



Social Mining & Big Data Ecosystem

H2020 - www.sobigdata.eu September 2015- August 2019

**GARR Conference – 16th November 2017** 

@SoBigData (https://twitter.com/SoBigData)

https://www.facebook.com/SoBigData



### **SOBIGDATA VISION AND VALUES**

## Big data "proxies" of social life

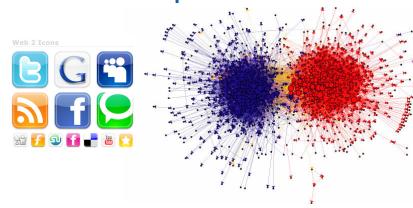
#### **Shopping patterns & lifestyle**



#### Desires, opinions, sentiments



#### Relationships & social ties

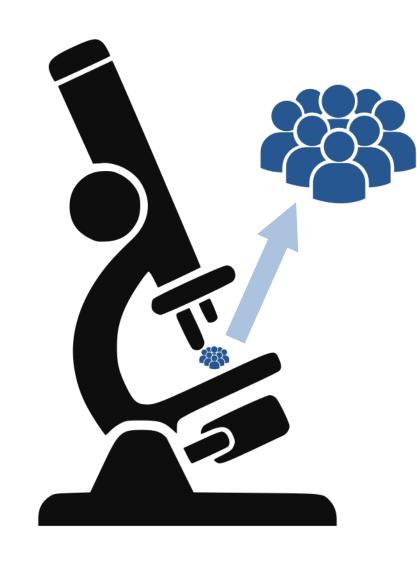


#### **Movements**



GARR - 16th November

Social mining: making sense of big data to understand society



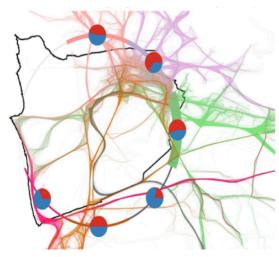
## What is Social Mining

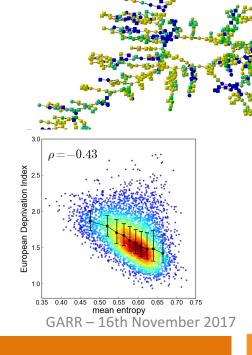
- Automated discovering patterns and models of human behaviour across the various social dimensions that have big data "proxies"
  - desires and opinions

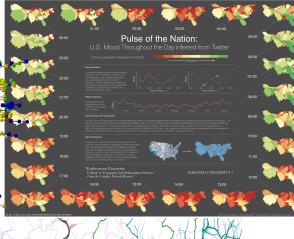
relationships and social ties

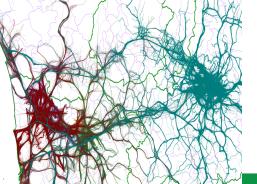
life-styles

mobility

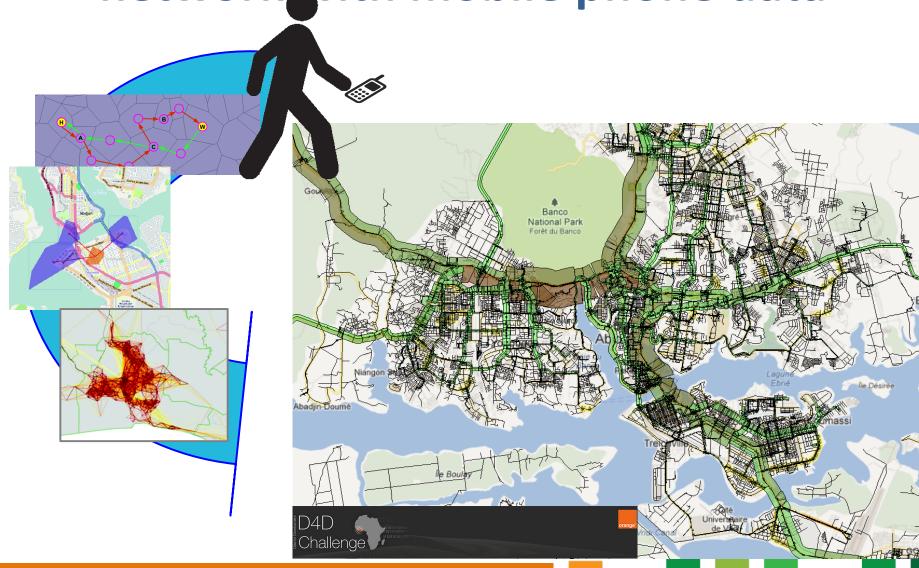




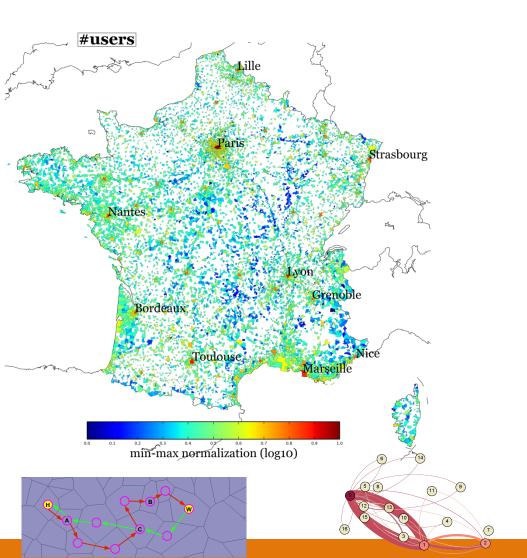


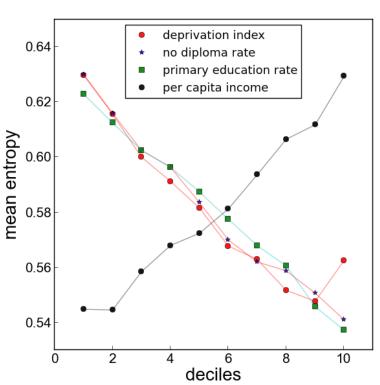


## Estimating traffic fluxes on road network with mobile phone data



## Estimating Country well-being with mobility Diversity





$$d_i^{(n)} = \sum_{j=1}^{|V|} \frac{1}{k_j} M_{ij} p_j^{(n-1)} \forall i$$

$$p_j^{(n)} = \sum_{i=1}^{|U|} \frac{1}{k_i} M_{ij} d_i^{(n-1)} \forall j$$

## **Social Sensing**

Real-time crisis mapping

TABLE 6. Results of evaluating our system's ability to detect areas damaged by the Emilia earthquake.

