

Human-in-the-Loop Data Management

CMPT 884, FALL 2016

JIANNAN WANG

Introduce Yourself

What's your name?

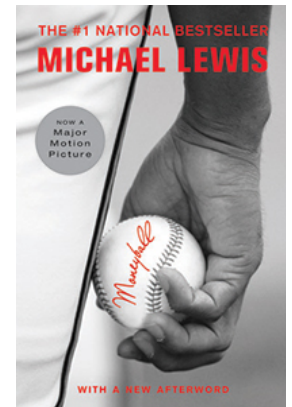
Where are you from?

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

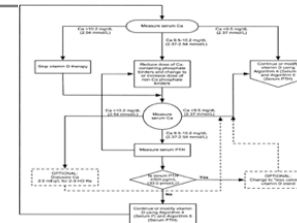
- Machine Learning, Statistical Methods
- Prediction, Business Intelligence

Machines

- Clusters and Clouds
- Warehouse Scale Computing

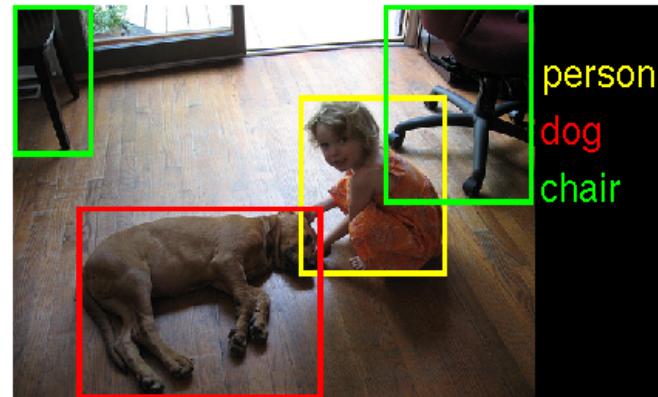
People

- Crowdsourcing, Human Computation
- Data Scientists, Analysts



An Example of Using Three Resources

What are in the image?



How to solve the problem?

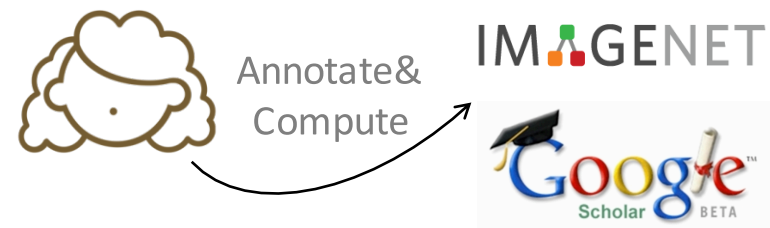
Deep Learning (Algorithms)
GPU Cluster (Machines)
ImageNet (People)

Human-in-the-loop Data Management

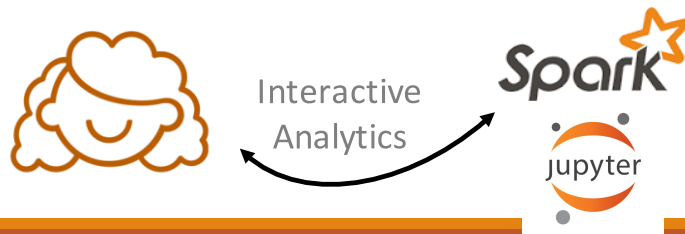
Data Producer



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



Hybrid Human and Machine



Human-based



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

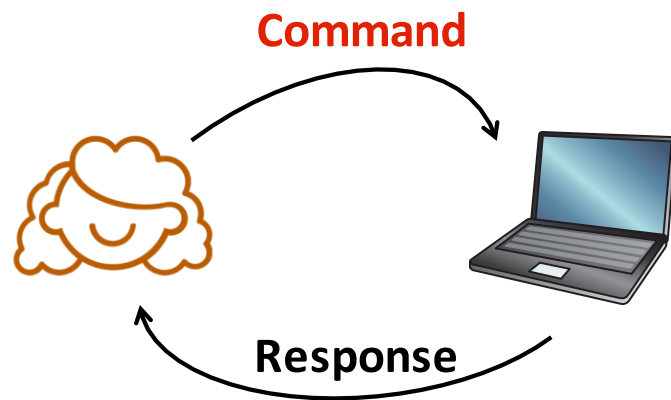
(Human as Data Processor, 13 papers)

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>