# CrowdDB: Answering Queries with Crowdsourcing

Franklin, Michael J., Donald Kossmann, Tim Kraska, Sukriti Ramesh, and Reynold Xin. "CrowdDB: answering queries with crowdsourcing." In SIGMOD Conference, pp. 61-72. 2011.

> Presented by Patricia Ortega February/2013



## Outline

- Introduction
- Problem definition
- Crowdsourcing
- CrowdDB
- **User Interface Generation**
- Query Processing
- **Experiment and Results**
- Conclusion





## Introduction

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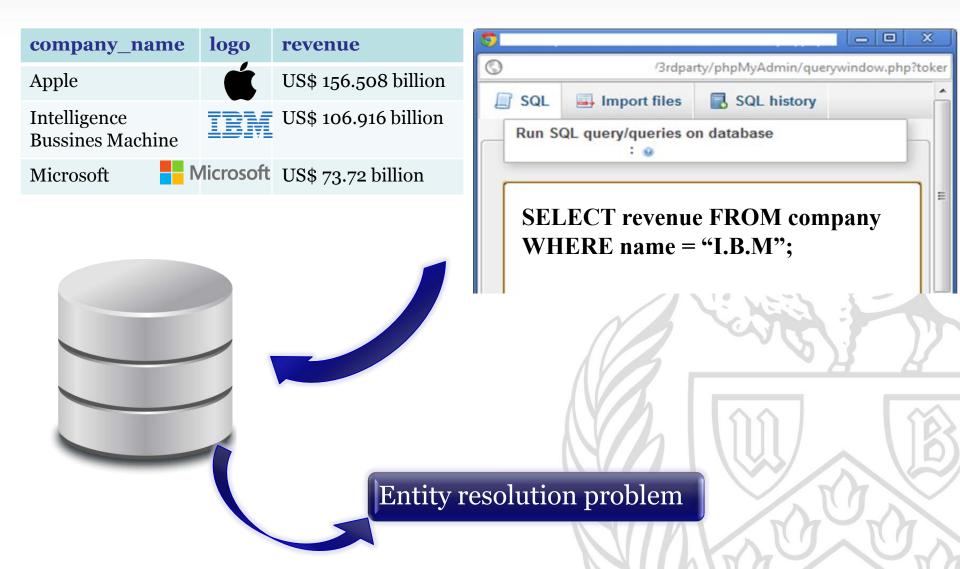
What do they have in common?



