# Crowdsourced Databases: Query Processing with People

Adam Marcus, Eugene Wu, David Karger, Sam Madden, Rob Miller

MIT CSAIL



# Crowdsourced Databases: Query Processing with People

How to Crowdsource the Introduction of Your Talk to Other Research Groups

### CIDR Deadline



## Good ideas in Databases Jane, John

#### Abstract

Traditional databases *fill in here*. This leads to *fill in here*. We propose *find a good name*, which *what does it do?*.

### OR



### **CIDR 2011**

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- Anastasia Ailamaki, EPFL
- Michael Franklin, UC Berkeley
- Joe Hellerstein, UC Berkeley

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D2 $\checkmark$ $f(x)$ $\Sigma$ = dna@cs.yale.edu									
	A	В	С	D					
1	Name	Affiliation	Phone	Email					
2	Daniel <u>Abadi</u>	Yale	203.436.1265	dna@cs.yale.edu					
3	Anastasia <u>Ailamaki</u>	EPFL							
4	Gustavo Alonso	ETH Zurich							
5	Sihem Amer Yahia	Yahoo! Research							
6	Magdalena <u>Balazinska</u>	Washington							
7	Michael Cafarella	Michigan							

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7 Michael Cafarella	Michigan			



### Turker Interface

Answer a Questio	n					
Requester: Ada		<b>ard:</b> \$0	.05 per HI	T HITs Avail	able: 1	Duration
Qualifications F	Required. None					
What is the email	address and phone	number	of <b>Joe He</b> l	<b>lerstein</b> , who w	orks at <b>U</b>	Berkeley?
		1				
Email Address			Number			
	Human	Inte	elliaei	nce Task		
			···· 5 •·			

(HIT)

## **Uses of Human Computation**

- Data cleaning/integration (ProPublica)
- Finding missing people (Haiti, Fossett, Gray)
- Translation/Transcription (SpeakerText)
- Word Processing (Soylent)
- Outsourced insurance claims processing
- Data journalism (Guardian)

## Challenges in Human Computation

- Workers are not silicon drones
- Worker latency is minutes, hours
- Optimization parameters (price, accuracy) and model are different

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- Workers are not silicon drones
- Worker latency is minutes, hours
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These problems are currently addressed in an ad-hoc way

Qurk is a declarative workflow management system that allows human computation over relational data

Qurk is a declarative workflow management system that allows human computation over relational data

(a system which throws humans in the direct path of query execution)

## Data Model/Query Model

- UDFs that compile to HTML forms
- List types
  - Make relational with Pig-style FLATTEN operator
  - Collapse to single values with UDAs

```
TASK pcInfo(name, affiliation)

Text: "What is the email address and phone number of %s, who works at %s", name, affiliation

Response: Form(("Email Address", String),

("Phone Number", String))
```

SELECT name, FLATTEN(pcInfo(name, affiliation)) FROM pc INTO temporary Q1

# SELECT name, FLATTEN(pcInfo(name, affiliation)) FROM pc INTO temporary Q1

```
(h@berkeley, 123-456-7890)
Joe, ((h@berkeley, (999)4445555))
(bad @ response, 1234567890)

flatten

Joe, h@berkeley, 123-456-7890
Joe, h@berkeley, (999)4445555)
Joe, bad @ response, 1234567890
```

SELECT name, majorityVote(normalize(phone)), majorityVote(normalize(email)), FROM Q1 GROUP BY name

## SELECT name, majorityVote(normalize(phone)), majorityVote(normalize(email)),

### FROM Q1 GROUP BY name

```
Joe, h@berkeley, 123-456-7890
Joe, h@berkeley, (999)4445555)
Joe, bad @ response, 1234567890
```

Joe, h@berkeley, (123)456-7890 Joe, h@berkeley, (999)444-5555 Joe, bad@response, (123)456-7890

## SELECT name, majorityVote(normalize(phone)), majorityVote(normalize(email)),

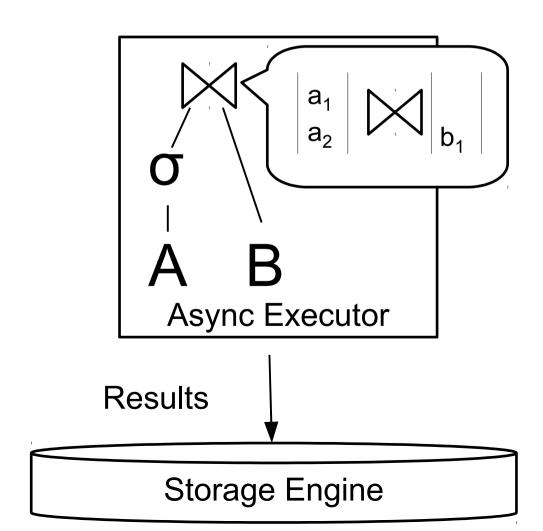
### FROM Q1 GROUP BY name

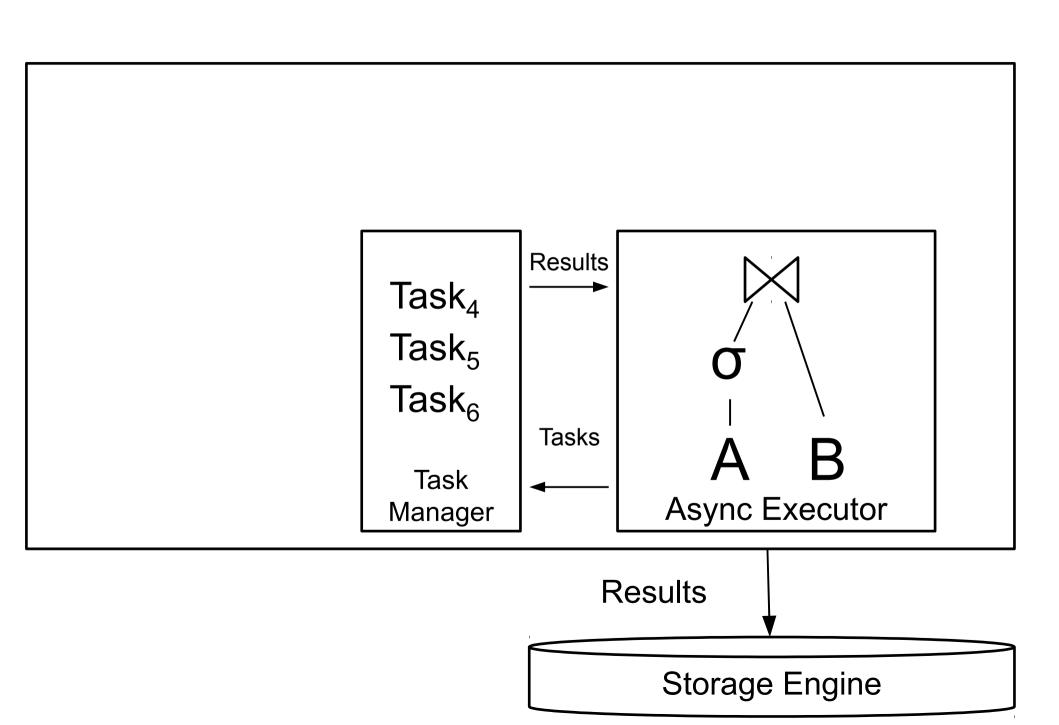
```
Joe, h@berkeley, 123-456-7890
Joe, h@berkeley, (999)4445555)
Joe, bad @ response, 1234567890
Joe, h@berkeley, (123)456-7890
Joe, h@berkeley, (999)444-5555
Joe, bad@response, (123) 456-7890
                majorityVote
Joe, h@berkeley, (123)456-7890
```

