City Data Fusion

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Listening to the pulse of our cities during City Scale Events

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Agenda

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- Context
- Ingredients and challenges
- Research hypothesis
- Streaming Linked Data Framework
 - design principles
 - Architecture
 - Components
- Testing the research hypothesis
 - London Olympic Games 2013
 - Milano Design Week 2013
- Conclusions



The digital reflection of reality is sharpening



Streams of information flows through our cities thanks to:

the pervasive deployment of sensors in our cities



the usage of (location-based)

social networks

the wide adoption of smart phones (equipped with sensors)



the availability of datasets about urban environment





We can feel the pulse of our cities by







- Characteristics
 - Lasting days
 - Hundreds of venues
 - Thousands of events
 - Hundreds of thousands of visitors
- Questions
 - Which are the most attractive events?
 - What do visitors think about the events they join?
 - What is their mood before, during and after the event they join?
- Ground truth
 - The program of the event
 - News about the event





Example of City Scale Event The Milano Design Week 2013 (MDW'13)

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- Problem
 - Sponsor and organizer of a city scale event needs to quantify the return on investment
- Existing solutions
 - Spread people around the hundreds of event to asses the success of the various events is effective, but expensive
- Challenge
 - Obtaining comparable results by analysing public social streams



















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Real data recorded on 13 April 2013 between 13:00 and 00:00

Step 2: subtract what is systematic















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Real data recorded on 13 April 2013 between 13:00 and 00:00

Tale Umbria





Open Street Nap

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Real data recorded on 13 April 2013 between 13:00 and 00:00

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Real data recorded on 13 April 2013 between 13:00 and 00:00



semantic technologies

- Address "variety" using Ontology Based Data Access
- Named Entity recognition and linkage
- Knowledge discovery (e.g., detecting systematicy)

streaming algorithms



- Address "velocity" of data stream
- Address "*volume*" by being able to process data that do not fit in main memory

crowd-sourcing techniques



Address "*veracity*" by cleansing and enriching data

Visual analytics



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- Allow no-expert access to data
- Tell stories out of data

Limitation of current systems

- Insufficient methods for making sense in realtime of heterogeneous data and social streams w.r.t. the vast collections of (open) data
- Lack of crowd-sourcing techniques whose incentives leverage needs of people in the urban environment
- Lack of visualization techniques tailored to non-experts







1. To scale order matters



3. Visualization must tell stories





Observation: order reflects recency, relevance, trustability

(v	Combinations	Continuous top-k Q/A	Order-aware reasoning
f order L	Relevance, ustability, etc.	Top-k Q/A	Top-k Reasoning
ypes o	Recency	DSMS/CEP	Stream reasoning
	Indexes	Traditional solutions	Scalable reasoning
~ <u>~</u>		No	Yes
		Types of r	reasoning

 harnessing orders is key to make sense in real-time of heterogeneous, massive and volatile data

Research hypothesis: urban-centric incentives!

MILANO

[source: http://www.behance.net/gallery/Maslows-Hierarchy-of-Needs-Infographic/4376921]



- **NNO**CENTIVE® amazon mechanical turk **Re**CAPTCHA[™]
- incentives designed for urban environment and life styles are key



Research hypothesis: visualizations must tell stories



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Testing the research hypothesis

- 1. To scale order matters
 - Stream Reasoning
 - RDF Stream
 - Continuous SPARQL
 - Incremental Materialization for RDF Streams (IMaRS)
 - C-SPARQL Engine
 - RESTFul Services for RDF Stream Processors
 - Streaming Linked Data Framework
 - SPARQL Rank
 - Rank aware SPARQL algebra
 - ARQ-Rank
- 2. Crowdsourcing needs the urban-centric incentives
 - Urban Games With A Purpose
 - UrbanMatch
 - Urbanopoly
- 3. Visualization must tell stories
 - On going work





Testing the research hypothesis Focus of this key note

- 1. To scale order matters
 - Stream Reasoning
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- Input Data Formats
 - Streaming information: RDF streams
 - Background Information: RDF graphs
- Query Language
 - Continuous SPARQL
- Features
 - Adapters to access the social streams, e.g., twitter
 - Ability to record and replay portions of the social stream
 - Possibility to decorate the social stream with sentiment information
 - Possibility to continuously analyzing the social stream
 - Possibility to built complex application composing complex networks of decorators and analyzers
 - Possibility to publishing and visualizing results of continuous analysis





Streaming Linked Data Framework **Design Principles**



- Follow publish/subscribe patter
 - Publisher and subscribers do not have to know each other
 - Subscribers can join and leave in any moment
- Adopt a reliable message-passing
 - Guarantees delivery order
- Minimise latency by using main memory
 - Avoiding disk I/O bottleneck



Streaming Linked Data Framework Architecture







Streaming Linked Data Framework Architecture - Adapters





Available adapters include: twitter, instagram, pachube, and linked sensor data



Streaming Linked Data Framework Architecture – RDF stream Bus





Streaming Linked Data Framework Architecture - Publishers





* Barbieri, D.F., Della Valle, E.: A proposal for publishing data streams as linked data - a position paper. In: LDOW. (2010)

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Streaming Linked Data Framework Architecture – Recorders and Re-players



A re-player re-plays a recorded stream. It can vary the speed.