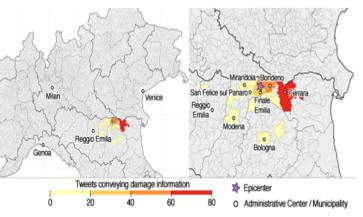
|                                               | <b>Evaluation metrics</b> |        |             |          |           |       |  |  |
|-----------------------------------------------|---------------------------|--------|-------------|----------|-----------|-------|--|--|
| Task                                          | Precision                 | Recall | Specificity | Accuracy | F-measure | MCC*  |  |  |
| Detect all damaged areas                      | 0.895                     | 0.202  | 0.992       | 0.797    | 0.330     | 0.365 |  |  |
| Detect areas that suffered significant damage | 0.867                     | 0.813  | 0.992       | 0.982    | 0.839     | 0.830 |  |  |

<sup>\*</sup>MCC: Matthews correlation coefficient

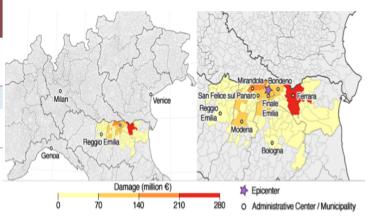
TABLE 7. Results of evaluating our system's ability to detect areas damaged by the Sardegna flood.

| Task                                                | Evaluation metrics |        |             |          |           |       |  |
|-----------------------------------------------------|--------------------|--------|-------------|----------|-----------|-------|--|
|                                                     | Precision          | Recall | Specificity | Accuracy | F-measure | MCC*  |  |
| Detection of all damaged areas                      | 0.640              | 0.410  | 0.973       | 0.915    | 0.500     | 0.470 |  |
| Detection of areas that suffered significant damage | 0.500              | 0.643  | 0.973       | 0.960    | 0.563     | 0.545 |  |

<sup>\*</sup>MCC: Matthews correlation coefficient

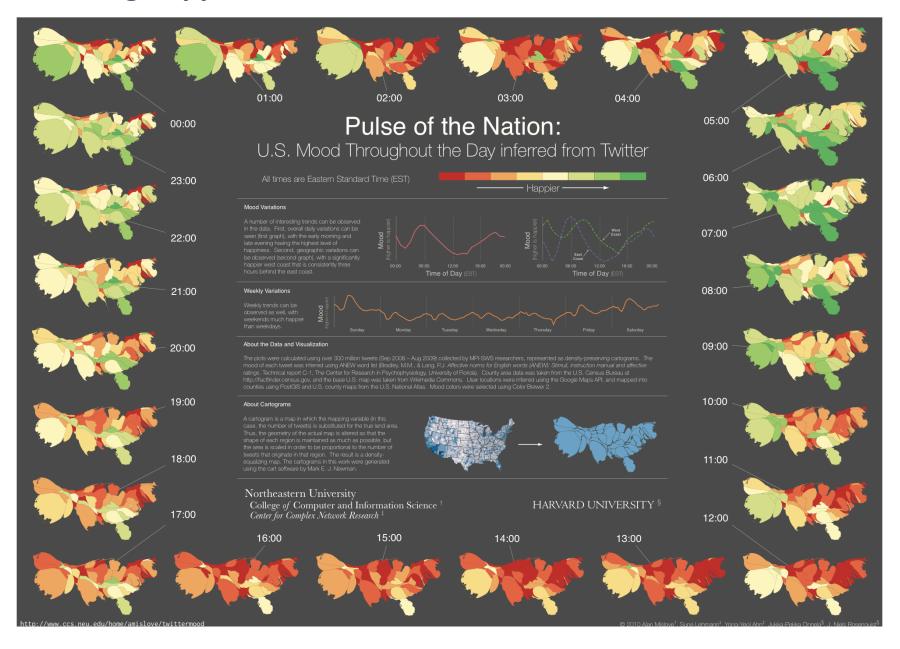


Overview and detail of the crisis map generated by our system.



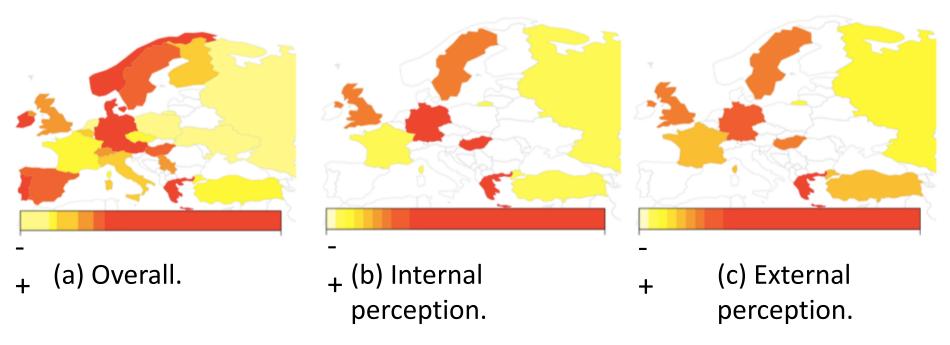
Overview and detail of the authoritative post-event damage assessment map. Authoritative data about the economic losses are provided by Emilia Romagna regional district (http://www.openricostruzione.it).