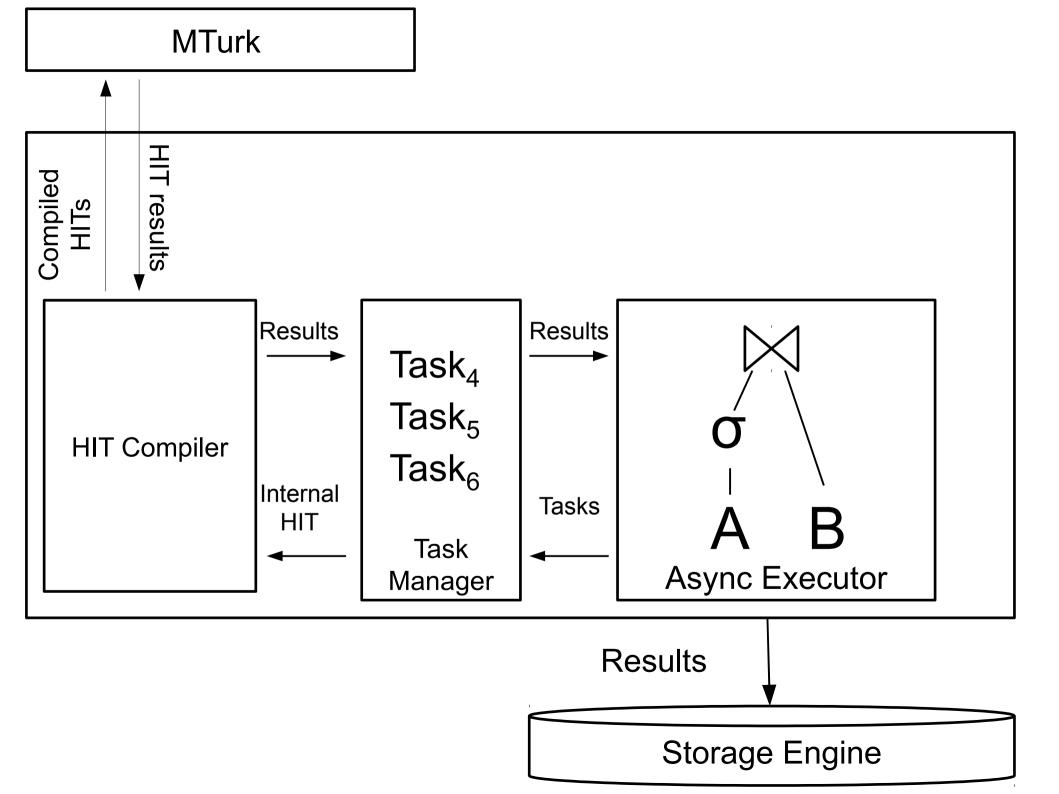
### Haiti Join

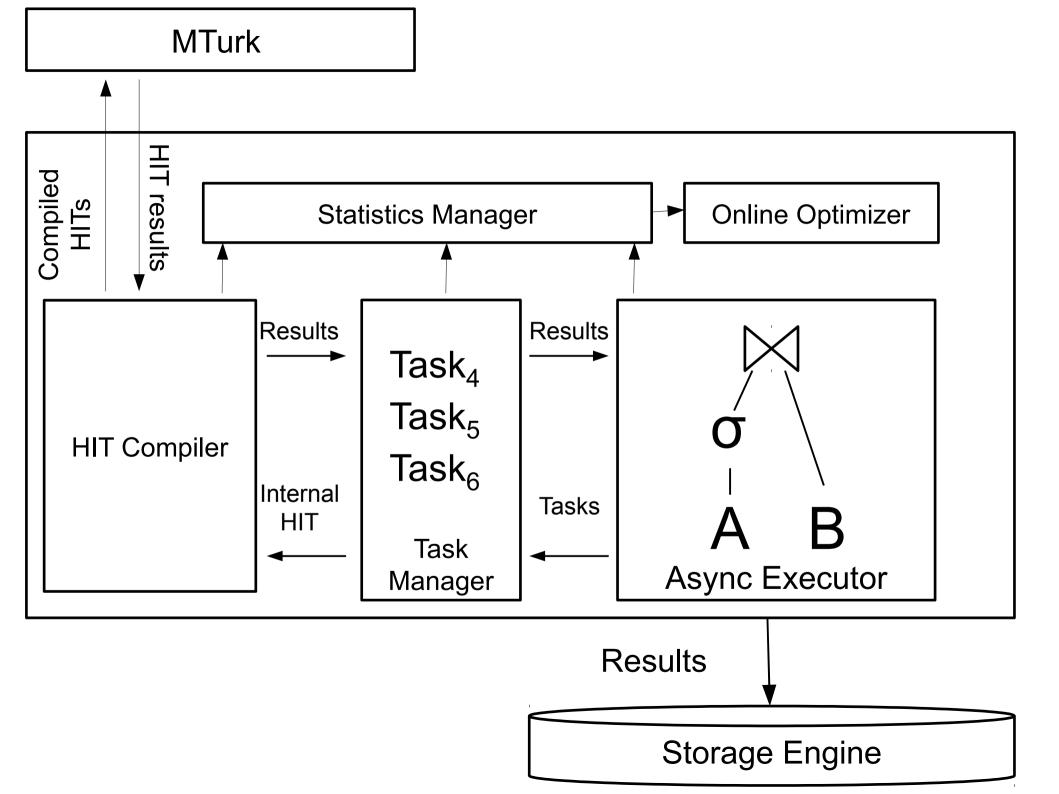
SELECT survivor.name FROM survivor, missing WHERE majorityVote(samePerson(survivor.img, missing.img))

