Slides Available here: www.gianlucademartini.net/crowdsourcing/searchsolutions

Crowdsourcing for Data Processing and Search

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Gianluca Demartini

- BSc, MSc at U. of Udine, Italy
- PhD at U. of Hannover, Germany
 - Entity Retrieval

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- Worked at the eXascale Infolab U. Fribourg (Switzerland), UC Berkeley (on Crowdsourcing), Yahoo! (Spain), L3S Research Center (Germany)
- Lecturer in Data Science at the iSchool, U. of Sheffield
- Tutorials on Entity Search at ECIR 2012 and RuSSIR 2015, on Crowdsourcing at ESWC 13, ISWC 13, SearchSolutions 2015

www.gianlucademartini.net

Research Interests

- Entity-centric Information Access (2005-now)
 - Structured/Unstruct data (SIGIR 12), TRank (ISWC 13)
 - NER in Scientific Literature(WWW 14) Prepositions (CIKM 14)
- Hybrid Human-Machine Systems (2012-now)
 - ZenCrowd (WWW 12, VLDBJ), CrowdQ (CIDR 13)
 - Memory-based Information Systems (WWW 14, PVLDB)
- Better Crowdsourcing Platforms (2013-now)
 - Pick-a-Crowd (WWW 13), Malicious Workers (CHI 15)
 - Scale-up Crowdsourcing (HCOMP 14), Dynamics (WWW 15)

Learning Objectives

- Demonstrate an understanding of crowdsourcing applications to search problems with its opportunities as well as its limitations;
- Demonstrate knowledge of the common techniques to be used in crowdsourced task design to improve the quality of the collected data;
- Discuss how crowdsourcing can be leveraged in combination with machine-based algorithms for data processing problems and to answer complex search queries;
- Discuss the benefits and challenges of applying crowdsourcing solutions for **search within the enterprise**.

Slides Available here:

www.gianlucademartini.net/crowdsourcing/searchsolutions

Introductions

- Name, role
- Interest / experience in Crowdsourcing / Data Processing / Search

Tutorial Outline

- Part 1
 - Introduction to Crowdsourcing (30min)
 - Ensuring Quality in Paid Crowdsourcing (60min)
- Part 2
 - Hybrid Human-Machine Data Integration (30min)
 - Crowd-Powered Search (30min)
 - Enterprise Crowdsourcing for Search (30min)

Introduction to Crowdsourcing

Crowdsourcing

- Portmanteau of "crowd" and "outsourcing," first coined by Jeff Howe in a June 2006 Wired magazine article
- [Merriam-Webster] the practice of obtaining needed services, ideas, or content by soliciting contributions from a large group of people and especially from the online community rather than from traditional employees or suppliers



Crowdsourcing

- Leverage human intelligence at scale to solve
 - Tasks simple for humans, complex for machines
 - With a large number of humans (the Crowd)
 - Small problems: micro-tasks (Amazon MTurk)
- Examples
 - Wikipedia, Image tagging
- Incentives

- Financial, fun, visibility

• See also longer tutorial at ISWC 2013



Crowdsourcing Incentives

- Paid Crowdsourcing
- Fun (enjoyment)
 Gamification
- Community (belonging, desire to help)
 For example, Wikipedia

The Way Industry Looks At It

CROWDSOURCING LANDSCAPE Beta v1



Case-Study: Amazon MTurk

- Micro-task crowdsourcing marketplace
- On-demand, scalable, real-time workforce
- Online since 2005 (still in "beta")
- Currently the most popular platform
- Developer's API as well as GUI

Amazon MTurk



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. Register Now

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



Amazon MTurk

- Requesters create tasks (HITs)
- The platform takes a fee (30% of the reward)
- Workers preview, accept, submit HITs
- Requesters approve, download results

• If the results are approved, workers are paid

Demographics of MTurk workers in 2009

y of residence

Country of residence

- United States: 46.80%
- India: 34.00%
- Miscellaneous: 19.20%

2013 Statistics: 1M workers 10% active



Gender Breakdown



Demographics of MTurk workers in 2009



Household Income for US workers



Household Income for Indian workers

Year of Birth for US workers



Year of Birth for Indian workers



Demographics of MTurk workers in 2009



Number of HITs completed per week

Time spent on Mechanical Turk per week







For bugs reports or feature requests, please contact Panos Ipeirotis

If you want to cite this website, please cite the paper Analyzing the Amazon Mechanical Turk Marketplace, P. Ipeirotis, ACM XRDS, Vol 17, Issue 2, Winter 2010, pp 16-21.