#### Measuring happiness with twitter data



## Sentiment Analysis

- Internal and external perception by country
  - Index ρ the ratio between pro refugees users and against refugees users
  - Red means a higher nredominance of nositive sentiment higher of





## 3 Million Brexit Tweets Reveal Leave Voters Talked About Immigration More Than Anything Else

Groundbreaking analysis shows immigration, not sovereignty or the NHS, dominated the conversation – and making British judges responsible for British law was a key theme for Leave supporters.



James Ball
BuzzFeed Special
Correspondent



Chris Applegate
Editorial Developer, UK

posted on Dec. 9, 2016, at 2:03 p.m.













https://www.buzzfeed.com/jamesball/3-million-brexit-tweets-reveal-leave-voters-talked-about-imm?utm\_term=.jmDQE9JNR#.fuOOrb145

## What's special in social mining?

- Any data science experiment is composed by:
  - data acquiring (open data, crowdsourcing, crowdsensing,)
  - model building (data mining, machine learning, network science, ...and very complex validation phase),
  - creation of an exploration scenario (what-if analysis) (different validation setting),
- ....similar to many other data-driven science process,...but data are produced by humans

### What is needed

- responsibility and trust
  - FACT: Fairness, Accuracy, Confidentiality and Transparency
- harness social mining for scientific advancement and for the social good
  - FAIR: Findable, Accessible, Interoperable, Reproducible
- responsible open data science





TO CONSTRUCT THE Multidisciplinary European Infrastructure on Big Data and Social Data Mining (the Social Mining CERN) providing an integrated ecosystem for ethic-sensitive scientific discoveries and advanced applications of social data mining on the various dimensions of social life, as recorded by "big data".





### The pillars for reaching the goal

- an ever-growing, distributed data ecosystem for procurement, access and curation of big social data, within an ethic-sensitive context, based on
  - innovative strategies for acquiring social big data for research purposes,
  - using both opportunistic means offered by social sensing technologies and
  - participatory means based on user involvement as prosumers of social data and knowledge.



### The pillars for reaching the goal

- an ever-growing, distributed platform of interoperable, social data mining methods and associated skills:
  - tools, methodologies and services for mining, analysing, and visualising complex and massive datasets,
  - harnessing the techno-legal barriers to the ethically safe deployment of big data for social mining.



## The pillars for reaching the goal

 Building the Social Mining community of scientific, industrial, and other stakeholders (e.g. policy makers),



## The path to achieve the goals

- Integrate European national infrastructures and centres of excellence in big data analytics, social mining and data science
  - 1. Text and Social Media Mining (TSMM)
  - 2. Social Network Analysis (SNA)
  - 3. Human Mobility Analytics (HMA)
  - 4. Web Analytics (WA)
  - 5. Visual Analytics (VA)
  - 6. Social Data (SD)

## Integrating national research Infrastructures





### The Consortium



























## The path to achieve the goals

- Grant access (both virtual and trans-national on-site) to the SoBigData RI to multidisciplinary scientists, innovators, public bodies, citizen organizations, SMEs, as well as data science students at any level of education.
- joint research, and extensive networking and innovation actions

#### **Big Data Ecosystem**

- Open Data
- Restricted Data
- Virtual Collections



- Text & Social Media Mining
- Social Network Analysis
- Human Mobility Analytics
- Web Analytics
- Visual Analytics
- Social Data



Ethical and Legal Framework



E-infrastructure



#### **Transnational Access**

Open calls Exploratory projects



#### Networking

Training
Dissemination
Innovation Accelerator

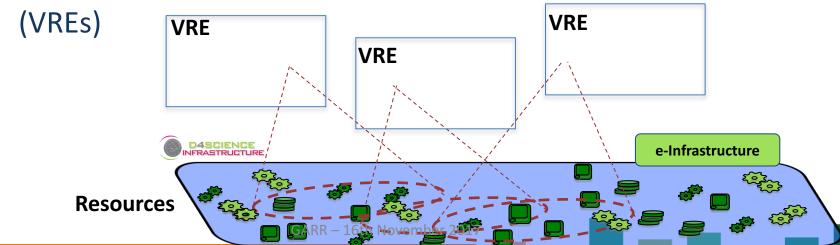


### Lines of actions of the RI

- Overcome data and methods fragmentation
- Attract critical mass of users
- Multidisciplinary community Building
- Creating a new generation of DataScientists
- Improve SME competiveness

## SoBigData e-infrastructure

- It is powered by the D4Science platform
  - Used as a production system by other communities
    - Users: +3600 (SoBigData is 16.5%, fast growing), Computing: 1500 cores,
       Storage: 400 TB)
- Supports basic services
  - Authentication, Authorization, Accounting framework
  - Resource Catalogue
  - Virtual Research Environments management framework



## SoBigData VREs



**Exploratories** 

Access

Training

Dissemination

Consortium

**EU Project** 

Blog

Q

**Virtual Access** 

Home » Access

The Virtual Access offers online services for big data and social mining research. A web front-end comprising of a catalogue of SoBigData resources (data and services) and a set of SoBigData Virtual Research Environments (VREs).

VREs are web-based working environments equipped with a number of applications, enabling scientists to have access to the set of data, services, and algorithms needed to perform their investigation in a **collaborative way**.