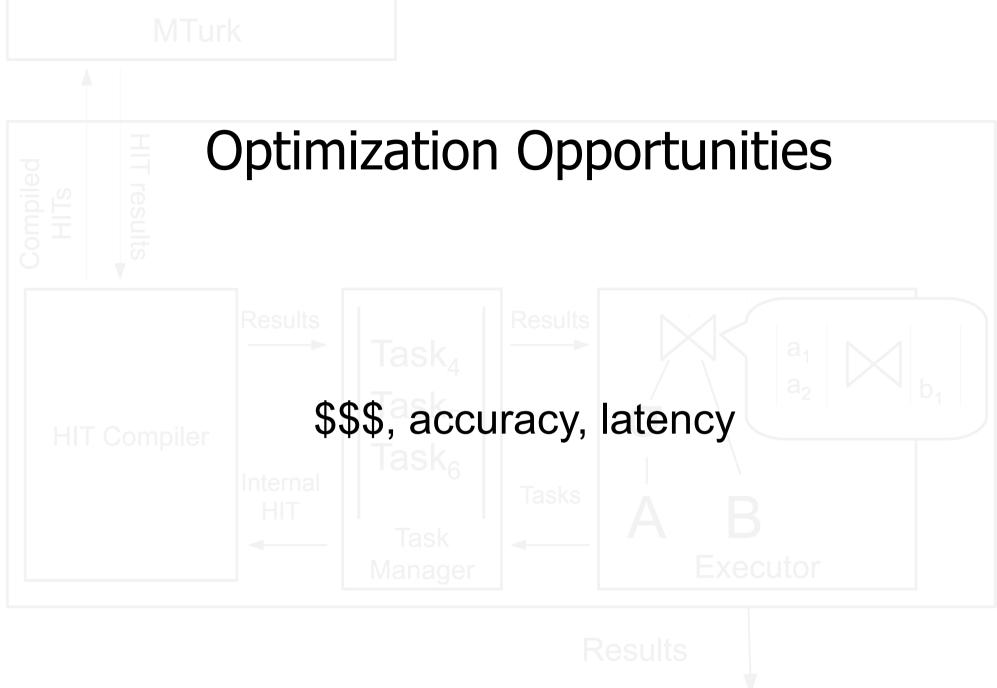
## Qurk System Design





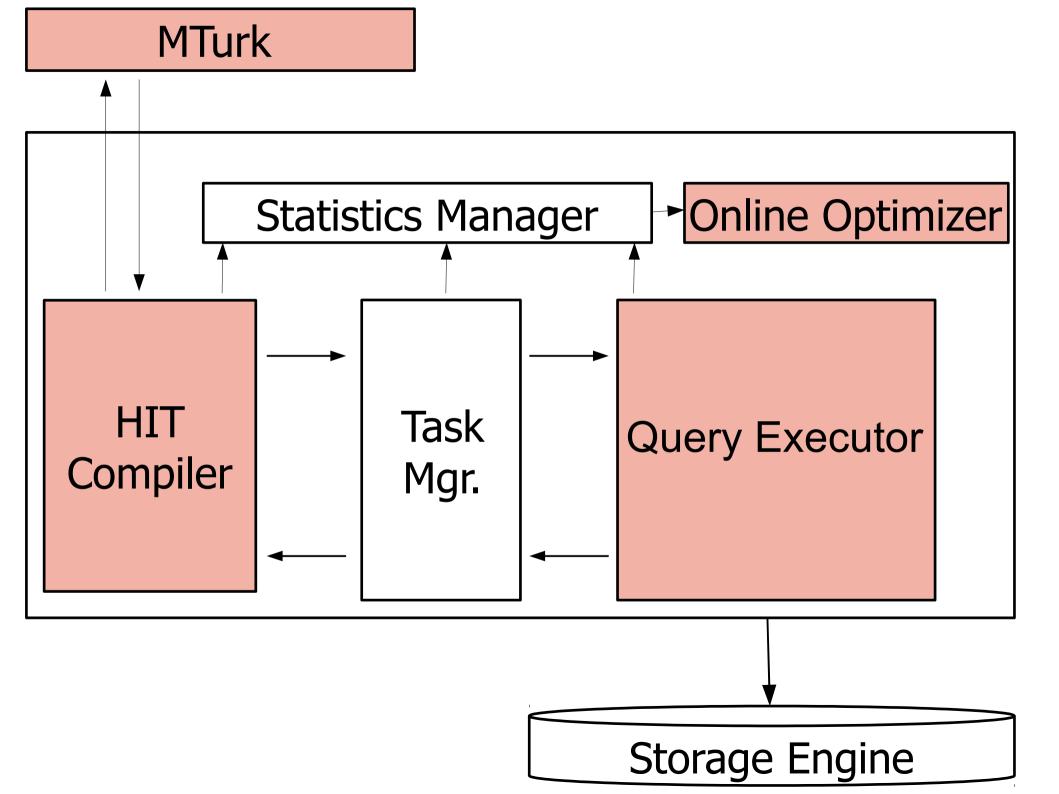


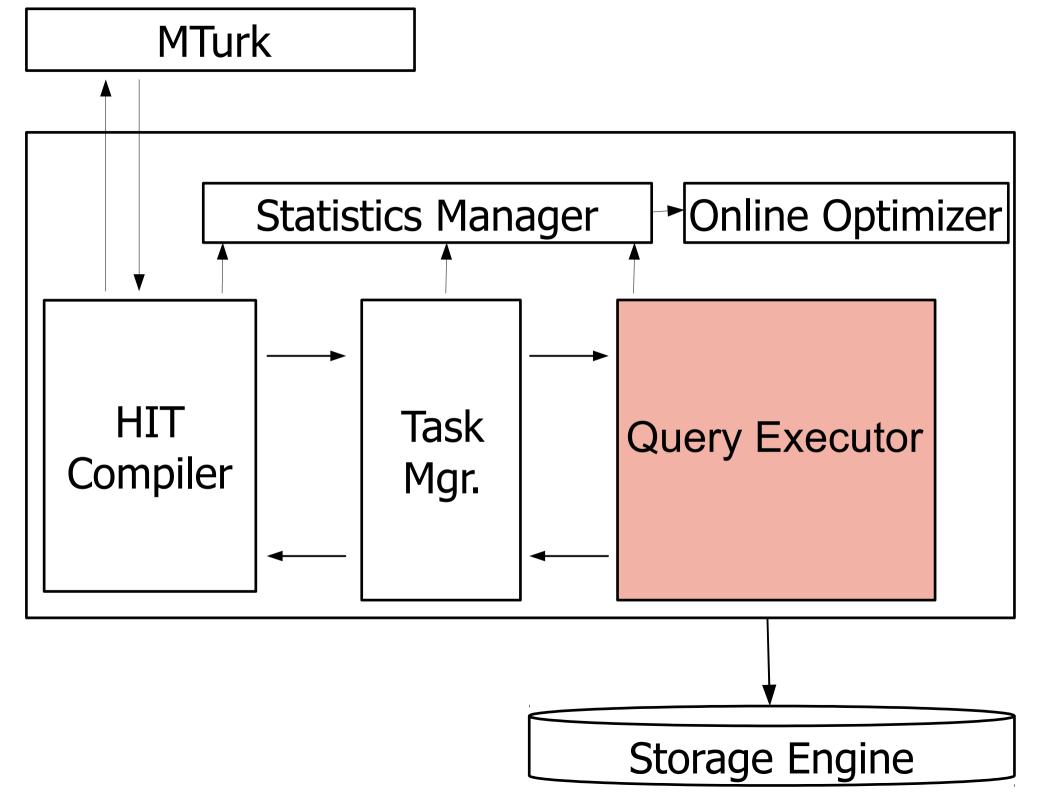




Results

Storage Engine





```
σ

male?

σ

adult?

Photos
```

σ male? σ adult?



Is this a picture of an adult? • Yes • No

σ male? σ adult? Photos



Is this a picture of an adult? • Yes • No

σ | male? | σ | adult? Photos



## **Batch Tuples**

σ male? σ adult? Photos



### Turker 1

Is this a picture of an adult? • Yes • No Is this a picture of a male? • Yes • No



### Turker 2

## **Batch Tuples**

σ male? σ adult? Photos



#### Turker 1



Is this a picture of an adult? • Yes • No Is this a picture of a male? • Yes • No

### **Batch Tuples**

# Early Experimental Result: Batching Tuples

- + Maintains accuracy
  - Increases latency

### Join Optimizations

Avoid cross product

### Join Optimizations

- Avoid cross product
- Technique 1: batching

### Join Optimizations

- Avoid cross product
- Technique 1: batching
- Technique 2: join heuristics reduce search space
  - e.g., image equi-join: gender, ethnicity match

### Turker Fatigue

```
MWXMNMWNMNWMNXMN • Has V • No V XWMXNWMNXXWMNWX • Has V • No V NXNMXWNXWMXNWMN • Has V • No V WMNWXNXWMWXMWXMWXN • Has V • No V NWXNXWMWXMWXMNW • Has V • No V NMWNMNWMNXMNXNX • Has V • No V XWMWXMWXMNWNXWV • Has V • No V NXMNXNMXVVVNXWM • Has V • No V
```

