

WeKnowIt

Making the Collective Intelligence of Social Media Searchable

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CERTH – ITI

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<http://www.weknowit.eu>

Contents

- Introduction
- Social Media - Collective Intelligence
- WeKnowIt approach
- Community detection in Social Media
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Web 2.0 content

flickr

- 3,190 uploads in the last minute
- 3.2 million things geotagged this month
- 4,754,012,299 photos (2 July 2010)

YouTube

- 24h of video content uploaded every minute
- 2 billion movies watched every day

facebook

- More than 400 million active users
- More than 200 million users log on at least once each day
- 2.5 billion photos uploaded each month



Winner



The winner of the WeKnowIt Grand Travel Challenge

Tags, content everywhere

Upload, tag, share, search



amsterdam beach berlin birthday blackandwhite blue boston bw california cameraphone canada car cat cats chicago china christmas church city clouds concert day dog england europe family festival florida flower flowers food france friends garden germany graduation graffiti green hawaii holiday home india italy japan june kids london london macro may me mexico moblog music nature new newyork newyorkcity newzealand night nyc paris park party people photo portrait red sanfrancisco scotland seattle sky snow span spring street summer sunset taiwan thailand tokyo travel tree trees trip uk unbound urban usa vacation vancouver washington water wedding white winter yellow zoo

Can we do more things?

The screenshot shows a search interface with a 'Search' bar at the top. Below the search bar, there are tabs for 'Photos', 'Groups', and 'People'. A 'SEARCH' button is visible, along with options for 'Full Text' and 'Tags Only' search. The main content area is titled 'Tag Clusters' and lists three clusters of photos based on tags:

- Photos with tags like nyc, newyork and manhattan
- Photos with tags like fruit, red and green
- Photos with tags like ipod, iphone and music

Below the clusters, there are several photo thumbnails with captions. Some captions include 'From sonnyfung', 'From War', 'From (karen)', 'From HAZEL- B. G. B.', 'From amy johanna', 'From nrvica', 'From fernando780', 'From jordanmarric...', and 'From tsunedri'. The interface also shows sorting options like 'Sort: Relevant', 'Recent', and 'Int', and view options like 'Small', 'Medium', 'Detail', and 'Slideshow'.

**By combining information from many photos
- tags, it seems that we can extract
Stable patterns
in tagging systems over time**



flickr®

Deutsches Eck from Ehrenbreitstein
Fortress, Koblenz, Germany



When you're high up on the hill above Koblenz at Ehrenbreitstein Fortress you can get a great panoramic view of the city and the surrounding area.

by [schaengel](#)

121 comments 69 faves

Tagged with [koblenz](#), [ehrenbreitstein](#) ...
Taken on November 15, 2009, uploaded
November 17, 2009

See more of [schaengel](#) photos, or visit
his profile.



... and more: Travel trends using flickr



Trace Flickr users from a chronologically ordered set of geographically referenced photos

Who are the Italians and who are the Americans?

MIT SENSEABLE CITY LAB, "The World's eyes"

Collective Intelligence!