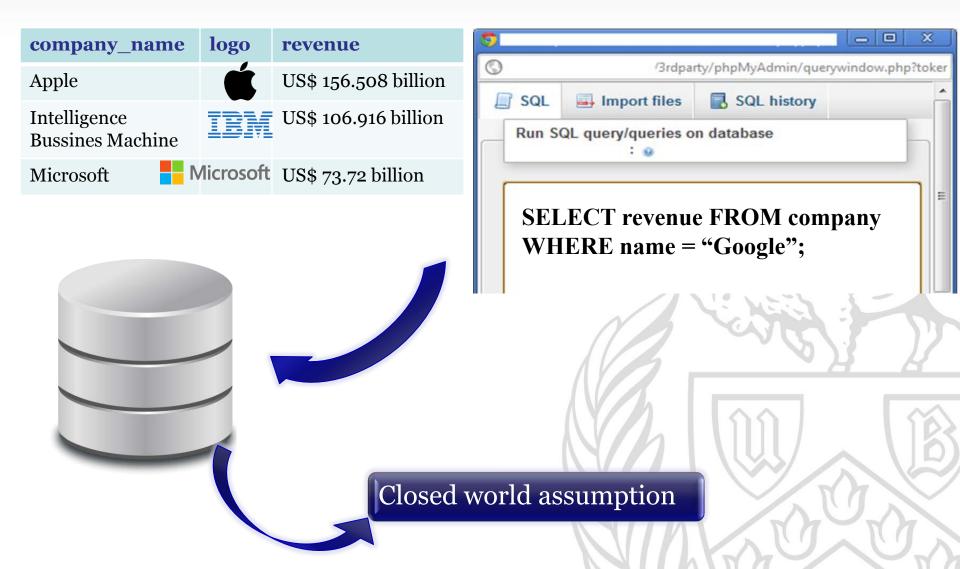




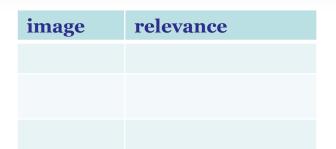


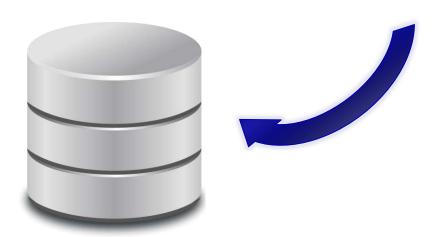






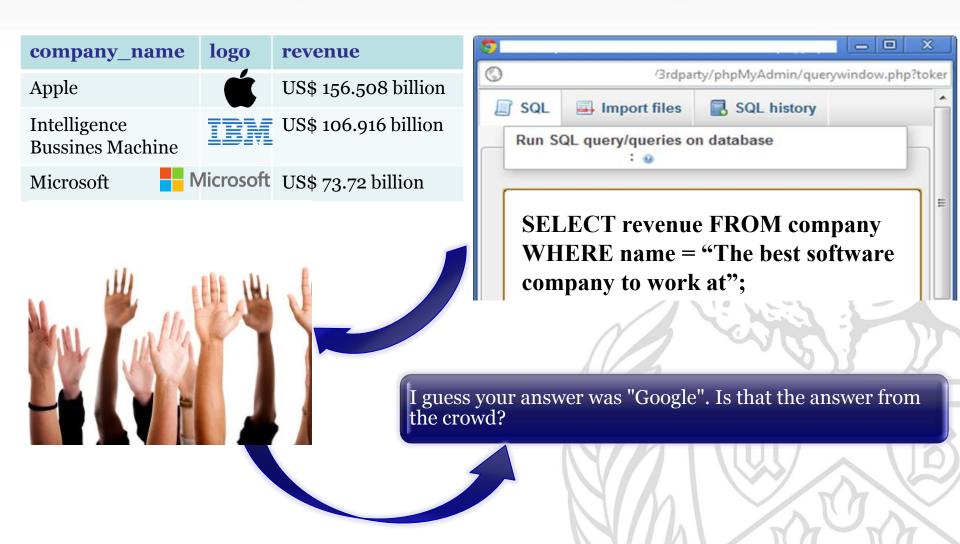














# Crowdsourcing

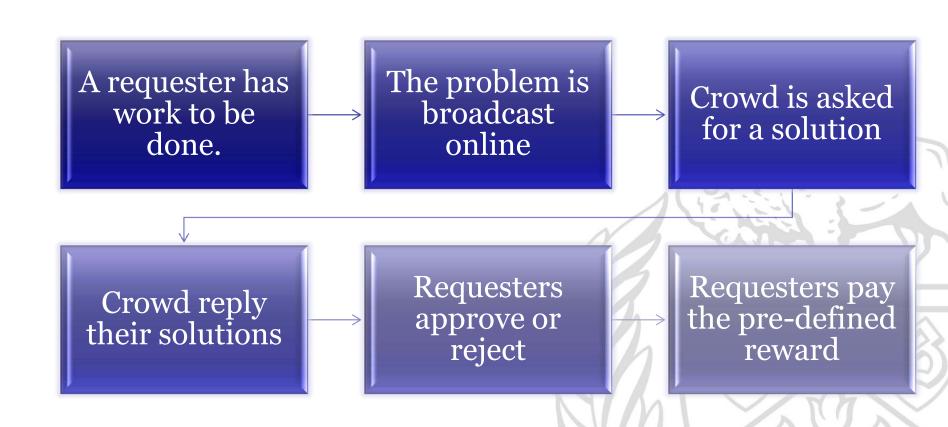
Two main human capabilities that allow corrects answers:

- Finding new data
- Comparing data



### 函

# Crowdsourcing



# Crowdsourcing – Mechanical Turk **Basics**

- Microtasks: No requires special training, typically less than a minute.
- HIT(Human Intelligent Task): The smallest entity of work that could be accepted by a worker.
- Assignment: HIT can be replicated into multiple assignments. A worker can process at most a single assignment per HIT.
- HIT Group: AMT automatically groups HIT's by requester, tittle, description and reward.

## AMT Workflow

- Requesters post HITs.
- AMT post them into compatible HIT groups.
- Worker search, accept and process the assignment.
- Requesters approve or reject.
- For each task completed requesters pay the predefined reward, bonus and commission to Amazon.