Go to the **SoBigData LAB** to perform your experiments



Go to VRE **Exploratories** to see the thematic environments of SoBigData



Go to the **Catalogue** to have a list of our methods and datasets



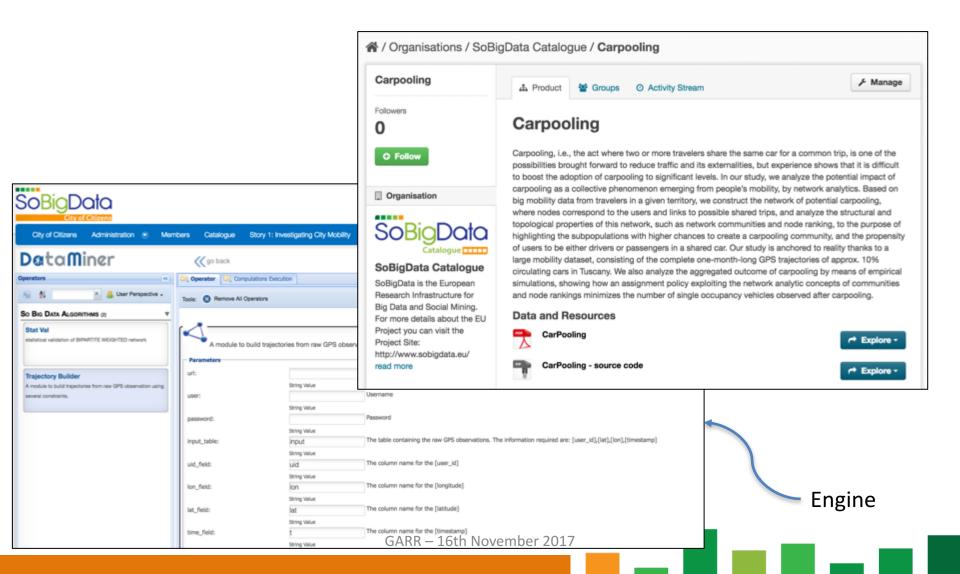
Go to SoBigData Virtual Research
Environments - VREs

- City of Citizens
- Societal Debates
- Wellbeing
- Migration
  - Sport Science



Smaph

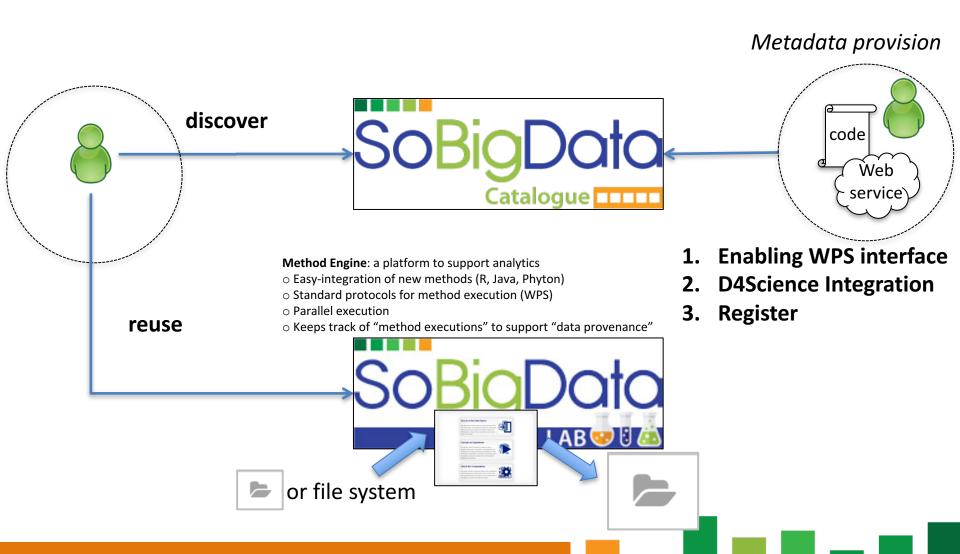
## SoBigData Lab and Catalogue VREs



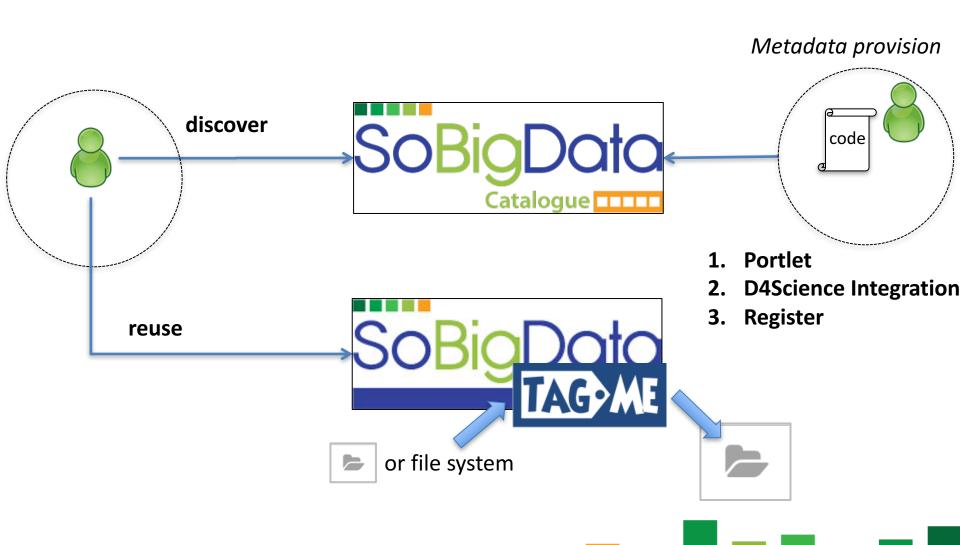
## Methods as Software: patterns of sharing and reuse