Streaming Linked Data Framework Architecture - Analysers







Streaming Linked Data Framework Architecture - Decorators





A decorator adds information to streamed RDF graphs that match a given patter.

E.g., Dictionary-based sentiment classifiers* (known to be efficient for short texts concentrating on a single topic, such as tweets) was used in this work to decorate each tweet.

* Tsytsarau, M., Palpanas, T., Denecke, K.: Scalable Detection of Sentiment-Based Contradictions. In: DiversiWeb workshop, WWW, Hyberabad, India (2011)



Streaming Linked Data Framework Architecture - Visualizers







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Testing the research hypothesis The London Olympic Games 2012 (LOG'12)

- Problem
 - To manage a big event requires tracking in real time the movement of crowds.
- Existing solutions
 - CCTV, and mobile network data analysis are effective, but expensive
- Challenge
 - Obtaining comparable results by analysing public social streams





Test bed LOG'12 Case study #1 tracking attention of the crowds

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- The problem
 - Managers of Big event want to track in real time if the event is capturing the attention of the audience.
- Input
 - 20 thousand of tweets streamed by Twitter between 9pm July 27th and 1am July 28th
- Ground truth
 - The Opening Ceremony Broadcast
 - <u>http://www.youtube.com/watch?v=4As0e4de-rI</u>
 - The wikipedia page describing the Opening Ceremony
 - <u>http://en.wikipedia.org/wiki/</u>
 <u>2012_Summer_Olympics_opening_ceremony</u>



Test bed LOG'12 – study case 1 The Opening Ceremony







Interesting phenomena are visible at different scales



World





Continent

City







 The Opening Ceremony is clearly visible in the volume of tweets containing LOG'12 related hashtags



Global scope







 Bursts of hashtags usage capture what attracts the attention of those watching the ceremony world wide



Detailed Analysis

Moments of Ceremony	# of hashtags	Fraction
Total	189	100.00%
Hashtagged with an emotional state	34	17.99%
Correctly hashtagged	72	38.10%
- Right on time (1 min tolerance)	50	26.46%
- After the event (15 min tolerance)	13	6.88%
- Before the event (15 min tolerance)	9	4.76%



- Audience loosed attention while the ceremony was progressing
- Audience emotions where more evident in the first and the last part of the ceremony

		0
1. Countdown	(21:00-21:04)	1 2
2. Green and Pleasant Land	(21:04–21:09)	90% - E
3. Pandemonium	(21:09–21:25)	
4. Happy and Glorious	(21:25–21:35)	80% -
5. Second to the right	(21:35–21:47)	70% -
6. Interlude	(21:47–21:52)	60% -
7. Frankie and June say	(21:52–22:09)	50% -
8. Abide with Me	(22:09–22:20)	40% -
9. Welcome	(22:20–00:00)	30% -
10.Bike a.m.	(00:00-00:07)	0070
11.Let the Games Begin	(00:07–00:24)	20% -
12.There Is a Light	(00:24–00:38)	10% -
13.And in the end	(00:38–00:46)	0%
		1 2





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- The problem
 - detect the events given the position of a set of venues and socially listening their surroundings
- Input
 - three million tweets streamed by Twitter between July 25th and August 13th 2012
- Conditions

Туре	Venue	Events	Capacity (seats)
Large	Olympic stadium	athletic games	80,000
Medium	Aquatic centre	swimming, diving and synchronized swimming	17,500
Small	Water polo arena	Water polo	5,000

- Ground truth
 - calendar of Olympic Games



Test bed LOG'12 – case study 2 SLD application



(Metro Burst -> Outside Burst -> In ?venue Burst) within 30 min => event in ?venue



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Test bed LOG'12 – case study 2 **Evaluation**



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Test bed LOG'12 Case study #3 visualizing crowds movements



- The problem
 - Visualize the movement of the crowds
- Input
 - three million tweets streamed by Twitter between July 25th and August 13th 2012
- Conditions

Туре	Venue	Events	Capacity (seats)
Large	Olympic stadium	athletic games	80,000
Medium	Aquatic centre	swimming, diving and synchronized swimming	17,500
Small	Water polo arena	Water polo	5,000

Ground truth

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• Expert judgement ;-)

Test bed LOG'12 – case study 3 **Evaluation**



Opening Ceremony at the Olympic Stadium



A crowded event at the Aquatic Centre (July 31st, 2012)