### Turker Fatigue

MWXMNMWNMNWMNXMN 

Has V 

No V XWMXNWMNXXWMNWX 

Has V 

No V NXNMXWNXWMXNWMN 

Has V 

No V WMNWXNXWMWXMWXN 

Has V 

No V NWXNXWMWXMWXMNW 

Has V 

No V NMWNMNWMNXMNXNX 

Has V 

No V XWMWXMWXMNMWNXW 

Has V 

No V NXMNXNMX V VNXWM • Has V • No V **Skew Correction** MWXMNMWNMNWMM 

Has V 

No V XWMXNWMVXXWMN 

Has V 

No V NXNMXWNXWMXNW 

Has V 

No V WMNWXVVXWMWXMW 

Has V 

No V NWXNXWMWXMWXMN 

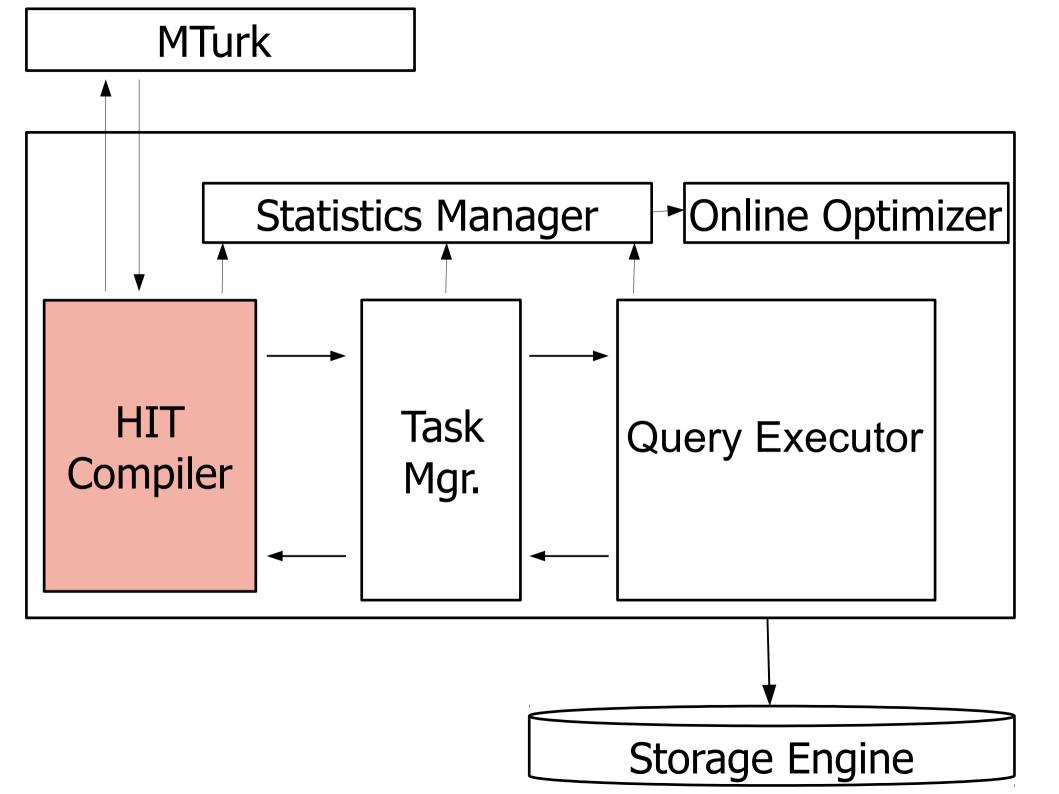
Has V 

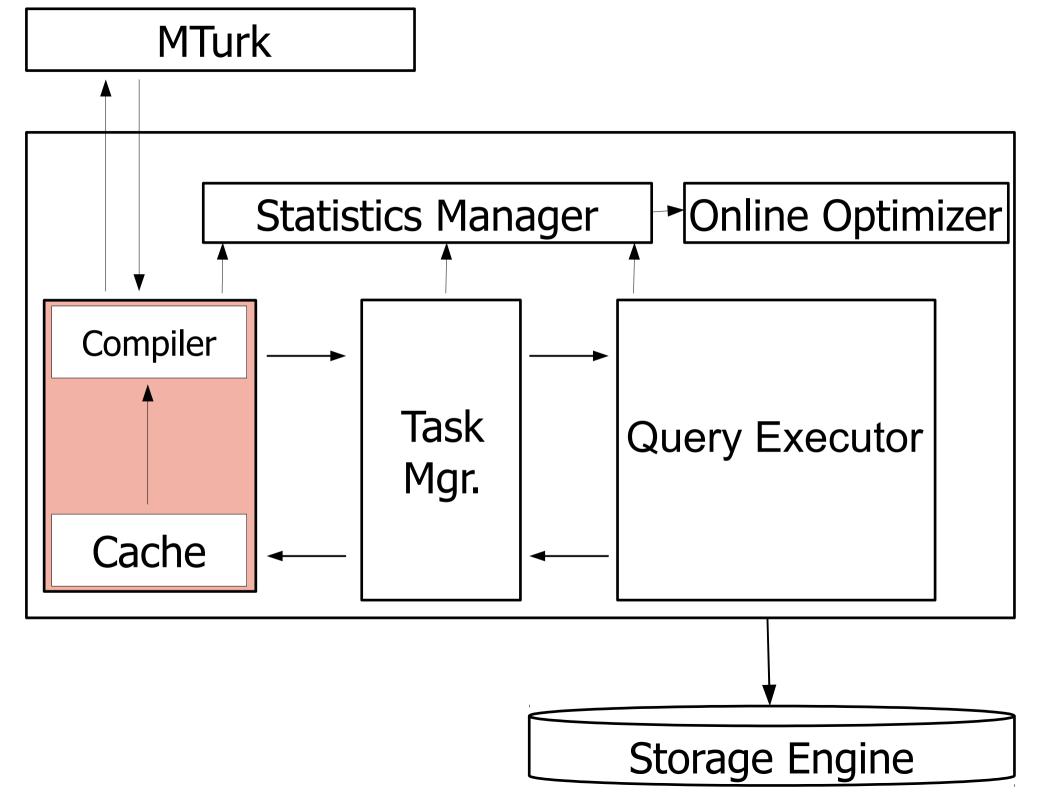
No V NMWNMNV VMNXMNX • Has V • No V XWMWXMWXMNMWNM • Has V • No V NXMNXNMXVVNXWM • Has V • No V

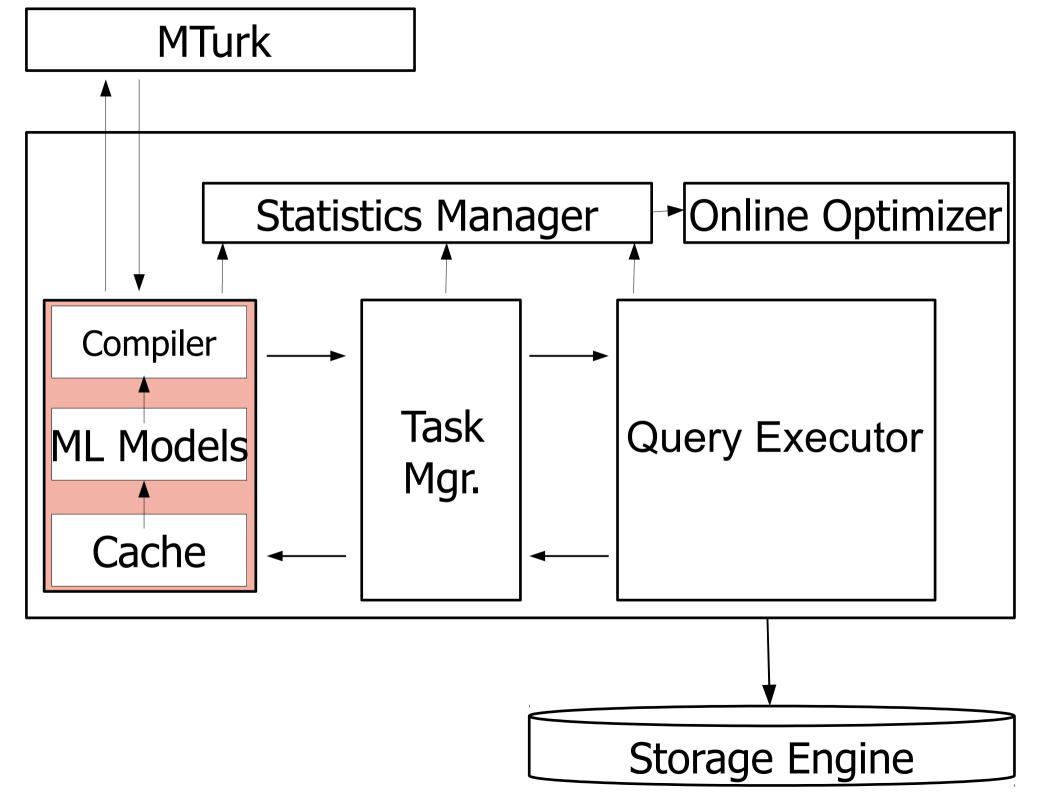
### Turker Fatigue

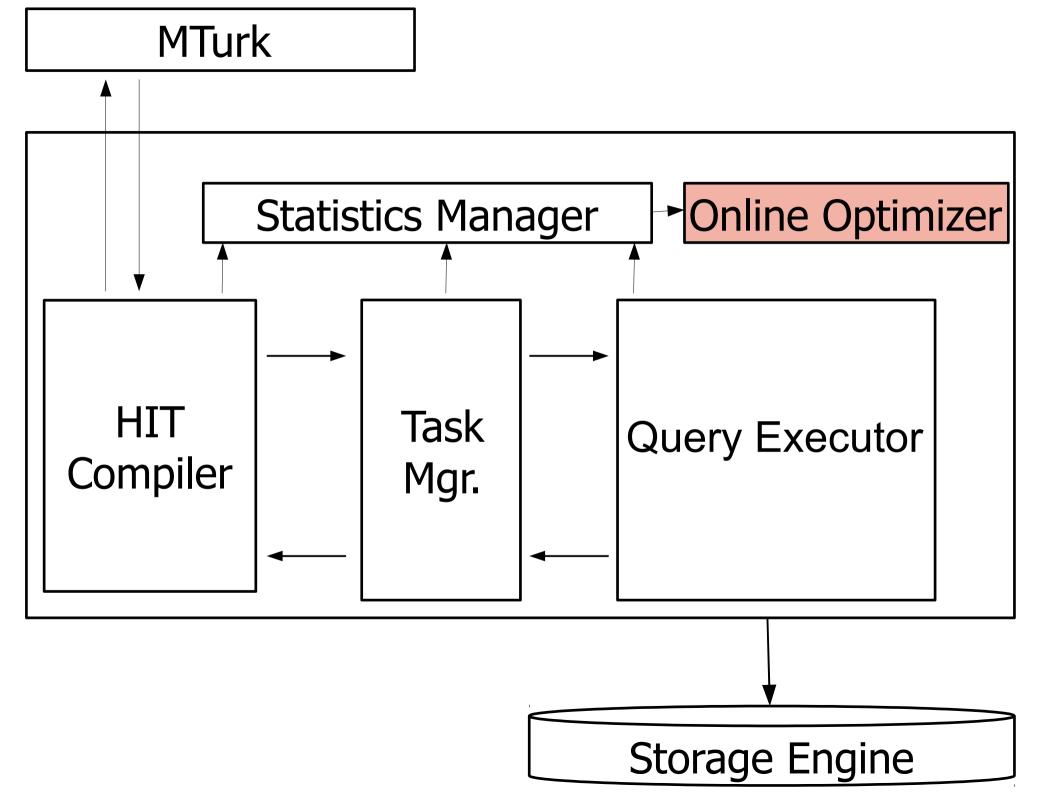
# Early Experimental Result: Skew Correction