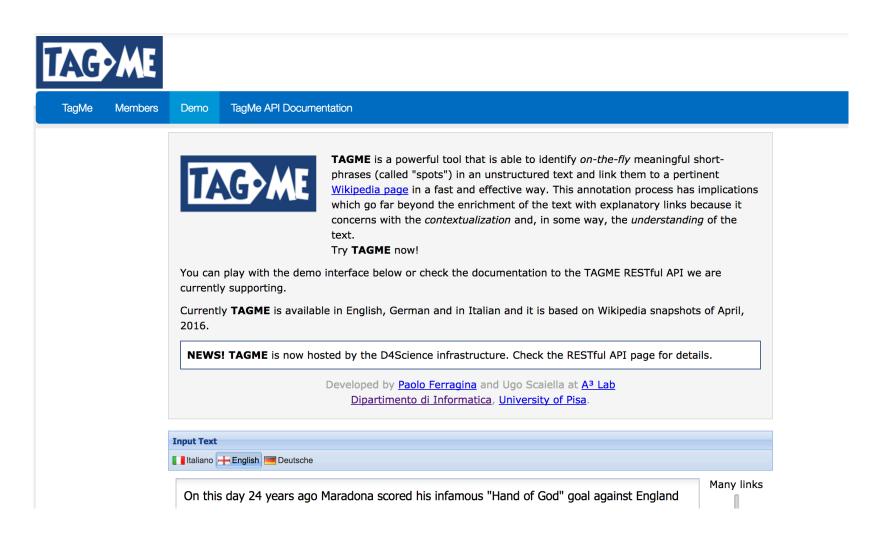
## Method as a Service: patterns of sharing and reuse



## Web application: patterns of sharing and reuse

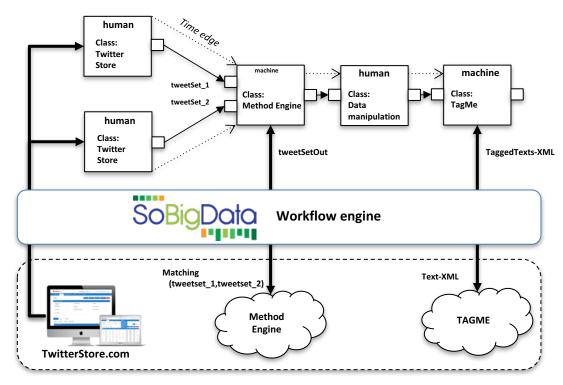


## **Application VRE**



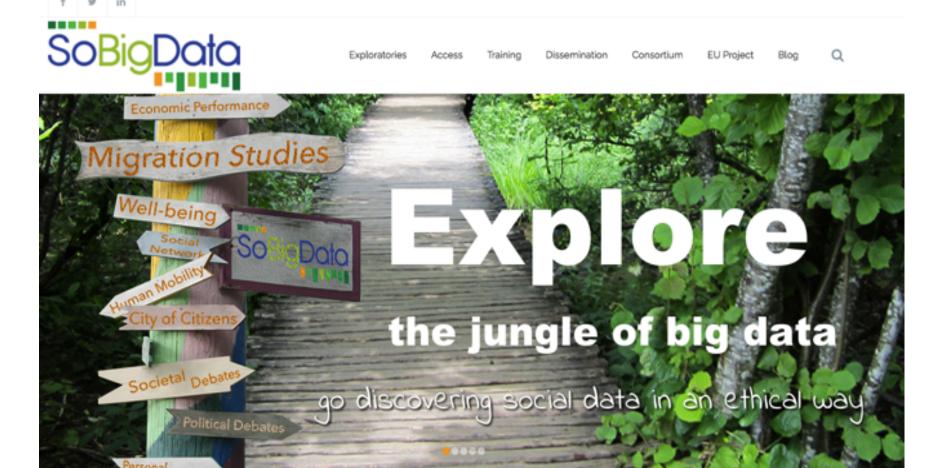
## (Hybrid) Workflow Language

- Issue: repeatability is limited to individual methods as a service
- Goals: Workflow language to model and share scientific processes to enable their repeatability
- Challenges: enabling repeatability by modeling combinations of manual and autonomic steps





## **Exploratories**

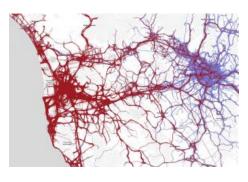




## **Exploratories**

# Virtual Research Environments tailored on specific domains where Social Mining is applied

- Promotes results sharing among scientists and communities
- Promotes the use of RI through Virtual and Transnational Access

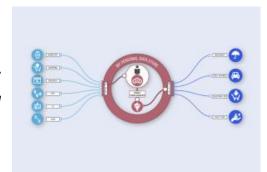


#### City of Citizens

This exploratory tells stories about cities and people living in it. We describe those territories by means of data, statistics and models.

#### Well-being & Economic Performance

Can Big Data help us to understand relationships between economy and daily life habits? We use data of purchases in supermarkets and investigate people's behavior.





