

Human-in-the-Loop Data Management

CMPT 884, FALL 2016

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<https://sfu-db.github.io/cmpt884-fall16>

Introduce Yourself

What's your name?

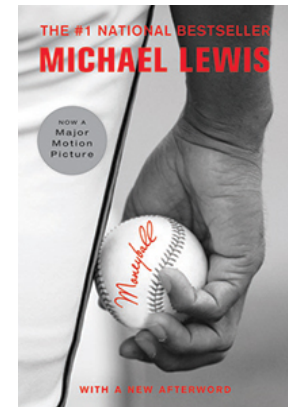
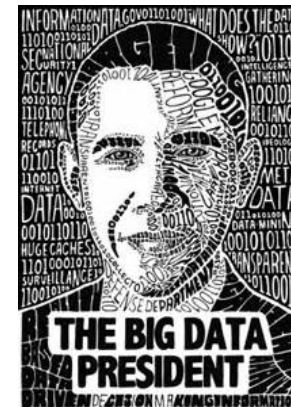
Where are you from?

M.Sc. or Ph.D.? Which year?

What do you want to get out of the course?

A Problem That Everybody Cares About!

How to manage data and extract value from it?



Key Resources

Algorithms

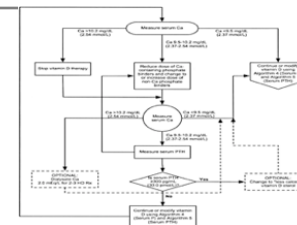
- Machine Learning, Statistical Methods
- Prediction, Business Intelligence

Machines

- Clusters and Clouds
- Warehouse Scale Computing

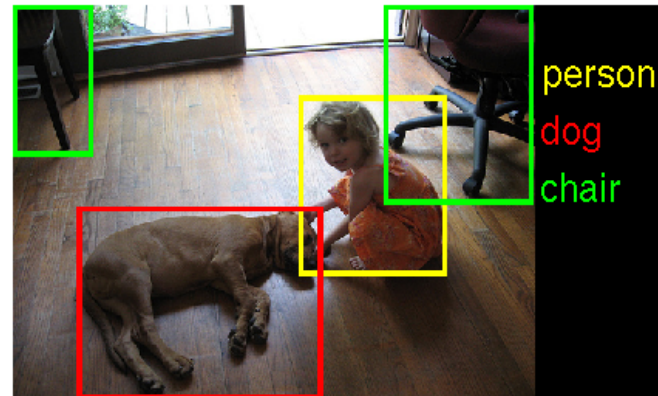
People

- Crowdsourcing, Human Computation
- Data Scientists, Analysts



An Example of Using Three Resources

What are in the image?



How to solve the problem?

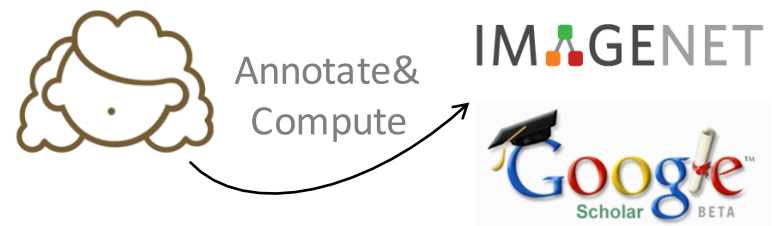
Deep Learning (Algorithms)
GPU Cluster (Machines)
ImageNet (People)

Human-in-the-loop Data Management

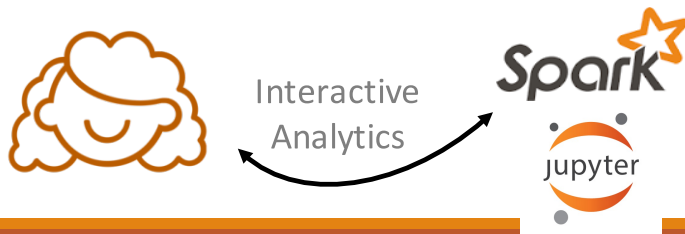
Data Producer



Data Processor



Data Scientist



Data Consumer



A very hot topic

HCOMP 2013

Conference on Human Computation & Crowdsourcing
November 6-9, 2013 - Palm Springs, California USA

The Beckman Database Research Self-Assessment Meeting

2.5 Roles of Humans in the Data Life Cycle

Back when data management was an enterprise-driven activity, it was built databases and database-centric applications, business analysts (data-based) reporting tools, end users generated data and queried and up administrators tuned and monitored databases and their workloads. To

HILDA 2016

Workshop on Human-In-the-Loop Data Analytics

June 26, 2016 | Co-located with **SIGMOD 2016** in San Francisco, CA

Course Objectives

- Introducing students **the cutting-edge research on Human-in-the-loop Data Management**

Part 1: Crowdsourced Data Management

(Human as Data Processor, 13 papers)

Part 2: Interactive Analytics

(Human as Data Scientist, 17 papers)



30 papers

Part 1: Crowdsourced Data Management

(Human as Data Processor, 13 papers)

Machine-based



-  Quality
-  Time
-  Money

Hybrid Human and Machine



-  Quality
-  Time
-  Money

Human-based



-  Quality
-  Time
-  Money

Part 1: Crowdsourced Data Management

(Human as Data Processor, 13 papers)