Mass user-generated content Web 2.0

Little understanding

Analysis techniques:
Content, Social, Mass

Loose interaction

Organisations – Processes

No benefits from community and mass content

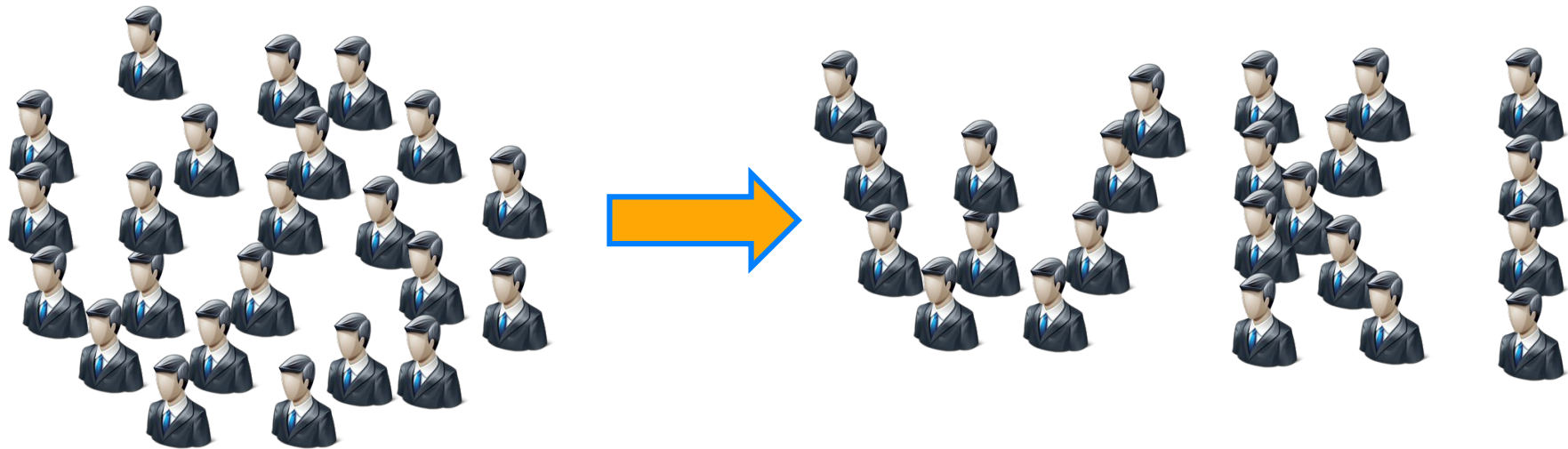
Users & Devices

Need for new services

weknowit 

Defining Collective Intelligence

Collective Intelligence is the Intelligence which emerges from the collaboration, competition and coordination among individuals.



...an Intelligence greater than the sum of the individuals' intelligence

WeKnowIt and CI

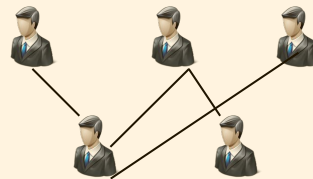
Collective Intelligence

Media Intelligence



User-generated content,
social context

Social Intelligence



Social Networks

Mass Intelligence



Blogs, forums,
ratings, voting

Organizational Intelligence



Knowledge
Management

Personal Intelligence

Upload and Access



Personal Intelligence



- >> Login, Upload
- >> Tag recommendation,
- >> Spam detection



Organisational Intelligence

- >> Log Merger

Access

Location	Name	Keyword	Role
Sheffield	Angela Rowland	answer phone message	fo
Ecclefield	Edie	sheffield city council	fo
Langley	Brian	george harcock	fo
GREAVES ROAD	jake	ecclefield school	fo
JORDON LANE	Angie	team officer	fo
NETHER LAKE	Christine	chemical incident	fo
HARBOROTH	Newton	police presence	fo



Emergency Response

Media Intelligence



Picture arrives at emergency response

- >> Automatic localisation of photo



Mass Intelligence

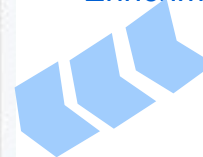
Many contributors

- >> Clustering
- >> Tag Harmonization
- >> Enrichment from add. sources

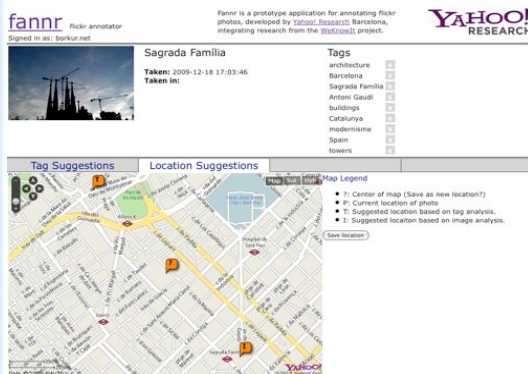


Social Intelligence

- >> ER Alert Service



Travel prototype



Post Travel

Mass Intelligence

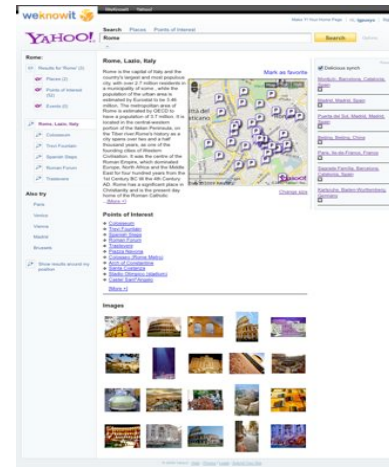


- >> Media Collection: flickr, query logs,
- >> Automatic generation of ranked facet lists of POIs