#### Societal Debates

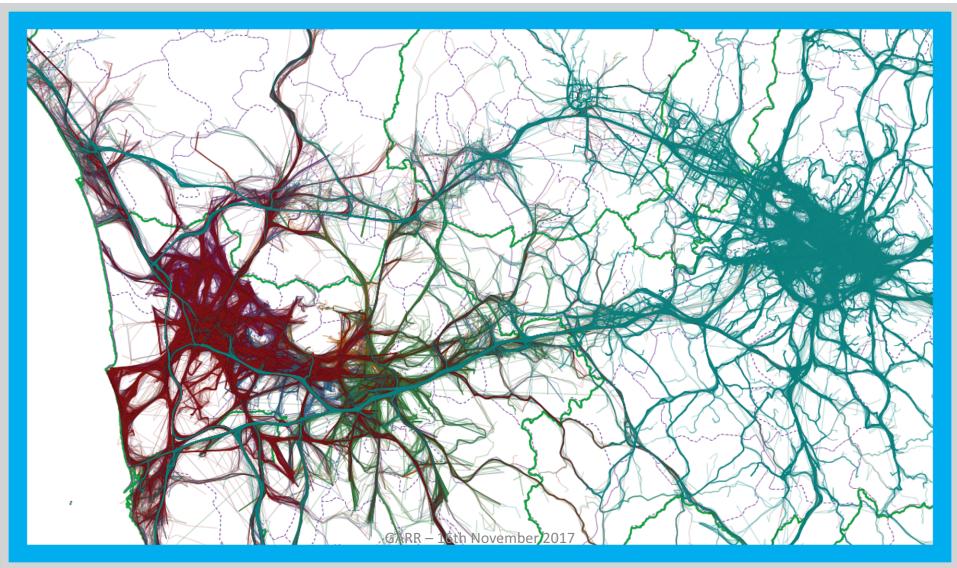
We study public debates on social media and newspaper. We can identify themes, following the discussions around them and tracking them through time and space.

#### Migration Studies

Could Big Data help to understand the migration phenomenon? We try to answer to some questions about migrations in Europe and in the world.



## Exploratory: Big Data for City of Citize s



### Investigating City Mobility

How do people move into the city? How does the traffic change during the day? And how does it vary during the week? How does the turism presence affect the traffic? Our data scientits already study the traffic in the Italian cities of Pisa and Florence by analyzing Big Data sources such as mobile phone traces, veicular gps and social media data as proxy of human behaviour.

The results could be useful for both local administrators and citizens. The local administrators could have a tool to quantify accurately city's traffic and understand city's usage, so they could take better decisions to manage mobility. Citizens could take informations to know traffic situation in real time and they could choose the best and fastest way.

Our studies could be useful in carpooling, too. Indeed, Big Data analysis can suggest to citizens who can share the travel with them.



Datasets Metho

Methods & Tools Thematic Clusters

People

Contact

GPS data of private cars

CDR data from a telecommunication company

Foursquare check-ins

Flickr geo-localized photos

Official administrative information (OAI)

Living the story in SoBigData Virtual Research Environment

(only alpha version available - final service will be available in September 2016)

| Datasets             | Methods & Tools | Workflows | Thematic Clusters | People                   | Contact                    |  |  |
|----------------------|-----------------|-----------|-------------------|--------------------------|----------------------------|--|--|
|                      |                 |           |                   |                          |                            |  |  |
| Method               |                 |           | Partner           | SoBigDa                  | SoBigData RI - Integration |  |  |
| Urban Profiles       |                 |           | AALTO             | Service h                | Service hosted, Download   |  |  |
| Urban Mobility Atlas |                 | CNR       | Web Serv          | Web Service              |                            |  |  |
| Trajectory Builder   |                 |           | CNR               | Service h                | Service hosted, Download   |  |  |
| Borders              |                 |           | CNR               |                          |                            |  |  |
| Sociometer           |                 |           | CNR               | Download                 |                            |  |  |
| Trip Builder         |                 | CNR       | Web Page          | Web Page                 |                            |  |  |
| Car Pooling          |                 | CNR       | Service h         | Service hosted, Download |                            |  |  |
| MyWay                |                 | CNR       | Service h         | Service hosted, Download |                            |  |  |
| Privacy Risk         |                 | CNR       | Service h         | Service hosted, Download |                            |  |  |
| O/D Matrix           |                 | CNR       | Web Serv          | Web Service              |                            |  |  |
| Mobility Prof        | îles            |           | CNR               | Service h                | osted, Downloa             |  |  |
| Exploration (        | of Time         |           | FRH               | Download                 |                            |  |  |
| Statistical Va       | alidation       | ·         | SNS               |                          |                            |  |  |

#### Living the story in SoBigData Virtual Research Environment

(only alpha version available - final service will be available in September 2016)

## **Exploratory VREs**



GARR - 16th November 2017

Last Updated

AccessibilityMode

CreationDate
Creator
Field/Scope of use

http://www.sobigdata.eu/ read more 6 dicembre 2016, 14:53 (UTC+01:00)

Trasarti, Roberto, roberto.trasarti@isti.cnr.it

Download

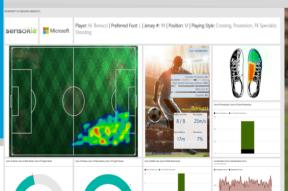
Predicting trajectories







Data Science for sports analytics







## **Ethics and trust**



### Legal and Ethical framework

Define and implement the legal and ethical framework of the SoBigData RI, in accordance with the European and national legislations

## Monitor of research

Monitor the compliance of experiments and research protocols with the framework

#### Privacy-by-design

The development of big data analytics and social mining tools with value-Sensitive Design and privacy-by-design methodologies

## The ethics of SoBigData

- Gathering large quantities of data may have serious consequences:
  - consequences range from personal harm,
  - to issues of autonomy, injustice and inequality.
- SoBigData adheres to a value-sensitive design approach:
  - design solutions to overcome ethical dilemma's, in this case those between the utility of the data gathered vs. the protection of the individuals subject to the research.