5-year Analysis of MTurk workload

- Mturk-tracker.com
 - Collects metadata about each visible batch (Title, description, rewards, required qualifications, HITs available, etc), that is, set of similar tasks or HITs
 - Records batch progress (every ~20 minutes)
 - Covers 130M tasks



amazon

Djellel Eddine Difallah, Michele Catasta, Gianluca Demartini, Panagiotis G. Ipeirotis, and Philippe Cudré-Mauroux. **The Dynamics of Micro-Task Crowdsourcing -- The Case of Amazon MTurk**. In: 24th International Conference on World Wide Web (**WWW 2015**), Research Track. Firenze, Italy, May 2015.

Country-Specific HITs



Workers from US, India and Canada are the most sought after.

Distribution of Batch Size



"Power-law"

Batch Size over time



How much are HITs paid?





5-cents is the new 1-cent

Requesters and Reward over time



Increasing number of New and Distinct Requesters

One month of MTurk Requesters

Top-1000 Requesters, report for October 25, 2015 to November 24, 2015

	Requester name	÷	hits	÷	rew
Speechpad			32114		\$288,834.35
CastingWords			11727		\$6,817.26
Chris Callison-Burch			18812		\$5,597.21
p9r			76873		\$4,239.22
Stanford GSB Behavioral Lab			3262		\$2,579.85
Jon Brelig			46459		\$2,483.66
Farhan Memon			9177		\$1,835.40
OCMP5			33243		\$1,651.25
nada hashmi			457		\$1,623.00
VidAngel			126		\$1,583.80

Top requesters



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1 WEEK FTTT \$1.50/minute 1 DAY \$2.50/minute

INTERNATIONAL \$1.75/minute 27



Watch Movies Your Way - However the **BLEEP** you want!



Distribution of HIT Types

Year 2009 2010 2011 2012 2013 2014 10⁵ 10[°] 10[°] 10[°] 10[°] 10[°] 10⁰ CC SU VV CA

Task Category

10¹

Less Content Access batches

Content Creation: the most popular

Classify HITs into types (Gadiraju et. al 2014)

- Information Finding (IF)
- Verification and Validation (VV)
- Interpretation and Analysis (IA)
- Content Creation (CC)
- Surveys (SU)
- Content Access (CA)

Is the Market Elastic?



Intercept = 2.5Slope = 0.5%

20% of new work gets completed within an hour

Summary

- HIT reward has increased over time
- Audio transcription: the most popular task
- Demand for Indian workers has decreased
- Surveys are most popular for US workers
- 1000 new requesters per month join
- 10K new HITs arrive and 7.5K HITs get completed every hour
- Check #mturkdynamics for more findings

Why Crowdsourcing for IR Evaluation?

- Easy, cheap and fast labeling
- Ready-to use infrastructure MTurk payments, workforce, interface widgets – CrowdFlower quality control mechanisms, etc.
- Allows early, iterative, frequent experiments Iteratively prototype and test new ideas – Try new tasks, test when you want & as you go
- Proven in major IR shared task evaluations
 CLEF image, TREC, INEX, WWW/Yahoo SemSearch

Gamification of IR Evaluation

GeAnn: <u>http://www.geann.org/</u>

- Relevance judgments with Gamification:
 - Text relevance
 - Image relevance

Quality through Flow and Immersion: **Gamifying Crowdsourced Relevance Assessments**. Eickhoff, C., C. G. Harris, A. P. de Vries, and P. Srinivasan. SIGIR 2012.