Systems and Programming Models

1. [CrowdDB: Answering Queries Using Crowdsourcing](#)
2. [TurKit: Human Computation Algorithms on Mechanical Turk](#)
3. [CrowdForge: crowdsourcing complex work](#)

Part 1: Crowdsourced Data Management

(Human as Data Processor, 13 papers)

Quality / Latency Control

4. [Get Another Label? Improving Data Quality and Data Mining Using Multiple, Noisy Labelers](#)
5. [SQUARE: A Benchmark for Research on Computing Crowd Consensus](#)
6. [CLAMShell: Speeding up Crowds for Low-latency Data Labeling](#)

Part 1: Crowdsourced Data Management

(Human as Data Processor, 13 papers)

Data Annotation

7. Labeling images with a computer game
8. ImageNet: A Large-Scale Hierarchical Image Database
9. Cheap and Fast — But is it Good? Evaluating Non-Expert Annotations for Natural Language Tasks

Part 1: Crowdsourced Data Management

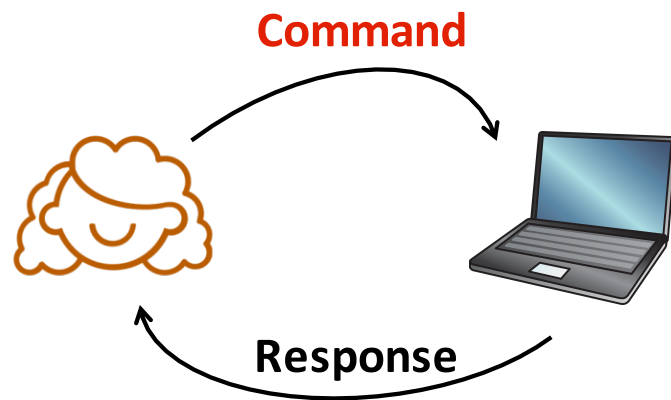
(Human as Data Processor, 13 papers)

Crowdsourced Operators

10. [Human-powered Sorts and Joins](#)
11. [CrowdER: Crowdsourcing Entity Resolution](#)
12. [Leveraging Transitive Relationships for Crowdsourced Joins](#)
13. [Using the crowd for top-k and group-by queries](#)

Part 2: Interactive Analytics

(Human as Data Scientist, 17 papers)



Interactive **Data Cleaning**
Interactive **Visualization**
Interactive **Machine Learning**
Interactive **SQL Analytics**

Part 2: Interactive Analytics

(Human as Data Scientist, 17 papers)

Background

14. Enterprise data analysis and visualization: An interview study
15. The Emerging Role of Data Scientists on Software Development Teams
16. IPython: A System for Interactive Scientific Computing

Part 2: Interactive Analytics

(Human as Data Scientist, 17 papers)

Interactive Data Cleaning

17. [SampleClean: Fast and Accurate Query Processing on Dirty Data](#)
18. [Wrangler: Interactive Visual Specification of Data Transformation Scripts](#)
19. [Scorpion: Explaining Away Outliers in Aggregate Queries](#)

Part 2: Interactive Analytics

(Human as Data Scientist, 17 papers)

Interactive Visualization

20. [Polaris: A System for Query, Analysis, and Visualization of Multidimensional Relational Databases](#)
21. [Prefuse: a toolkit for interactive information visualization](#)
22. [SEEDB: Efficient Data-Driven Visualization Recommendations to Support Visual Analytics](#)
23. [imMens: Real-time Visual Querying of Big Data](#)

Part 2: Interactive Analytics

(Human as Data Scientist, 17 papers)

Interactive Machine Learning

24. [Power to the People: The Role of Humans in Interactive Machine Learning](#)
25. [Active Learning Literature Survey \(Sec 1-4\)](#)
26. [ActiveClean: Interactive Data Cleaning For Statistical Modeling](#)

Part 2: Interactive Analytics

(Human as Data Scientist, 17 papers)

Interactive SQL Analytics

27. [Implementing Data Cubes Efficiently](#)
28. [BlinkDB: queries with bounded errors and bounded response times on very large data](#)
29. [Dremel: Interactive Analysis of Web-Scale Datasets](#)
30. [Spark SQL: Relational Data Processing in Spark](#)

Course Objectives

1. Introducing students **the cutting-edge research on Human-in-the-loop Data Management**
2. Training students to master **basic skills for being a researcher**

Skills

Reading Papers

Giving Talks

Reviewing Papers

Asking Questions

How you will be trained

Reading 27+3 Papers

- A quick scan of 27 papers
- A virtual reimplementations of 3 papers

Giving 1 Talk

- Choosing 1 paper to present (35min+15 min Q&A)

Writing 2 reviews

- One from Part 1 and the other from Part 2

Asking 10 Questions

- Asking at least 10 questions in the Q&A sessions

Grading

Paper Presentation: 25%

Questions: 10%

Paper Review: 15%

Assignments: 15%

Final Project: 35% (5% proposal + 10% presentation + 20% report)

What's next

Fill in the form by the end of Sunday 9/11

<https://goo.gl/forms/FPExVnosd00CCpDj1>