## Mechanical Turk APIs

### Create new HIT:

createHit(tittle,description,question,keywords,rewa rd, duration, maxAssignments, lifetime): Hitld

### List of assignments of a HIT

getAssignmentsForHIT(HitId):list(ansId,workerId,Ans wer)

### Approve/Reject

approveAssignment(ansId)/rejectAssignment(ansId)



# CrowdDB – Design Considerations

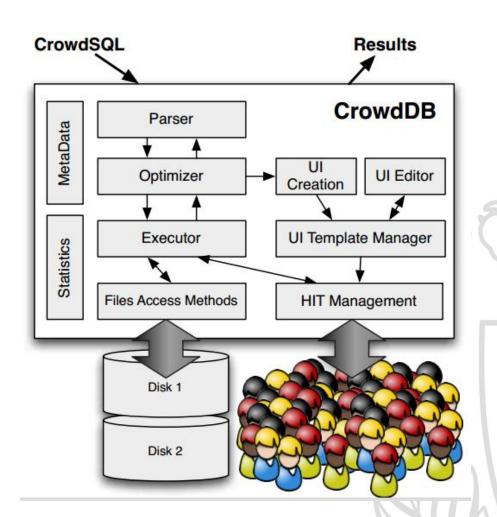
- Performance and variability
  - Work speed
  - Work cost
  - Work quality
- Task design and ambiguity
  - Natural language ambiguity
  - **UI** Design



# CrowdDB – Design Considerations

- Affinity and learning
  - Workers develop skills, and relationships with requesters.
- Relatively small worker Pool
  - Impact in parallelism and throughput
- Open vs. closed world
  - Possible return unlimited number of answers. (Query planning, execution cost, answer quality)

## CrowdDB- Architecture



## **Crowd Components**

### Turker Relationship Manager:

Handles: approving/rejecting assignments, paying, etc.

### User Interface Management:

 CrowdSQL extends data definition language to annotate tables, information used later to create UI.

### HIT Manager:

Manages interaction CrowdDB and crowdsourcing platform



# CrowdSQL

Is a SQL extension that support crowdsourcing.

Minimal extension

 Support use case with missing data and subjective comparisons.

# CrowdSQL - Considerations

### **SQL DDL extensions**

### **Keyword CROWD:**

- Incomplete data can occurs:
  - Specific attributes of tuples
  - Entire tuple

### Crowdsourced column

```
CREATE TABLE Department (
university STRING,
name STRING,
url CROWD STRING, phone STRING,
PRIMARY KEY (university, name) );
```

### **Crowdsourced Table**

```
CREATE CROWD TABLE Professor (
name STRING PRIMARY KEY,
email STRING UNIQUE,
university STRING,
department STRING,
FOREIGN KEY (university, department)
REF Department(university, name));
```

## CrowdSQL - Considerations

### **SQL MDL semantics**

### **Keyword CNULL:**

- Equivalent to NULL
- · Means that value should be crowd sourced at its first use.
- Default value of CROWD column

```
INSERT INTO Department (university,
INSERT INTO
                                    name, url)
Department (university, name)
VALUES ("UC Berkeley", "EECS");
                                    VALUES ("ETH Zurich", "CS",
                                     "inf.ethz.ch");
```

## CrowdSQL - Considerations

### **Query semantics**

- Suppor any kind of query on CROWD tables and columns.
- Incorporates crowdsourced data as part of processing SQL queries.

```
SELECT url FROM Department
WHERE name = "Math";
```

```
SELECT * FROM Professor
WHERE email LIKE "%berkeley%" AND
dept = "Math";
```

# CrowdSQL – Subjective comparisons

To support subjective comparisons has to built in functions (CROWDEQUAL and CROWDORDER):

 CROWDEQUAL: ~= (takes 2 paraters lvalue, rvalue, ask the crowd to decide if values are equals)

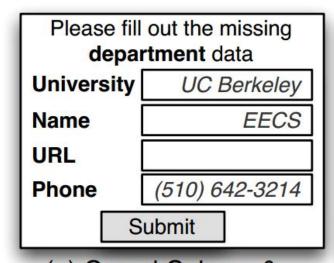
```
SELECT profile FROM department
WHERE name ~= "CS";
```

# CrowdSQL – Subjective comparisons

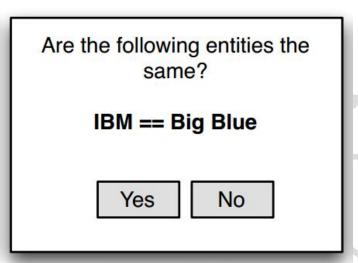
CROWORDER: Used to ask crowd rank the result.

```
CREATE TABLE picture (
p IMAGE,
subject STRING);
SELECT p FROM picture
WHERE subject = "Golden Gate Bridge"
ORDER BY CROWDORDER (p,
"Which picture visualizes better %subject");
```

Key: Provide effective user interfaces.