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Ensuring Quality in Paid Crowdsourcing

A Crowdsourcing Task

Choose the best category for this image



View Instructions 4

Select the room location in home for this picture. Seating areas outside are outside not living. Offices or dens are living not bedrooms. Bedrooms should contain a bed in the picture.

kitchen
living
bath
bed
outside
High-level Issues in Crowdsourcing

- Process
 - Experimental design, annotation guidelines, iteration
- Choose crowdsourcing platform (or roll your own!)
- Human factors
 - Payment / incentives, interface and interaction design, communication, reputation, recruitment, retention
- Quality Control / Data Quality
 - Trust, reliability, spam detection, consensus labeling

Task Design

- Ask the right questions
- Workers may not be experts so don't assume the same understanding in terms of terminology
- Instructions matter!
- Show examples
- Hire a technical writer
 - Engineer writes the specification
 - Writer communicates

Task Design - UI

- Generic tips
 - Experiment should be self-contained.
 - Keep it short and simple. Brief and concise.
 - Be very clear with the task.
 - Engage with the worker. Avoid boring stuff.
 - Always ask for feedback (open-ended question) in an input box.

Bad Example

- Asking too much, task not clear, "do NOT/reject"
- Worker has to do a lot of stuff

Help us describe How-To Videos! Earn \$2.50 bonus for every 25 videos entered!

Watch a how-to video, and writs a keyword-friendly synopsis describing the video.

- 1. Click on the link to watch the Film & Theater how-to video =>> 332492 Get a 35mm film look with a depth of field adapter
- 2. Write a description of the video linked in 4 or more sentences.
- 3. Be detailed in your description. Describe how the procedure is done.
- 4. Description should be at least 100 words.
- 5. Description should be fewer than 2000 characters.
- 6. Use the character and word counters below to help you stay within the limits.
- 7. You must complete 25 video descriptions in order to earn the \$2.50 bonus. Bonuses are distributed after HITs have been completed. The more HITs completed and approved, the more you will earn.
- 8. It is not necessary to repeat the headline in your entry. It will NOT count toward your word count.
- 9. Do NOT describe the following: the format, where the video comes from, or how long the video is. This information is IRRELEVANT.
- 10. Do NOT describe the video in the following manner. "She turns around to face the camera. Then she faces left." Follow the examples below.

Current Word Count: 0 Current Character Count: 0 / 2000

Criteria for REJECTION:

- 1. Entries with obvious and stultiple spelling or grammatical errors will be rejected.
- 2. Entries with fewer than 100 words will be automatically rejected.
- 3. Test copied from the web or other places will be rejected. Multiple plagianzed answers will lead to being BLOCKED. You may use a quotation, but the majority of your content must be ORIGINAL.
- 4. Incomplete and blank answers will be rejected. Multiple blank answers will result in being blocked.
- 5. Tasks submitted without descriptions will be rejected.
- 6. Tasks submitted with inaccurate descriptions will be rejected as well.
- 7. Do NOT add any personal opinions. Entries with personal opinions or reviews will be automatically REJECTED.
- 8. If you notify us that a link is broken, we appreciate it but will not be able to accept the submission. The notification will result in rejection.
- 9. Entries that transcribe the video will be REJECTED.

Good Example

- All information is available
 - What to do
 - Search result
 - Question to answer

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ILTON KEYNES	ALL RESULTS 1-	20 of 7,020,000 results - <u>Advanced</u> Sponsored sit	es
ilton Keynes Map	Milton Keynes - Wikipedia, the free encyclopedia	Milton Key	nes Hotels
filton Keynes l <mark>estaurants</mark>	Milton Keynes, often abbreviated MK, is a large town in Bucking England, about 45 miles (72 km) north-west of London. It is also t History · Urban design · Culture · Education	namshire, in the south east of Save up to 50 he capital of Our Best Price www.expedia	% on Hotels and Now Get e Guarantee. com +

Form and Metadata

- Form with a close question (binary relevance) and open-ended question (user feedback)
- Clear title, useful keywords
- Workers need to find your task

Describe you	ir HIT
Title	Pick the best category
	Describe the task to workers. Be as specific as possible, e.g. "answer a survey about movies", instead of "short survey", so workers know what to expect.
Description	Pick the best category
	Give more detail about this task. This gives workers a bit more information before they decide to view your HIT.
Keywords	category, categorize
	Provide keywords that will help workers search for your HITs.