## The ethics of SoBigData

- How do we create an infrastructure in which such methods can be disseminated and improved upon?
- Data Management Plan plays a key role:
  - Each data has its privacy requirements and fact checks and responsibility
- Anonymization techniques are part of the research
- Researchers will be trained in applying the necessary procedural safeguards



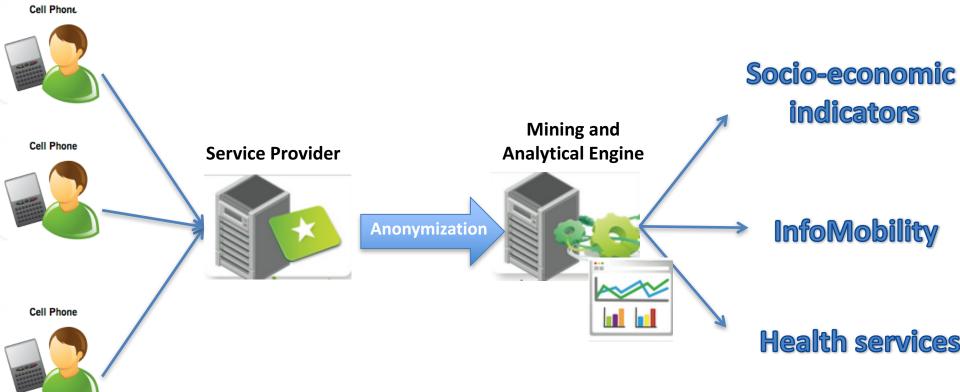


#### **REGULAR ARTICLE**

**Open Access** 

## Privacy-by-design in big data analytics and social mining

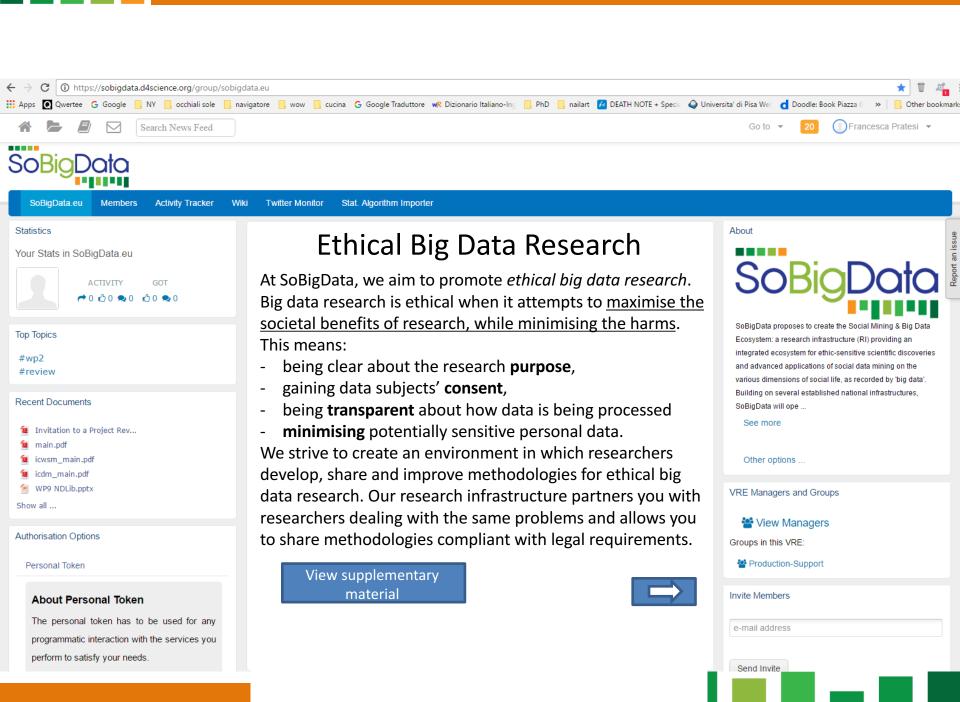
Anna Monreale<sup>1,2\*</sup>, Salvatore Rinzivillo<sup>2</sup>, Francesca Pratesi<sup>1,2</sup>, Fosca Giannotti<sup>2</sup> and Dino Pedreschi<sup>1</sup>