(a) Crowd Column & Crowd Tables w/o Foreign Keys

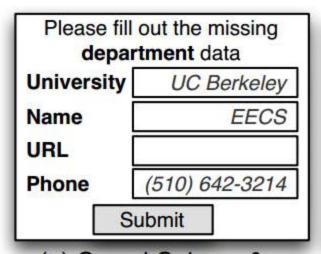


(b) CROWDEQUAL

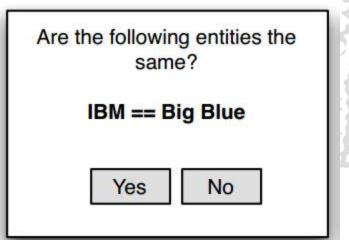
UI key to success in crowdsourcing:

- At compile time, creates templates to crowdsourcing missing information (HTML5, JavaScript)
- These templates are instantiated at runtime providing a UI for a concrete tuple or set of tuples.

Key: Provide effective user interfaces.



(a) Crowd Column & Crowd Tables w/o Foreign Keys



(b) CROWDEQUAL

Key: Provide effective user interfaces.

Which picture visualizes better "Golden Gate Bridge"				
	GOLDEN GATE ARIDO			
•	0			
Submit				

(c) CROWDORDER

Please fill out the <b>professor</b> data				
Name	Richard M. Karp			
Email				
University				
Department				
	Submit			

(d) Foreign Key(normalized)

Key: Provide effective user interfaces.

Please fill out the missing professor data		,	Please fill out the missing department data	
Name	Richard M. Karp	'	University	
Email		1	Name URL	
Department	add		Phone	
	Submit			Submit

(e) Foreign Key (denormalized)



# Query Processing – Crowd Operators

Current version of CrowDB has three crowd operators:

CrowdProbe:

Crowd missing information about CROWD columns and new tuples. (Uses generated UI)



# Query Processing – Crowd Operators

- CrowdJoin:
  - Implement an index nested-loop-join over two tables (at least one crowdsourced).
  - Creates HIT's for each tuple in the inner relation.

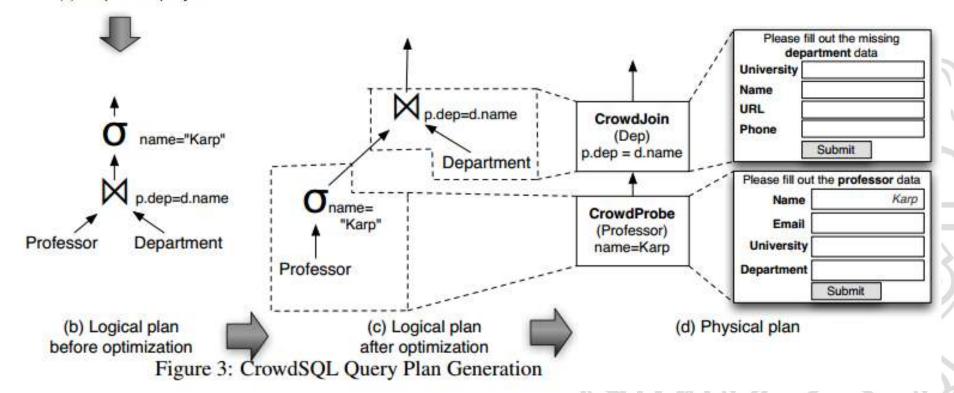
# Query Processing – Crowd Operators

- CrowdComprare:
  - Implements CROWDEAQUAL and CROWDORDER.
  - Instantiate UI.
  - Typically used inside another traditional operator(sorting or predicate evaluation).

# Query Processing – Plan Generation

SELECT \* FROM professor p. department d WHERE p.department = d.name AND p.university = d.university AND p.name = "Karp"

(a) PeopleSQL query





## **Experiments and Results**

Experiments run with CrowdDB and AMT. Ran over 25,000 HITs on AMT during October 2010

- Parameters:
  - Price
  - Jobs per HIT and
  - Time of day.
- Measured the response time and quality of the answers provided by the workers.



## **Experiments and Results**

### Micro-benchmarks:

- Simple jobs involving finding new data or making subjective comparisons.
- Goal:
  - Observe the behavior of workers for the types of tasks required.
  - Obtain insight to develop costs models for query optimization.