How Much to Pay?

- Price commensurate with task effort
 - Ex: \$0.02 for yes/no answer + \$0.02 bonus for optional feedback
- Ethics & market-factors
 - e.g. non-profit SamaSource contracts workers refugee camps
- Uptake & time-to-completion vs. Cost & Quality
 - Too little \$\$, no interest or slow
 - too much \$\$, attract spammers
- Accuracy & quantity
 - More pay = more work, not better (W. Mason and D. Watts, 2009)

Quality Control

- Extremely important part of the experiment
- Approach as "overall" quality; not just for workers
- Bi-directional channel
 - You may think the worker is doing a bad job.
 - The same worker may think you are a lousy requester.

Quality Control

- Approval rate: easy to use, & just as easily defeated
- Mechanical Turk Masters
 - Recent addition, only for specific tasks
- Qualification test
 - Pre-screen workers' ability to do the task (accurately)
- Assess worker quality as you go
 - Trap questions with known answers ("honey pots")
 - Measure inner-annotator agreement between workers

Qualification tests: pros and cons

- Advantages
 - Great tool for controlling quality
 - Adjust passing grade
- Disadvantages
 - Extra cost to design and implement the test
 - May turn off workers, hurt completion time
 - Refresh the test on a regular basis
 - Hard to verify subjective tasks like judging relevance
- Try creating task-related questions to get worker familiar with task *before* starting task in earnest

Other quality heuristics

- Justification/feedback as quasi-captcha
 - Should be optional
 - Automatically verifying feedback was written by a person may be difficult (classic spam detection task)
- Broken URL/incorrect object
 - Leave an outlier in the data set
 - Workers will tell you
 - If somebody answers "excellent" for a broken URL => probably spammer

Dealing with bad workers

- Pay for "bad" work instead of rejecting it?
 - Pro: preserve reputation, admit if poor design at fault
 - Con: promote fraud, undermine approval rating system
- Use bonus as incentive
 - Pay the minimum \$0.01 and \$0.01 for bonus
 - Better than rejecting a \$0.02 task
- If spammer "caught", block from future tasks
 - May be easier to always pay, then block as needed

Build Your Reputation as a Requestor

- Word of mouth effect
 - Workers trust the requester (pay on time, clear explanation if there is a rejection)
 - Experiments tend to go faster
 - Announce forthcoming tasks (e.g. tweet)

Crowd Worker Communities



Powered by non-amazonian script monkeys

To many bubbles but YMMV with your patience level.



Djellel Eddine Difallah, Gianluca Demartini, and Philippe Cudré-Mauroux. Pick-A-Crowd: Tell Me What You Like, and I'll Tell You What to Do. In: WWW2013

51

Behavioral Patterns of Malicious Workers



Ujwal Gadiraju, Ricardo Kawase, Stefan Dietze, and Gianluca Demartini. **Understanding Malicious Behaviour in Crowdsourcing Platforms: The Case of Online Surveys**. In: Proceedings of the ACM Special Interest Group on Computer Human Interaction (**CHI 2015**).

OpenTurk.com

- Yet another a platform? Build on top of Mturk!
- Chrome Extension for push / notification
- 400+ users
- <u>http://bit.ly/openturk-extension</u>
- Open source:

https://github.com/openturk/extension

Majority Vote

- Ask N workers and pick the most popular answer
- Works for multiple-choice questions
 - Relevance judgments
 - Sentiment analysis / supervised machine learning
- For other task use iterations
 - Audio transcription
 - Ask one worker to transcribe, the next to correct, etc.
- Learning weights for workers

Entity Factor Graphs



2 workers, 6 clicks, 3 candidate links

Aggregation based on worker similarity

- "Community-Based Bayesian Aggregation Models for Crowdsourcing", Venanzi et al., WWW2014.
- Community-based Bayesian aggregation model
- Group workers by the type of errors they do





SQUARE

- A benchmark for crowd answer aggregation
 - Binary choices (e.g., sentiment)
 - Multiple-choices (e.g., relevance, word-sense disambiguation)
- Compares a number of aggregation techniques over a number of tasks

http://ir.ischool.utexas.edu/square/

Other benchmarks

- Simulations
 - BATC A Benchmark for Aggregation Techniques in Crowdsourcing
 - Understand effect on efficiency and effectiveness
 - Set aggregation parameters

