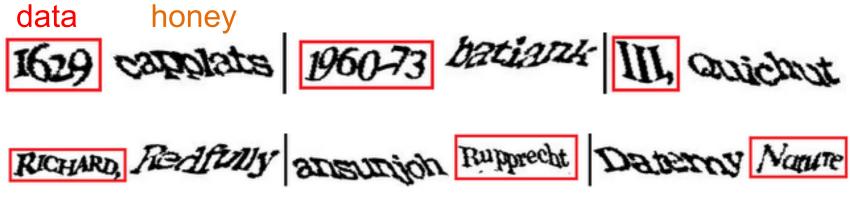
Gold Sets / Honey Pots



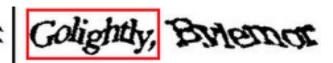
- Gold derived from
 - Experts
 - Crowd using high quorum
- Interject trap questions
- Block users in trap and invalidate answers
- + Often very effective
- + Cost efficient
- Not always applicable
- Digging gold is hard

Defeating Honey Pots: reCAPTCHA





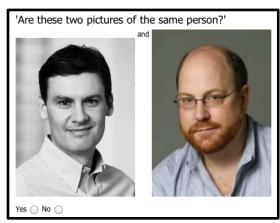






Redundancy: Quorum Votes







majority vote

result

- + Easy to implement
- + Hard to defeat
- Increased cost
- Masks cases of ambiguity or diversity, "tail" behaviors
- Does not cover bias



Challenges

- Quality
- User Interface Design
- Worker motivation
- Task decomposition
- Leverage worker knowledge/capabilities
- Optimization (time/cost)

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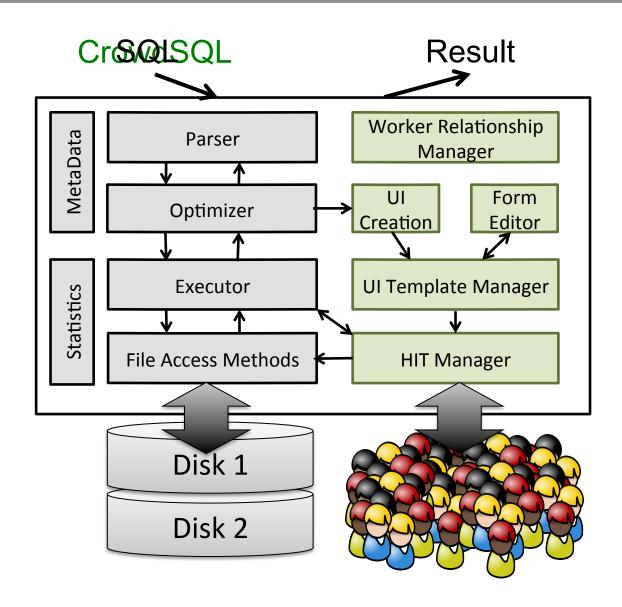




Use Cases

- Data collection:
 - How do my prices compare to the prices of my competitors
 - Finding job candidates (who is graduating from HPI next year)
 - Find green-tech companies in the Bay Area
 - **—** ...
- Data cleaning
 - Verifying customer addresses
 - Duplicate elimination
 - **—** ...
- Extending data
 - Labeling (spam/not_spam)
 - ...

CrowdDB





CrowdSQL

DDL Extensions:

Crowdsourced columns

```
CREATE TABLE company (
  name STRING PRIMARY KEY,
  hq_address CROWD STRING);
```

Crowdsourced tables

```
CREATE CROWD TABLE department (
   university STRING,
   department STRING,
   phone_no STRING)
PRIMARY KEY (university, department);
```

DML Extensions:

CrowdEqual:

```
SELECT *
FROM companies
WHERE Name ~ "Big Blue"
```

CROWDORDER operators (currently UDFs):

```
SELECT p FROM picture
WHERE subject =
    "Golden Gate Bridge"
ORDER BY CROWDORDER(p, "Which
pic shows better %subject");
```



Optimization: Quality

CROWD TABLE professor(name, e-mail) CROWD TABLE department(name, phone-nb) name="Carev" **SELECT** * p.dep=d.name FROM professor p, department d WHERE d.name = p.dep Professor Department AND p.name ="Michael J. Carey" Please fill out the missing Please fill out the missing professor data department data N ame Carev Department CS Name Department cs Please fill out the missing name **MTJoin** professor data **MTJoin** Phone E-Mail (Dep) Carey Name (Professor) Submit p.dep = d.nameSubmit p.name = "carey **MTProbe** E-Mail (Professor, Dep) Department name=Carey Please fill out the missing Department Please fill out the missing **MTProbe** professor data Phone department data (Professor) Name Carey MTProbe(Dep) Department name=Carey Submit Name E-Mail Department **Phone** Submit Submit (Department first) (Professor first) (De-normalized Probe) **Inefficient** ≈10% Error-Rate ≈80% Error-Rate



A Bigger(?) Underlying Issue

Closed-World





Open-World





What Does This Query Mean?