# Experiments and Results - Micro Benchmarks

Description: Simple tasks requiring workers to find and fill in missing data for a table with two crowdsourced columns:

```
CREATE TABLE businesses (
name VARCHAR PRIMARY KEY,
phone number CROWD VARCHAR (32)
address CROWD VARCHAR (256));
```

## Experiments and Results - Micro Benchmarks

- Table was populated with names of 3607 businesses (restaurants, hotels, and shopping malls) in 40 USA cities.
- Study the sourcing of the phone\_number and address columns using the following query:

SELECT phone number, address FROM businesses;

# Experiment 1: Response Time, Vary HIT Groups

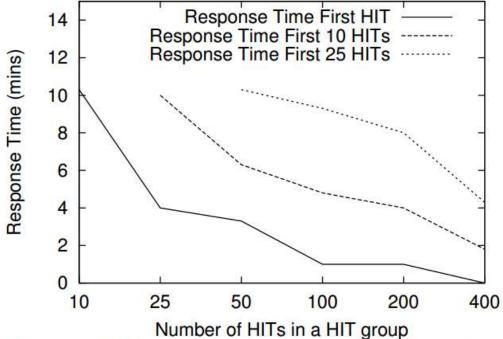


Figure 4: Response Time (min): Vary Hit Group (1 Asgn/HIT, 1 cent Reward)

Time of completion of 1,10,25 group HIT size.

Response time decrease dramatically as size of group is increased.

## Experiment 2: Responsiveness, Vary Reward

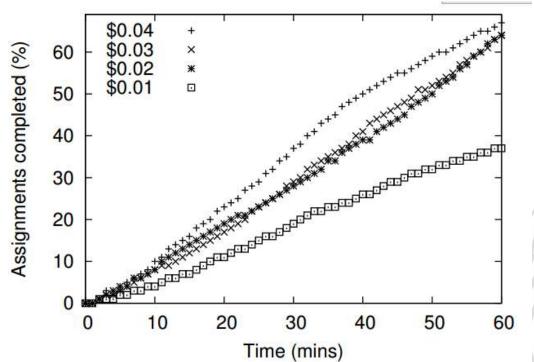


Figure 6: Completion (%): Vary Reward (100 HITs/Group, 5 Asgn/HIT)

How response time varies as a function of the reward.

## Experiment 2: Responsiveness, Vary Reward

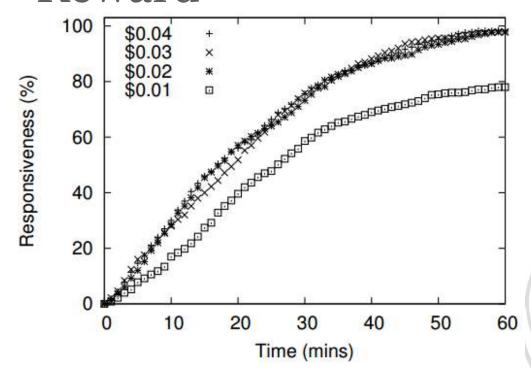
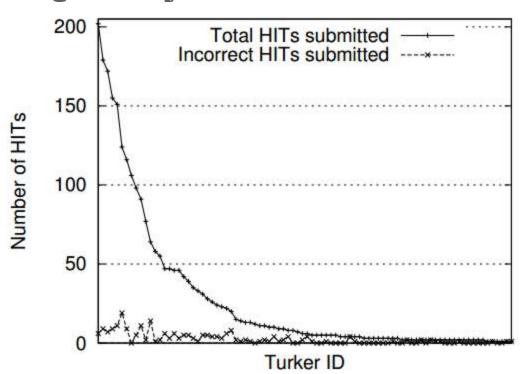


Figure 7: Completion (%): Vary Reward (100 HITs/Group, 5 Asgn/HIT)

Fraction of HITs that received at least one assignment as a function of time and reward

# Experiment 3: Worker Affinity and Quality



Number of HITs computed for a particular worker and the number of errors made for the worker

Figure 8: HITs/Quality by Worker (Any HITs/Group, 5 Asgn/HIT, Any Reward)



# Complex Queries: Entity Resolution on Companies

Non Uniform Name	<b>Query Result</b>	Votes
Bayerische Motoren Werke	BMW	3
International Business Machines	IBM	2
Company of Gillette	P&G	2
Big Blue	IBM	2

SELECT name FROM company WHERE name~="[a nonuniform name of the company]"

Figure 9: Entity Resolution on Company Names

# Complex Queries: Ordering Pictures



Figure 10: Pictures of the Golden Gate Bridge [1] ordered by workers. The tuples in the sub-captions is in the following format: {the number of votes by the workers for this picture, rank of the picture ordered by the workers (based on votes), rank of the picture ordered by experts).

## Conclusion

- CrowdDB is a relational query processing system that uses microtask-based crowdsourcing to answer queries that cannot otherwise be answered.
- Combination of human input with high-powered database processing:
  - · Extends the range of existing database systems.
  - Enables completely new applications and capabilities



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# Questions...?

Thank you.