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Hybrid Human-Machine Data Integration

Example: Hybrid Data Integration

paper	conf		
Data integration	VLDB-01		
Data mining	SIGMOD-02		

title	author	email	venue	
OLAP	Mike	mike@a	ICDE-02	
Social media	Jane	jane@b	PODS-05	

Generate plausible matches

- paper = title, paper = author, paper = email, paper = venue
- conf = title, conf = author, conf = email, conf = venue

Ask users to verify

Does attribute paper match attribute author?								
paper	conf		title	author	email			
Data integration	VLDB-01		OLAP	Mike	mike@a			
Data mining	SIGMOD-02		Social media	Jane	jane@b			
Yes No Not sure								

McCann, Shen, Doan: Matching Schemas in Online Communities. ICDE, 2008

Example: Hybrid Query Processing

Use the crowd to answer DB-hard queries

Where to use the crowd:

- Find missing data
- Make subjective comparisons
- Recognize patterns

But not:

 Anything the computer already does well



M. Franklin, D. Kossmann, T. Kraska, S. Ramesh and R. Xin . CrowdDB: Answering Queries with Crowdsourcing, *SIGMOD 2011* ⁶²

Facebook Buys Instagram for \$1 Billion

BY EVELYN M. RUSLI

2:02 p.m. | Updated

Facebook is not waiting for its initial public offering to make its first big purchase.

In its largest acquisition to date, the social network has purchased Instagram the popular photo-sharing application, for about \$1 billion in cash and stock, the company said Monday

HTML:

Facebook is not waiting for its initial public offering to make its first big purchase.In its largest acquisition to date, the social network has purchased Instagram, the popular photo-sharing application, for about \$1 billion in cash and stock, the company said Monday.

RDFa enrichment

<cite property="rdfs:label">Facebook</ cite> is not waiting for its initial public offering to make its first big purchase.In its largest acquisition to date, the social network has purchased <cite property="rdfs:label">Instagram">In its largest acquisition to date, the social network has purchased <cite property="rdfs:label">Instagram cite> , the popular photo-sharing application, for about \$1 billion in cash and stock, the company said Monday.

http://dbpedia.org/resource/Instagram owl:sameAs fbase:Instagram Google CNET > New Mobi Android Instagram for Android is now available At long last, Instagram finally releases the Android version of its app. by Jason Cipriani | April 3, 2012 10:07 AM PDT Instagram has been around since 2010, available only to iOS devices. Android users have been waiting patiently, with repeated promises of an Android version arriving soon.

http://dbpedia.org/resource/Facebook

ZenCrowd

- Combine both algorithmic and manual linking
- Automate manual linking via crowdsourcing
- Dynamically assess human workers with a probabilistic reasoning framework



ZenCrowd Architecture



Gianluca Demartini, Djellel Eddine Difallah, and Philippe Cudré-Mauroux. **ZenCrowd: Leveraging Probabilistic Reasoning and Crowdsourcing Techniques for Large-Scale Entity Linking**. In: 21st International Conference on World Wide Web (**WWW 2012**).

Entity Factor Graphs

- Graph components
 - Workers, links, clicks
 - Prior probabilities
 - Link Factors
 - Constraints
- Probabilistic Inference
 - Select all links with posterior prob >τ



2 workers, 6 clicks, 3 candidate links

Lessons Learnt

- Crowdsourcing + Prob reasoning works!
- But
 - Different worker communities perform differently
 - Many low quality workers
 - Completion time may vary (based on reward)
- Need to find the right workers for your task (see WWW13 paper)

ZenCrowd Summary

- ZenCrowd: Probabilistic reasoning over automatic and crowdsourcing methods for entity linking
- Standard crowdsourcing improves 6% over automatic
- 4% 35% improvement over standard crowdsourcing
- 14% average improvement over automatic approaches

http://exascale.info/zencrowd/

- Follow up-work (VLDBJ):
 - Also used for instance matching across datasets
 - 3-way blocking with the crowd