- + Improves Turker accuracy
  - Increases number of HITs



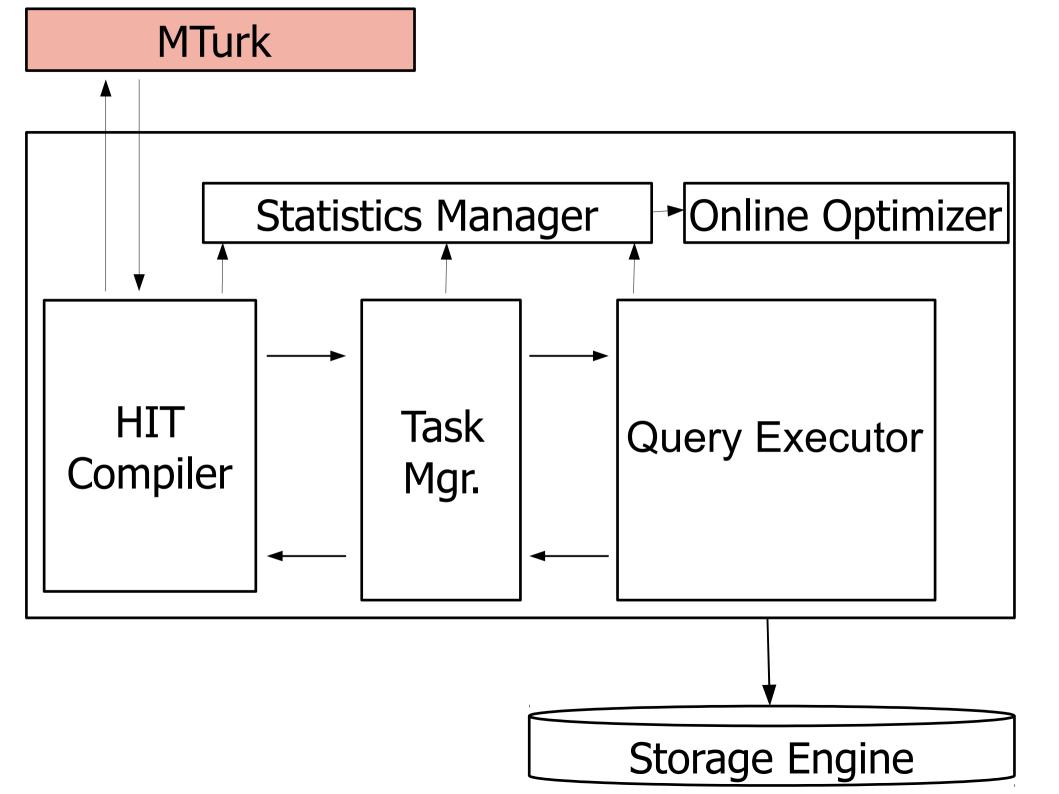






### Online optimizations

- Batch size per HIT
- Price per HIT
  - Small effect on accuracy
  - Large effect on latency
- Turkers per HIT



### **Human Computation Platforms**

- Paid Crowd (e.g., MTurk)
- Experts (e.g., Aardvark)
- Games (e.g., Image Labeling)

### Qurk

- Human computation inside relational databases
- Data model: SQL + Lists
- Asynchronous system design
- Lots of optimization opportunities

### Qurk

- Human computation inside relational databases
- Data model: SQL + Lists
- Asynchronous system design
- Lots of optimization opportunities
- Probably still can't con your way into CIDR

ask us for a demo! marcua@csail.mit.edu/eugenewu@mit.edu

### Image credits

 Clock: http://www.cs.wichita.edu/~vnambood/research.ht

 CAPTCHA: http://en.wikipedia.org/wiki/CAPTCHA

### **Statistics**

- ~80k HITs available at any time
- ~\$8K worth of work at any time
- ~1.2K projects at any time
- Source: http://www.mturk-tracker.com/general/

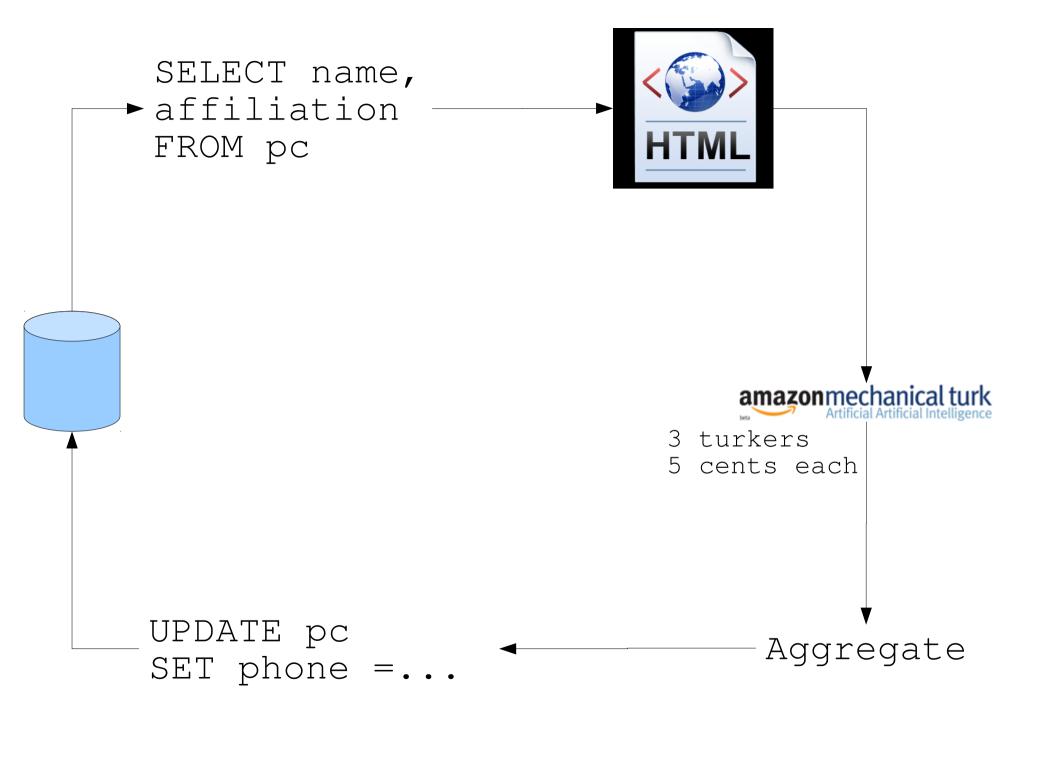
### Other Human Computation Platforms

- Outsourced insurance claims processing
- Data journalism (Guardian)
- CAPTCHA ( following finding )
- Games with a purpose (image labeling)

### Operator Implementations

 Non-blocking to allow pipelining (e.g., symmetric hash join)

- Join batching
- Rank by comparison, rating, or partial order



SELECT name, affiliation FROM pc



Data processed outside of DB Ad-hoc parameter tuning Primitive uncertainty reasoning Logical plan = Physical plan

UPDATE pc SET phone

Aggregate



**ProPublica Reporting Network** 

#### **ProPublica's Guide to Mechanical Turk**

by Srinivas Rao and Amanda Michel ProPublica, Oct. 15, 2010, 6:09 a.m. Amazon Mechanical Turk – or mTurk – is an online marketplace, set up by the online shopping site Amazon, where anyone can hire workers to complete short, simple tasks

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Our Investigations



About 3 years ago Sean P. Aune 9 P.

ProPublica Reporti

### ProPublica Turk Finds Leads

by Srinivas Rao and Amana ProPublica, Oct. 15, 2010, 6

Amazon Mechanical Tu shopping site Amazon,



3



As we reported previously, Amazon Mechanical Turk service is assisting in the search for adventurer Steve Fossett. Using quickly updated satellite imagery of the suspected crash area, people have been searching 256X256 squares for any sign of the aircraft.