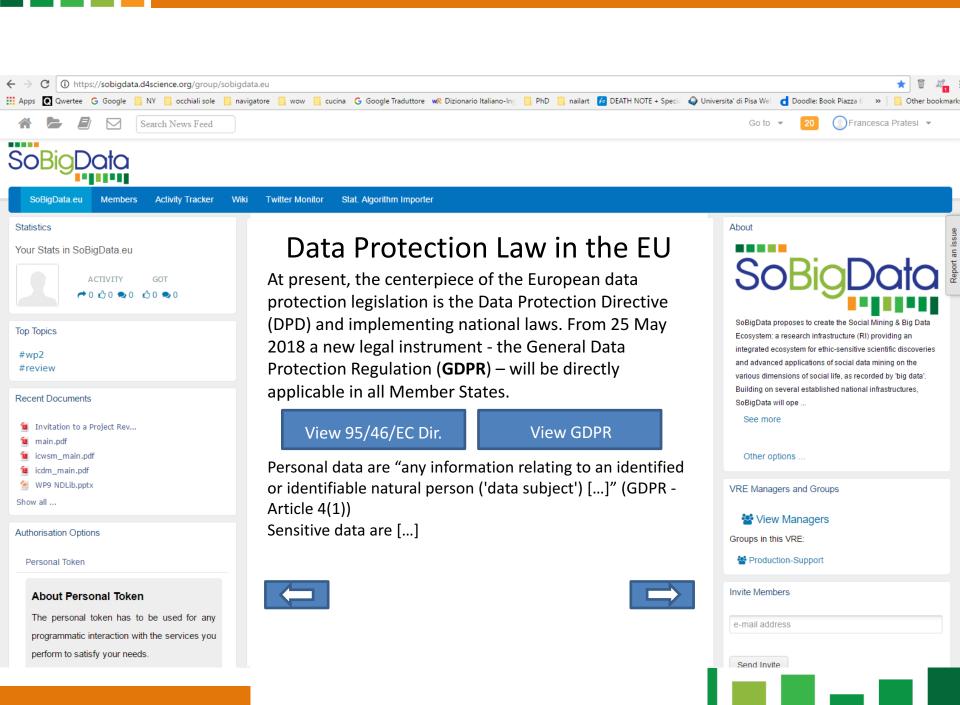


# Educating the responsible data scientists

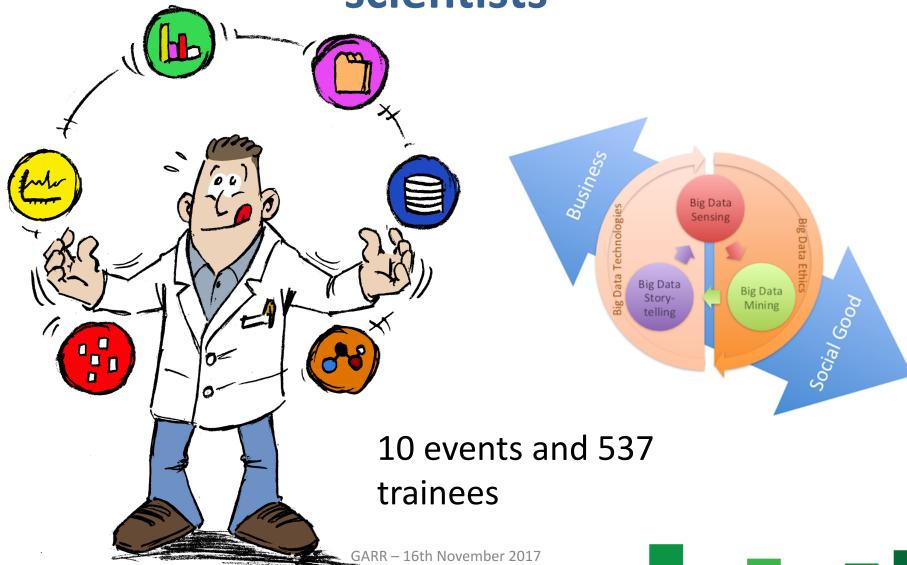
Based on a cooperation between ethicists and computer

- 1. A Massive Online Open Course (MOOC) which instructs all prospective researchers about the legal and ethical dangers of big data research and the steps they can take to minimise these;
- 2. A set of workflows that outline the steps researchers can take when designing their approach;
- 3. Information pop-ups which redirect researchers to state-of-the-art ethical methods.





# Creating a new generation of data scientists



## **Datathons**

- Zurich and Delft : young startuppers (40 participants)
- London: journalists
- Tartu: social scientist (220 participants)
- Pisa: High School students
  - Il level Master Bigdata Analytics (4th edition)
  - Master degree in data science (since 2004)
- Summer School
- Phd programme in Data Science in Pisa



An initiative devoted to **Tuscan** enterprises interested in exploring the potential of **big data** in improving the production processes, their market or their interaction with customers and suppliers.

The Tuscan Data Challenge gives opportunity to all the Tuscan companies, that usually do not use their **data**, to enter the world of Big Data and exploit its potential.

#### Who is Involved









## **Tuscan Big Data Challenge 2016**

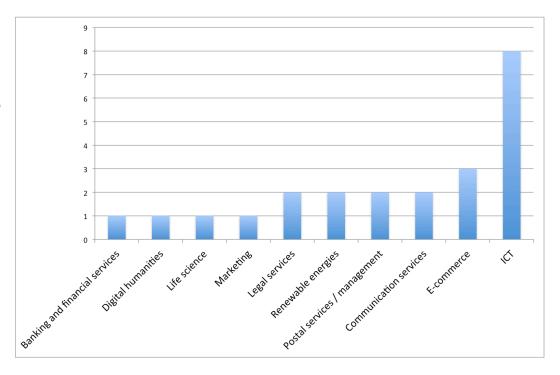
## **Results in Short**

Participants: 23 Tuscan SMEs

Selected: 9

## Type:

- Legal services
- Banking and financial services
- E-commerce
- ICT
- Renewable energies
- Postal services
- Digital humanities
- Life science
- Communication services
- Marketing





#### SoBigData TRANSNATIONAL ACCESS

Transnational Access supports short term scientific mission (between 2 weeks and 2 months) of researchers or teams at one of the installations of SoBigData that will provide big data computing platforms, big social data resources, and cutting-edge computational methods:

- Interact with the local experts, discuss research questions Run experiments on non-public datasets and algorithms Present results at workshops/seminars
- to enable multi-disciplinary social mining experiments with the Research Infrastructure **assets**: big data sets, analytical tools, services and skills.

Site: http://www.sobigdata.eu/access/transnational

Contacts: Gate - Kalina Bontheva < k.bontcheva@sheffield.ac.uk>

**SoBigData.it** - Claudio Lucchese < claudio.lucchese@isti.cnr.it>

**Fraunhofer** - Thorsten May <thorsten.may@igd.fraunhofer.de>

UT - Jaan Ubi < jaan.ubi@ut.ee> L3S - Thomas Risse < risse@L3S.de>

AALTO UNIVERSITY - Michael Mathioudakis < michael.mathioudakis@aalto.fi>

**Nervousnet** - Iza Moise <izabela.moise@gess.ethz.ch>

#### **KEY DATES**

#### Theme-driven exploratory calls:

Publish Call for Proposals: mid October 2017 Proposal Submission deadline: mid November 2017

Applicant Notification: **December 2017** 

Visits period: January – June 2018















#### Installations





general architecture for text engineering























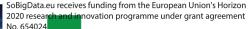
















## Become a SoBigData Supporter

Share and make findable your data science results

- Become part of the scientific network
- http://www.sobigdata.eu/join-us