ZenCrowd Architecture



Gianluca Demartini, Djellel Eddine Difallah, and Philippe Cudré-Mauroux. **ZenCrowd:** Leveraging Probabilistic Reasoning and Crowdsourcing Techniques for Large-Scale Entity Linking. In: 21st International Conference on World Wide Web (WWW 2012)

Blocking for Instance Matching

- Find the instances about the same real-world entity within two datasets
- Avoid Comparison of all possible pairs
 - Step 1: cluster similar items using a cheap similarity measure
 - Step 2: n*n comparison within the clusters with an expensive measure

3-steps Blocking with the Crowd

• Crowdsourcing as the most expensive similarity measure



tamr.com

• Data Integration solutions: algorithms+experts

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			Margaret Roberts		motorcycle engine	Purpuro-Cole Bros.	3.22+g	Sedan Europe	524	Engines, Motor
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http://www.tamr.com/oreilly-webinar-taming-data-variety/
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Crowd-Powered Search

Slow Search

- "Not All Searches Need to Be Fast"
 - Planning a vacation
 - Medical diagnosis
- Use additional time for human computation

Jaime Teevan. "Slow Search: Improving Information Retrieval Using Human Assistance", CIKM 2015.

Jaime Teevan, Kevyn Collins-Thompson, Ryen W White, and Susan Dumais. "SlowSearch". CACM, 57-8, Aug 2014.

Crowd-powered Search

- Search process
 - Understand query
 - Retrieve
 - Understand results
- Machines are good at operating at scale
- People are good at understanding

Extract Direct Answers w/ Crowdsourcing

	dog	temperatur	2						
	Norma	al Body Temper							
	The nor of 102°F	The normal dog body temperature is 101.5°F (38.6°C). A body temperature of 102°F (38.9°C) or above is considered a fever.							
	Source:	http://www.natural-d	og-health-remedi	es.com/dog-temp	erature.html				
	How to	Take Your Dog	<u>'s Temperat</u>	ure - Page 1	at ha				
	when your dog is ill, you may have to determine whether or not he has a fever by taking your dog's temperature . It's relatively easy								
Identify Cand	idates —		Filter Canc	lidates——					
Search Trails 75m trails, 13m URLs	Answer Trails w/ click	Candidates kthroughs to same URL	Destination 140,000 URLs wit	Probability th high dest. prob.	Question Querie 19,000 URLs w/ question	Answer Type			
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Extract Answe	er Conten	t							
Extract	Vote	Proofread	Vote	Title	Vote	Tail Answer			
3 info extractors	5 voters	3 improvements 101.5 deg°	3 voters	<i>3 title authors</i> Average Dog T	emp. Jog Temp	Average Dog Temp. 101.5 °F			

Bernstein et al., Direct Answers for Search Queries in the Long Tail, CHI 2012.

birthdate of the mayor of the capital city of italy

birthdate of the mayor of the capital city of italy						Ŷ	۹	
Web	Shopping	News	Images	Maps	More -	Search tools		

About 3,830,000 results (0.46 seconds)

Asmara - Wikipedia, the free encyclopedia

en.wikipedia.org/wiki/Asmara - Wikipedia -

Jump to Italian Eritrea - ... and when it was occupied by Italy in 1889 and was made the capital city of Eritrea in preference to Massawa by Governor Martini ...

Turin - Wikipedia, the free encyclopedia

en.wikipedia.org/wiki/Turin - Wikipedia -

Jump to **City** centre - Via Roma crosses one of the **main** squares of the **city**: the pedestrianised ... senate and, for few years, the **Italian** senate after the **Italian** unification), the ... to Saint John the Baptist, which is the **major** church of the **city**.

Milan - Wikipedia, the free encyclopedia

en.wikipedia.org/wiki/Milan - Wikipedia -

Its business district hosts the Borsa Italiana (Italy's main stock exchange) and the headquarters of the largest national banks and companies. The city is a major ...