SELECT COUNT(*) FROM IceCreamFlavors

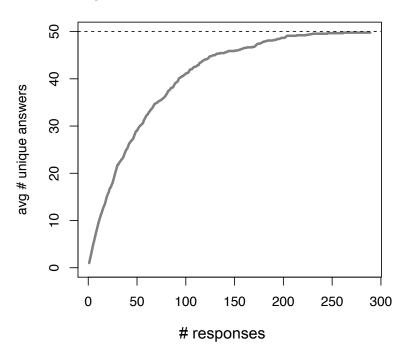


Trushkowsky et al. Getting it All From the Crowd, (in preparation) on arxiv

Estimating Completeness

SELECT COUNT(*) FROM US States US States using Mechanical Turk

Unique items over time



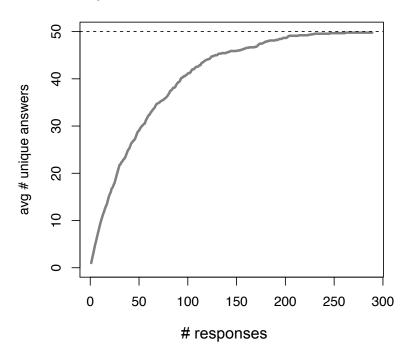




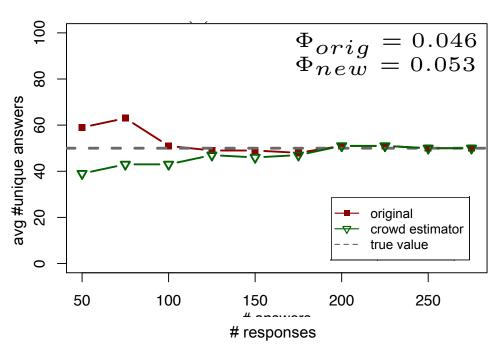
Estimating Completeness

SELECT COUNT(*) FROM US States US States using Mechanical Turk

Unique items over time



Crowd set-size estimation

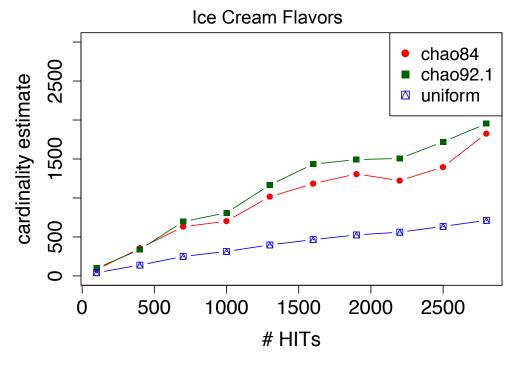




Estimating Completeness

SELECT COUNT(*) FROM IceCreamFlavors

- Ice Cream Flavors
 - Estimators don't converge
 - Very highly skewed(CV = 5.8)
 - Detect that # HITs insufficient (beginning of curve)



Few, short lists of ice cream flavors (e.g. "alumni swirl, apple cobbler crunch, arboretum breeze,..." from Penn State Creamery)

Pay-As-You-Go

- "I don't believe it is usually possible to estimate the number of species... but only an appropriate lower bound for that number. This is because there is nearly always a good chance that there are a very large number of extremely rare species"
 - Good, 1953
- So instead, can ask: "What's the benefit of m additional HITs?"

Ice Cream after 1500 HITs

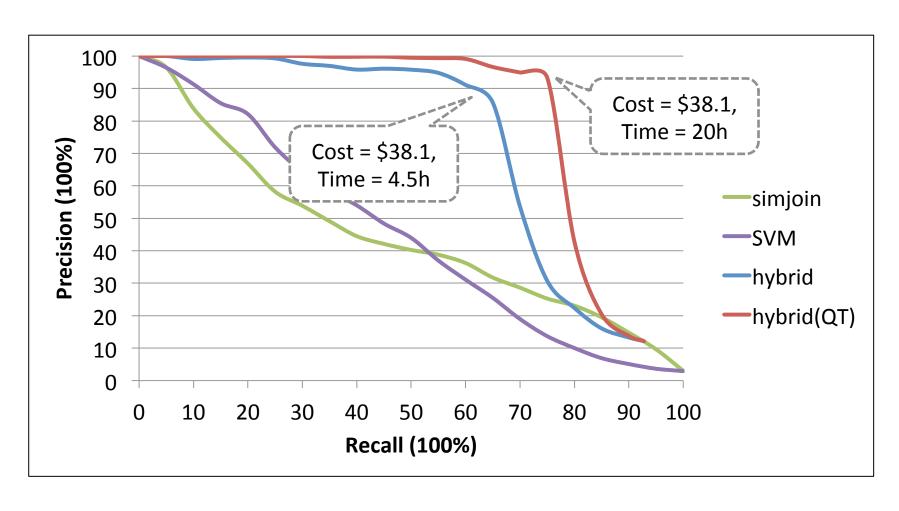
m	Actual	Shen	Spline
10	1	1.79	1.62
50	7	8.91	8.22
200	39	35.4	32.9



Entity Resolution

ID	Product Name	Price
r_1	iPad Two 16GB WiFi White	\$490
r_2	iPad 2nd generation 16GB WiFi White	\$469
r_3	iPhone 4th generation White 16GB	\$545
r_4	Apple iPhone 4 16GB White	\$520
r_5	Apple iPhone 3rd generation Black 16GB	\$375
r_6	iPhone 4 32GB White	\$599
r_7	Apple iPad2 16GB WiFi White	\$499
r_8	Apple iPod shuffle 2GB Blue	\$49
r_9	Apple iPod shuffle USB Cable	\$19

Hybrid Entity Resolution



J. Wang et al. CrowdER: Crowdsourcing Entity Resolution, PVLDB 2012



Human-Tolerant Computing

Adding People into the Analytics Lifecycle:

- Inconsistent answer quality
- Incentives
- Latency & Variance
- Open vs. Closed world
- Hybrid Human/Machine Design



Approaches:

- Statistical methods for error and bias
- Quality-conscious Interface design
- Cost (time, quality)-based optimization

Summary

The AMPLab looks into integrating Algorithms, Machines and People for big data analytics

- Crowdsourcing can help with Big Data analytics where machines are not enough
- CrowdDB is a first hybrid Crowd/Cloud data management system following this vision
- Full tutorial: Crowdsourcing Applications and Platforms: A Data Management Perspective. VLDB, 2011
- Try it out at mturk.com

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