Testing the research hypothesis The Milano Design Week 2013 (MDW'13)

MILANO

- Problem
 - Sponsor and organizer of a city scale event needs to quantify the return on investment
- Existing solutions
 - Spread people around the hundreds of event to asses the success of the various events is effective, but expensive
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- The problem
 - Is MDW visible in the social streams posted by people in Milano area? If yes in real-time,
 - 1. What are the districts from which MDW visitors post the most?
 - 2. What are the most frequently used hashtags?
 - 3. How do people feel before, during and after the event they join?
 - Can these question answered at a cost a SME can afford?
- Input
 - 106,770 tweets streamed by Twitter between April 9th and April 14th 2013





Test bed MDW'13 – case study 1 **SLD** application



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April 9-14, 2013

- Distinct users
 12.031
- Invocation of Linked Data Publisher
 - 1,136,052
- Cost
 - 25 €/month
 - 2 cores, 2GB





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# TOP	HASI	ITAG	i 10 hashtag maggi conteggiati ogni 15	ormente usati minuti.	nei tweet a Milano,	
triennale						
						15
fuorisalone					11	
inaugurazione					11	
					11	
milano						
					11	
design				10		
designweek				10		
designmeen			9			
milan						
			9)		
TDM6						
sindromeinfluenza		5				
andromennuenza	3					
museum						.•
2						







Test bed MDW'13 – case study 1 The results question 1





MDW 2013 is visible in the volume of micro-posts







April 9th, 2013 at 18.00	posts	April 11th, 2013 at 18.00	posts	Μ
fuorisalone	30	milano	25	is
designweek	28	fuorisalone	22	דו ה
nabasalone	20	design	10	
milano	9	designweek	6	n
design	6	32giornata	6	

MDW 2013 is visible in the top-5 hashtags used in the micro-posts

April 13th, 2013 at 18.00	posts	A
fuorisalone	28	inte
designweek	21	dire
nabasalone	17	caę
milano	10	mil
inter	8	ser

April 15th, 2013 at 18.00	posts
nter	20
diretta	11
cagliarii	6
nilan	4
seriea	3





Venue	posts	The most
cesati antiques & works of art	16653	attractive
Porta nuova 46/b	13416	found
Circolo Filologico	9891	lound
Adele Svettini Antichità	7366	
ALTAI	5592	\frown
Bigli19	5175	
Dudalina	4875	
Galleria DadaEast	3550	
borronichemicals	1078	
Antonio Lupi Showroom Milano	995	
Instituto Cervantes Milano	752	
GALLERIA D'ARTE CONTEMPORANEA CINESE	560	



Test bed MDW'13 – case study 1 The results question 3

- MILANO
- The number of tweets posted by the same user where not enough to answer question 3 (How do people feel before, during and after the event they join?)

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The problem



- Is the launch of ASUS products during MDW visible in the social streams posted by people around the world?
- If yes, not necessarily in real-time,
 - 1. What are the products that attract more attention?
 - 2. What is the global sentiment before, during and after the launch?
- Input
 - 107,044,487 tweets registered with SLD between April 3rd and April 30th, 2013 asking Twitter to send to SLD tweets containing 300 words related to MDW, ASUS and its products
- Ground Truth
 - News, movies, and other media published on the Web in the same time period
 - Tool: google advance search



Test bed MDW'13 – case study 2 The results



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Test bed MDW'13 – case study 2 The results



- As expected, the method did not handle sarcasm in a satisfactory manner
- some tweets about FonePad contained sarcastic sentences
 - e.g., "wanna buy it so bad!"
 - It was classified as negative
 - it was expressing positive sentiment.





Wrapping up



- Listening to social streams and proper visual analytics methods can unveil interesting city scale phenomena
 - where people are gathering
 - what people are interested in
 - where people interested in a given topic are
 - if an event is running
 - what people feels about some topic
 - if the people are perceiving the message an event organizer want to transmit
 - how the feeling of the people evolves over time

• ...





- lack of volume in social streams prevents to perform meaningful analysis
- Sarcasm and idioms hinder the possibility to capture the opinion of people from highly volatile social streams
- limited a priori knowledge about the event hampers the ability to link social content to background data



Wrapping up Next steps



- Basic research
 - continuous top-k query answering
 - crowdsource data cleansing and linking
 - determining what is systematic is difficult
- Applied research
 - profile a crowd
 - find opinion makers
 - predict social trends





Who	Semantic techs	Streaming algorithms	Crowd- sourcing	Visual analytics
CEFRIEL				
Density Design Lab – PoliMi				
DISI – University of Trento				
KDD Lab – ISTI, CNR, Pisa				
ML Group – SIEMENS				
Ontology Eng. Group – UPM				
Saltlux – Korea				
SKIL Lab - Telecom Italia				
Studio Labo				
Web IS - TU Delft				



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Thank you! Any question?

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