There has now been an update on how the search is going, and it seems there have been some interesting leads. After each report is filed of a possible spotting, Mechanical Turk compares your report with any other reports for the same image. At that point, if the system determines there may be something of interest in the

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About 3 years ago Sean P. Aune 9 🗭

ProPublica Reportii

Steve Fossett Search: Amazon Mechanical

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Amazon Mechanical Tr shopping site Amazon,



### Artificial Artificial Intellig

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Fossett. Using quickly updated satellite 256X256 squares for any sign of the ai

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#### amazonmechanical ti SpeakerText Automates And Crowdsources Video Transcripts (100 Beta Invites)

Erick Schonfeld

Sep 7, 2010

One of the big problems with video on the Web is that other than the title, description and some meta tags, it is mostly invisible to Google and other search engines. One way to make video more SEO-friendly is to add transcriptions, but



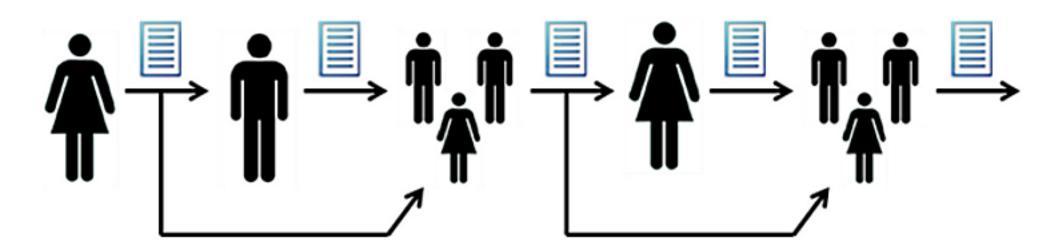
Tweet 4 Digg 1

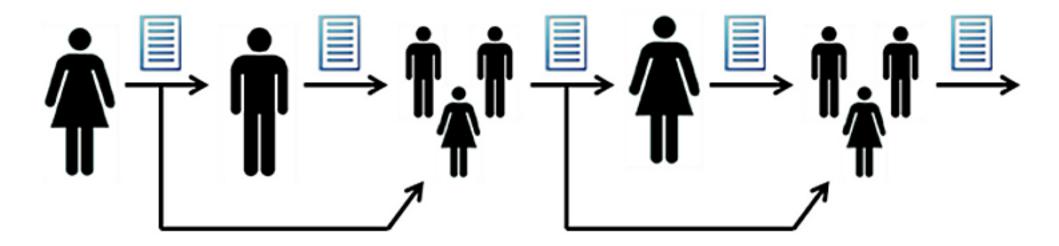
View Comments

that can get expensive. An angel-funded startup called SpeakerText is (re)launching today with a very clever way to automate the transcription process and attach the full transcript as part of the video player in a drop down window. You can see an example of how this works below. And if you publish a lot of videos and want to try it out yourself, we have 100 beta invites (use the code: techcrunch).

Ditch images, make it a list

Once a video is transcribed, it appears in a collapsible window below each player. Not only is





Automatic clustering generally helps separate different kinds of records that need to be edited differently, but it isn't perfect. Sometimes it creates more clusters than needed, because the differences in structure aren't important to the user's particular editing task. For example, if the user only needs to edit near the end of each line, then differences at the start of the line are largely irrelevant, and it isn't necessary to split based on those differences. Conversely, sometimes the clustering isn't fine enough, leaving heterogeneous clusters that must be edited one line at a time. One solution to this problem would be to let the user rearrange the clustering manually, perhaps using drag-and-drop to merge and split clusters. Clustering and selection generalization would also be improved by recognizing common text structure like URLs, filenames, email addresses, dates, times etc.



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Investigate your MP's expenses: Home

Overview

### Investigate your MP's expenses

#### Overview

Leading insurers are through operational

Many insurers have Services, a specializ process optimization and optimize their bu performance.

By using mature plat industry best practice Platform (formerly Na

Join us in digging through the documents of MPs' expenses to identify individual claims, or documents that you think merit further investigation. You can work through your own MP's expenses, or just hit the button below to start reviewing. (Update, Fri pm: we now have a virtually complete set of expenses documents so you should be able to find your MP's) Already created an account? Log in here.

We have 458,832 pages of documents. 27,486 of you have reviewed 222,875 of them. Only 235,957 to go...

Start reviewing

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Wrap this into mechanical turk slide, do as text

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#### Accenture II

Overview Why

#### **Inve**

#### Overview

Leading insurers are through operational

Many insurers have Services, a specializ process optimization and optimize their bu performance.

By using mature plat industry best practic Platform (formerly N:

Join us in digg documents th expenses, or virtually comp created an ac

> We have 4 222,875

Start re



#### reCAPTCHA IS A FREE ANTI-BOT SERVICE THAT HELPS DIGITIZE BOOKS.

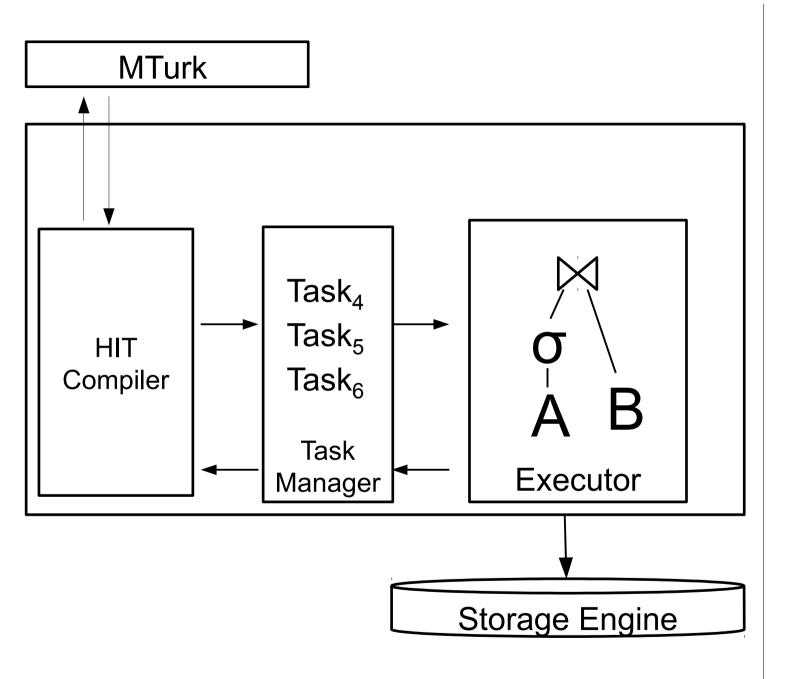


STRONG SECURITY

ACCESSIBLE TO BLIND USERS

30+ MILLION SERVED DAILY

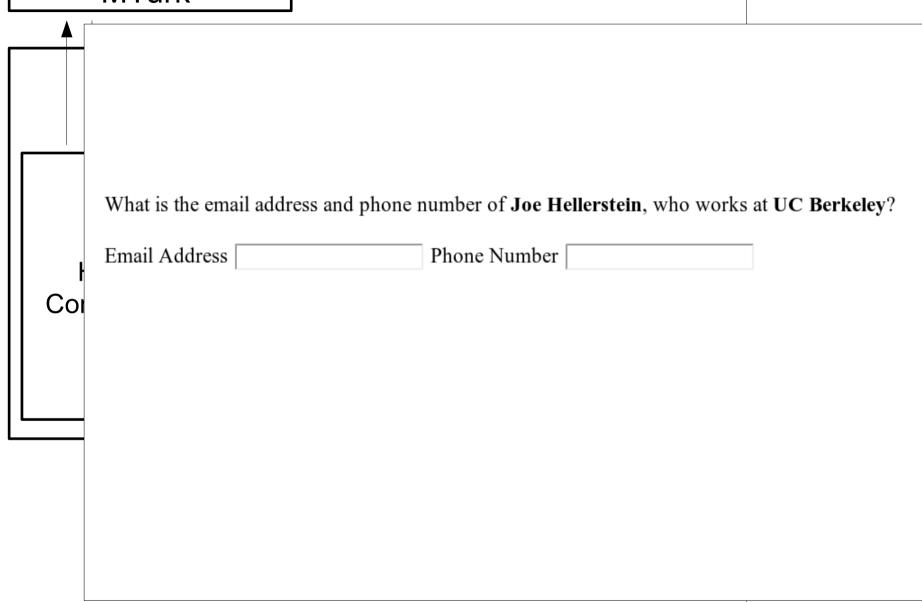
(want more about information integration, data cleaning)