Rome - Wikipedia, the free encyclopedia

capital city of italy



Design Addition of the first state of the st

mayor of rome

mayor	of rome						Ŷ	۹
Web	Images	Videos	News	Maps	More -	Search tools		
About 2	2,500,000 re	sults (0.30 s	econds)					

Ignazio Marino

The outgoing Mayor of Rome, Gianni Alemanno (PdL), stood for election for a second term as mayor. The center-left candidate, heart surgeon **Ignazio Marino** was be chosen by a multi-party primary election on 7 April 2013. Control of the 15 municipi of the Italian capital was decided in the election.

Rome municipal election, 2013 - Wikipedia, the free ... https://en.wikipedia.org/wiki/Rome_municipal_election,_2013

birthdate of ignazio marino

birthdate of Ignazio Marino

News

...

Web

Images

Videos Maps

More

More

Search tools

About 1,140,000 results (0.34 seconds)

March 10, 1955 (age 60 years)

Ignazio Marino, Date of birth



Gianni Alemanno March 3, 1958



Nicola Zingaretti October 11, 1965



Salvatore Buzzi November 15, 1955



Surgeon

Ignazio Roberto Maria Ma transplant surgeon and the of Rome. He is a member Democratic Party and hele Italian Senate from 2006 t as mayor of Rome. Wikipe

Born: March 10, 1955 (ag Italy

Education: Catholic Unive Sacred Heart (1979)

Party: Democratic Party

Motivation

• Web Search Engines can answer simple factual queries directly on the result page

Users with complex information needs are often unsatisfied

- Purely automatic techniques are not enough
- We want to solve it with Crowdsourcing!

CrowdQ

- CrowdQ is the first system that uses crowdsourcing to
 - Understand the intended meaning
 - Build a structured query template
 - Answer the query over Linked Open Data

Gianluca Demartini, Beth Trushkowsky, Tim Kraska, and Michael Franklin. **CrowdQ: Crowdsourced Query Understanding**. In: 6th Biennial Conference on Innovative Data Systems Research (**CIDR 2013**). About 124,000,000 results (0.33 seconds)

City	Mayor	Birthdate	
Rome, Italy	Gianni Alemanno	March 3, 1958	
Venice, Italy	Giorgio Orsoni	August 29, 1946	
Milan, Italy	Giuliano Pisapia	May 20, 1949	

Press to see more

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www.italylogue.com/italian-cities

Learn about the best **cities in Italy** to visit, and some **Italian cities** you might never have heard of before. These **cities in Italy** are **all** great for visitors.

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goitaly.about.com/od/planningandinformation/tp/topcities.htm

Italy has many beautiful and historic **cities** that are well worth a visit. Here are our picks for the ten best **cities** for visitors to **Italy**.

Italian Cities and Towns - Italy

en.comuni-**italian**i.it/

Information and statistics on Italian Pagions, Provinces and Municipalities, All Citics,

Q

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CrowdQ Architecture

Off-line: query template generation with the help of the crowd **On-line**: query template matching using NLP and search over open data



Hybrid Human-Machine Pipeline



Structured query generation Q= birthdate of actors of for MOVIE np SELECT ?y ?x WHERE { ?y <dbpedia-owl:birthdate> ?x . ?z <dbpedia-owl:starring> ?y . ?z <rdfs:label> 'F MOVIE mp' }

Results from BTC09:

<http://dbpedia.org/resource/Robin_Wright_Penn> 1966-04-08 <http://dbpedia.org/resource/Tom_Hanks> 1956-07-09 <http://dbpedia.org/resource/Sally_Field> 1946-11-06 <http://dbpedia.org/resource/Gary_Sinise> 1955-03-17 <http://dbpedia.org/resource/Mykelti_Williamson> 1960-03-04