### MTurk Task<sub>4</sub> Task<sub>5</sub> HIT Task<sub>6</sub> Compiler Task **Executor** Manager Storage Engine

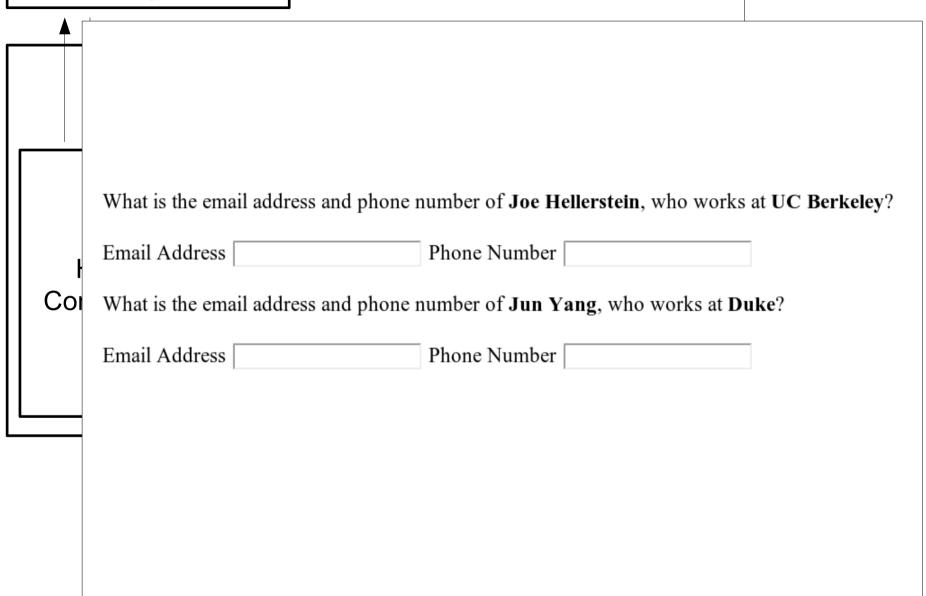
### **Batch Tuples**

### MTurk Batch Tuples



#### MTurk

### **Batch Tuples**



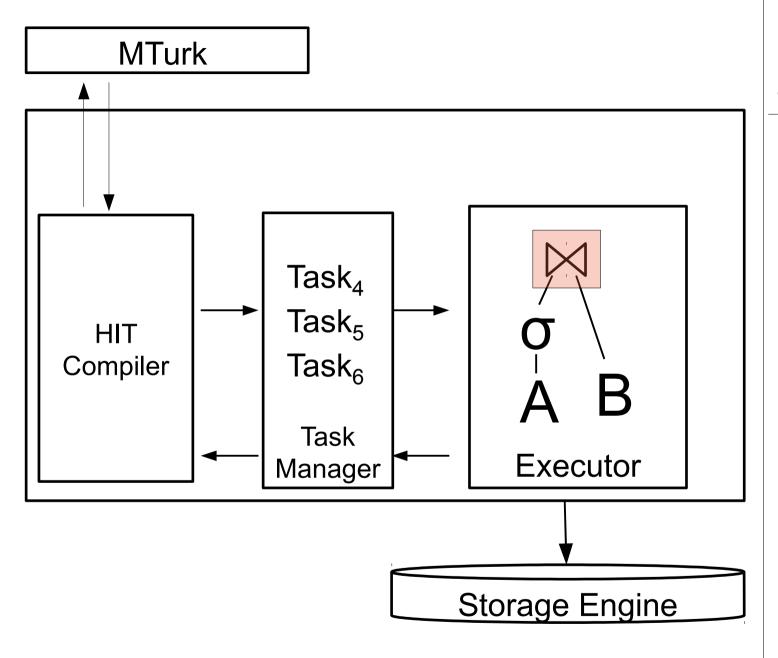
### MTurk Task<sub>4</sub> Task<sub>5</sub> HIT Task<sub>6</sub> Compiler Task **Executor** Manager Storage Engine

## Batch Tuples and Operators

Batch Tuples MTurk male? Coi adult?

**Batch Tuples** MTurk and Operators male? Coi Is this a picture of an adult? • Yes • No adult?

**Batch Tuples** MTurk and Operators male? Coi Is this a picture of an adult? • Yes • No adult? Is this a picture of a male? • Yes • No



Coi

tions

- Non-blocking for pipelining (e.g. symmetric hash join)
- Avoid |A|\*|B| join cost
- Necessary conditions for equality
  - image equi-join: gender, ethnicity match
- Perform |A| + |B| scans to find predicates, remove impossible join pairs

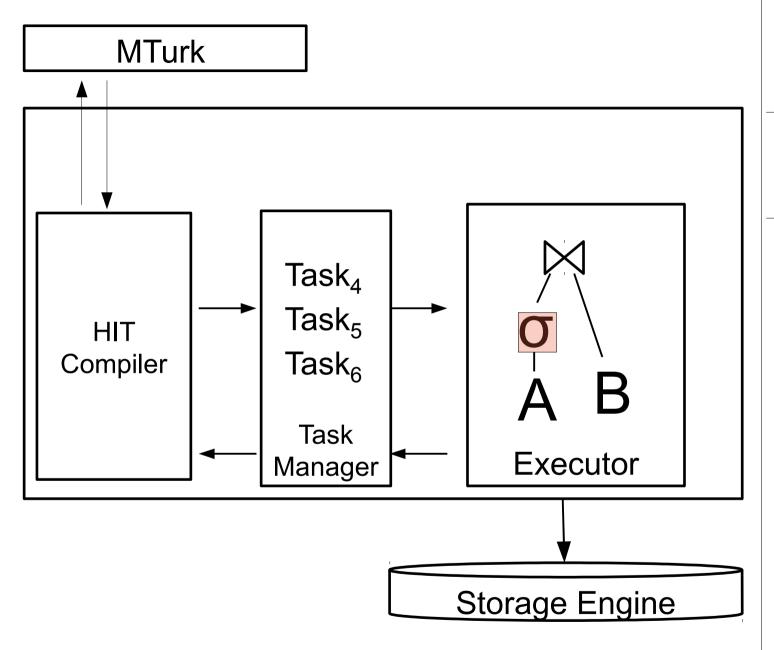
Coi

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  - image equi-join: gender, ethnicity match
- Perform |A| + |B| scans to find predicates, remove impossible join pairs

Alternative: cluster images



Selectivity Injection

#### MTurk

### Batch Tuples and Operators

Ongratar

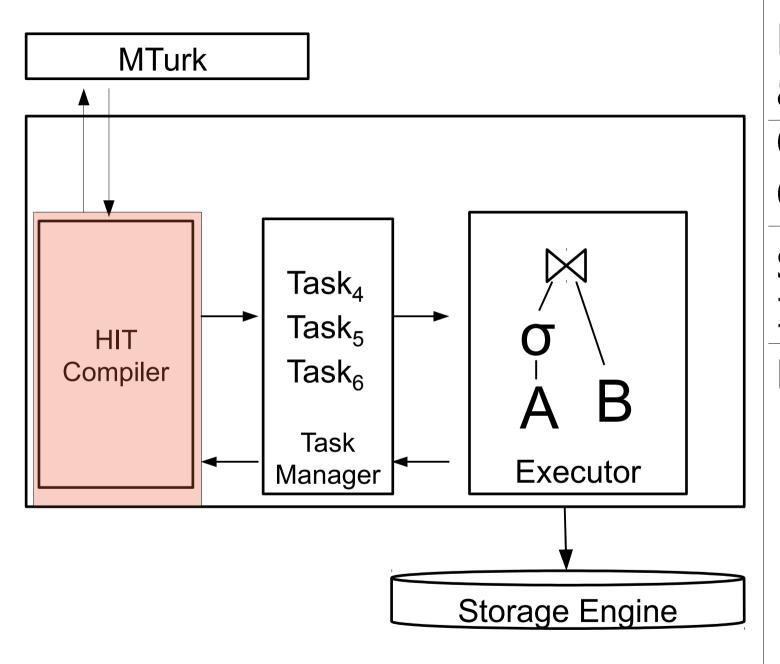
Which string has a 'V' in it?