Overview of hybrid systems

Year Cit.		Domain	Data Type	Human	Incentive	Time
			F	role		constrains
2006	[62]	Web	Images	Pre-p.	Fun	Batch
2007	[35]	Science	Images	Pre-p.	Community	Batch
2008	[64]	Web	Images	Post-p.	Access	Batch
2011	[52]	Database	Graph	Pre-p.	Monetary	Batch
2011	[30]	Database	Struct. data	Pre-p.	Monetary	Real-time
2011	[5]	Filtering	Video	Pre-p.	Monetary	Real-time
2012	[54]	Database	Struct. data	Post-p.	Monetary	Real-time
2012	[19]	Web	Unstruct. text	Post-p.	Monetary	Batch
2012	[56]	Data Integration	Struct. data	Post-p.	Monetary	Batch
2012	[66]	Entity Resolution	Struct. data	Post-p.	Monetary	Batch
2012	[68]	Entity Resolution	Struct. data	Post-p.	Monetary	Batch
2012	[8]	Search	Unstruct. text	Post-p.	Community	Real-time
2012	[42]	Captioning	Video	Pre-p.	Community	Real-time
2013	[34]	Info Extraction	Unstruct. text	Post-p.	Monetary	Batch
2013	[20]	Entity Resolution	Struct. data	Post-p.	Monetary	Batch
2013	[67]	Entity Resolution	Struct. data	Post-p.	Monetary	Batch
2013	[21]	Database	Struct. data	Pre-p.	Monetary	Batch
2013	[44]	Database	Struct. data	Post-p.	Monetary	Real-time
2013	[48]	Biomedical	Ontology	Pre-p.	Monetary	Batch
2013	[43]	Personal assistance	Unstruct. text	Pre-p.	Monetary	Real-time
2013	[27]	Biomedical	Unstruct. text	Post-p.	Fun	Batch
2014	[53]	Search	Image	Pre-p.	Monetary	Real-time
2014	[49]	Database	Struct. data	Post-p.	Monetary	Real-time
2014	[51]	Cult. Heritage	Image	Pre-p.	Monetary	Batch

Overview of hybrid systems

- Balance between systems that use the human component as pre-processing or postprocessing of data (11 vs 13)
- Mostly monetary reward
- Majority of systems perform batch data processing rather than real-time jobs
- In 2014 we can observe a decreased number of hybrid human-machine systems being propose : focus on solving core problems rather than building new systems

Summary

- Crowdsourcing big data can make you go bankrupt! -> hybrid systems
- When to ask a human, when to trust the machine
- Hybrid systems (human in the loop)
 - Pre-processing: training data for ML
 - Post-processing: based on confidence scores
 - Mix: active learning

Gianluca Demartini. Hybrid Human-Machine Information Systems: Challenges and Opportunities. In: Computer Networks, Special Issue on Crowdsourcing, Elsevier.

Tutorial Outline

- Part 1
 - Introduction to Crowdsourcing (30min)
 - Ensuring Quality in Paid Crowdsourcing (60min)
- Part 2
 - Hybrid Human-Machine Data Integration (30min)
 - Crowd-Powered Search (30min)
 - Enterprise Crowdsourcing for Search (30min)

Enterprise Crowdsourcing for Search

Enterprise Crowdsourcing

- Internal crowd
 - Employees of the company
 - Full-time annotators
 - Casual crowd workers
- Pro: Trust, Domain Knowledge
- Contra: Incentives, Load-balancing

Crowds for Enterprise Crowdsourcing

- Internal Crowd
 - IBM
 - Microsoft
 - Google
- External Crowd
 - Amazon MTurk
 - Yandex Toloka toloka.yandex.com

Crowds for Enterprise Crowdsourcing

- Mixed info.crowdflower.com/nda-contributors
 - NDA Crowds by Crowdflower
 - Top Tolokers become Yandex employees
- Tamr.com
 - Internal Expert-sourcing for data integration

Use of Crowdsourcing for data cleaning / extraction

- Locu / GoDaddy
 - http://www.oreilly.com/pub/e/3298
 - "learnings from 17 conversations with companies that make heavy use of crowd work"



Conclusions

- Crowdsourcing: a way to get manual data annotation / cleaning / processing at scale
- Applications to search
 - Evaluation / relevance judgments
 - Complex query understanding
 - Information Finding (e.g., customer care phone no)
 - Result extraction and aggregation in tabular format

Conclusions

- Challenges
 - Quality if public crowds are used
 - Many techniques can be used to guarantee high quality, commercial services are coming up
 - Deadlines: it is difficult to predict crowd execution time
 - Task reward can be used as a means to speed-up execution
 - Cost: can be reduced thanks to hybrid humanmachine systems