#### MTurk

## Batch Tuples and Operators

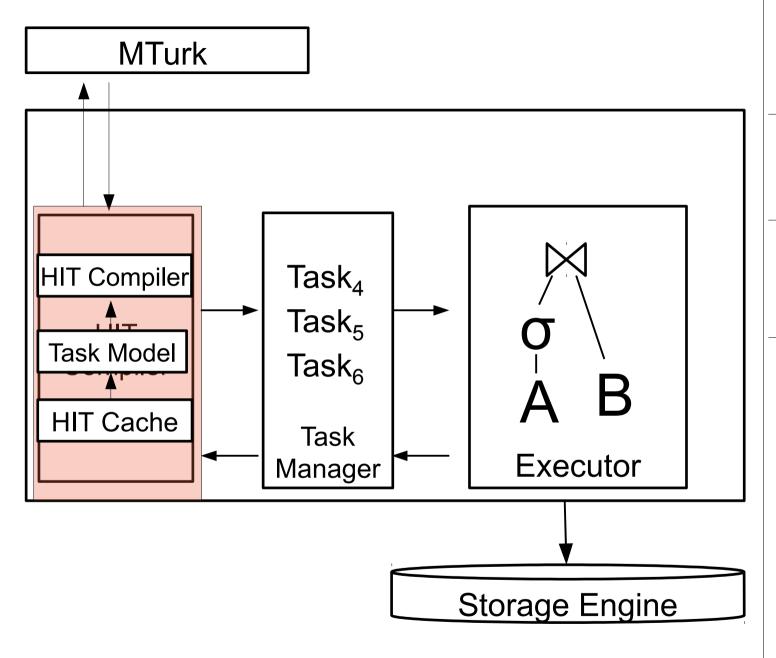
Ongratar

Which string has a 'V' in it?



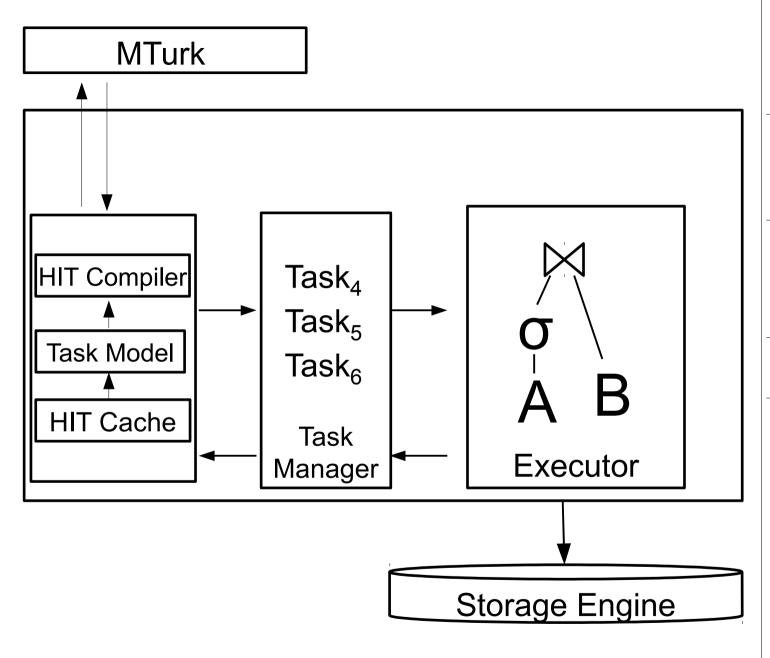
Selectivity Injection

HIT Avoidance



Selectivity Injection

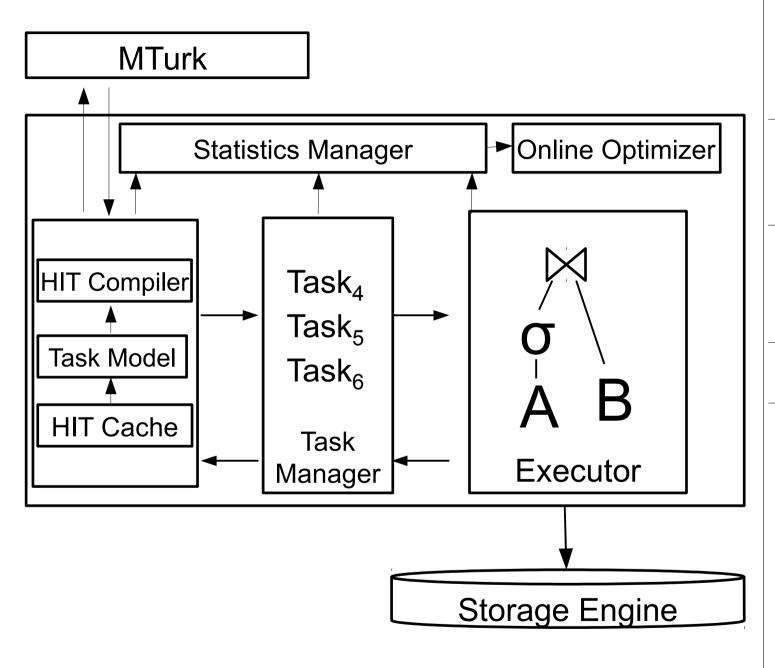
HIT Avoidance



Selectivity Injection

HIT Avoidance

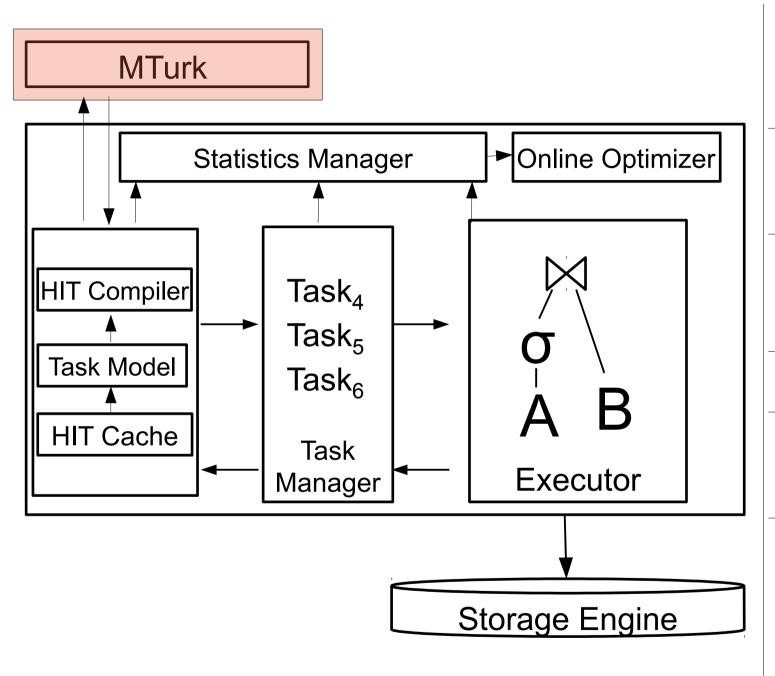
Online Optimizations



Selectivity Injection

HIT Avoidance

Online Optimizations

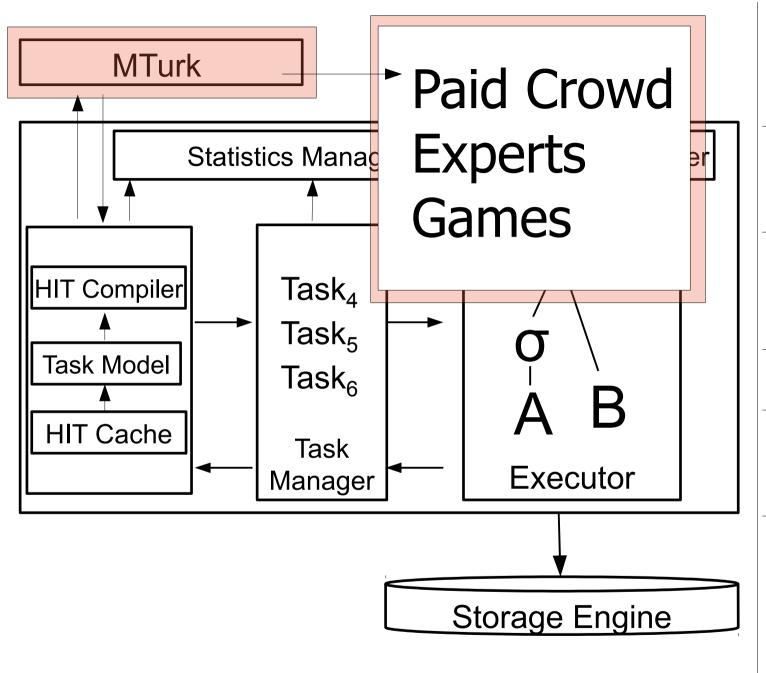


Selectivity Injection

HIT Avoidance

Online Optimizations

Human Computation Platforms



Selectivity Injection

HIT Avoidance

Online Optimizations

Human Computation Platforms