Media Intelligence



- >> Hybrid Clustering
- >> Image Localisation
- >> Tag suggestions



Travel Preparation



Personal Intelligence



Profile of contributor
>> Recommendations



Social Intelligence

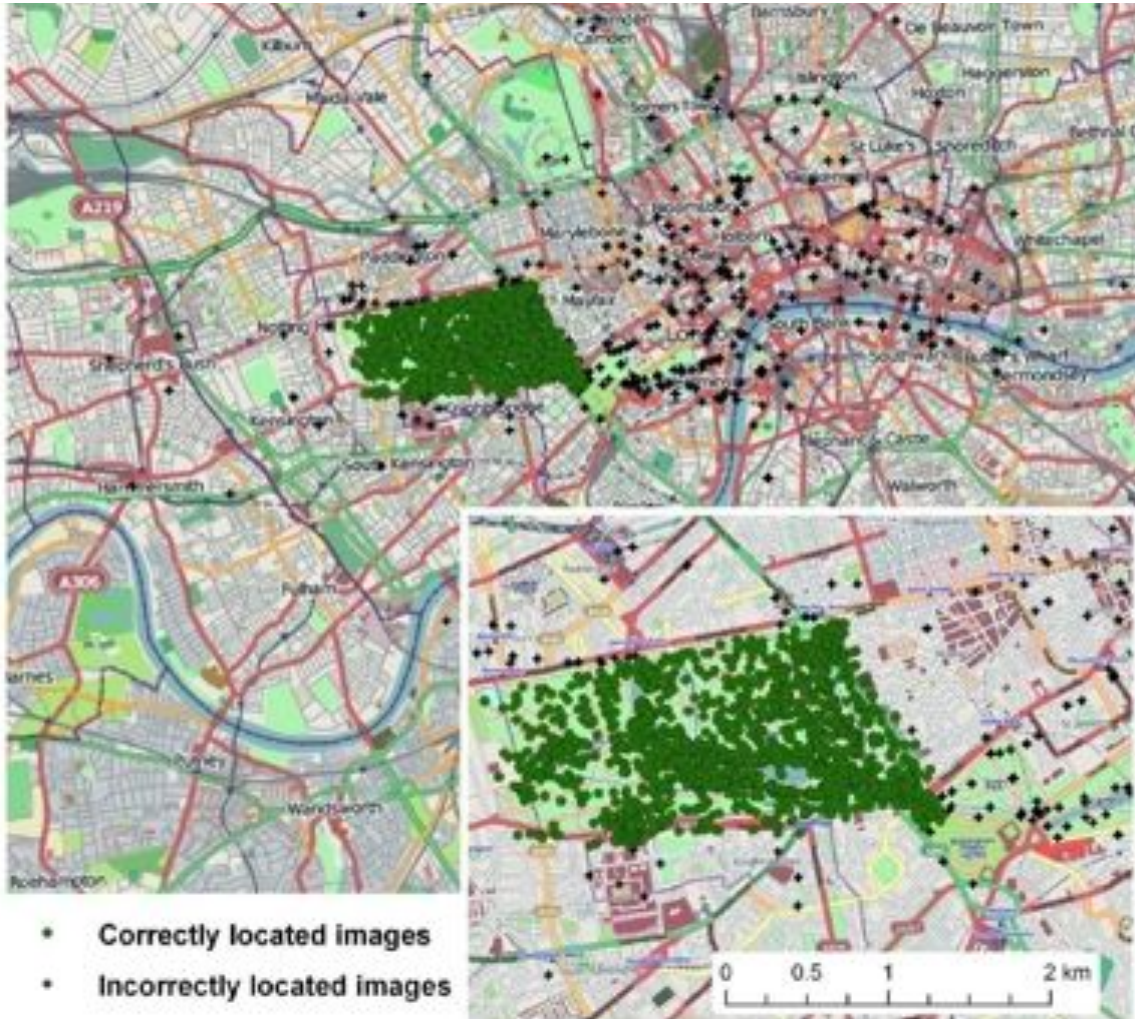
Can your social network be of help?

>> Friends position, alert



Mobile Guidance

Relevant Activities (*ER*)



Automatically describe city cores

Distinction between administrative and vernacular uses of place names

Potential for confusion in the dispatch of emergency services

Livia Hollenstein and Ross S. Purves, "Exploring place through user-generated content: using Flickr to describe city cores", JOURNAL OF SPATIAL INFORMATION SCIENCE

Relevant activities



MIT Center for Collective Intelligence

<http://cci.mit.edu/index.html>

The Climate Collaboratorium

Collective prediction ...*accurate predictions about future events such as product sales, political events, and outcomes of medical treatments....*

Collective intelligence in healthcare

Studying collective intelligence in today's organizations

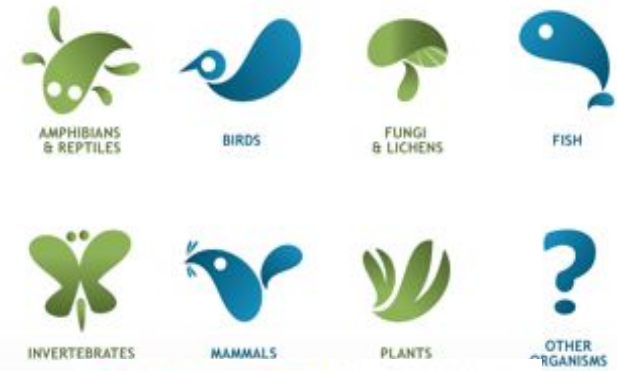
Relevant activities



<http://traffic.berkeley.edu>



Boston Citizens Connect

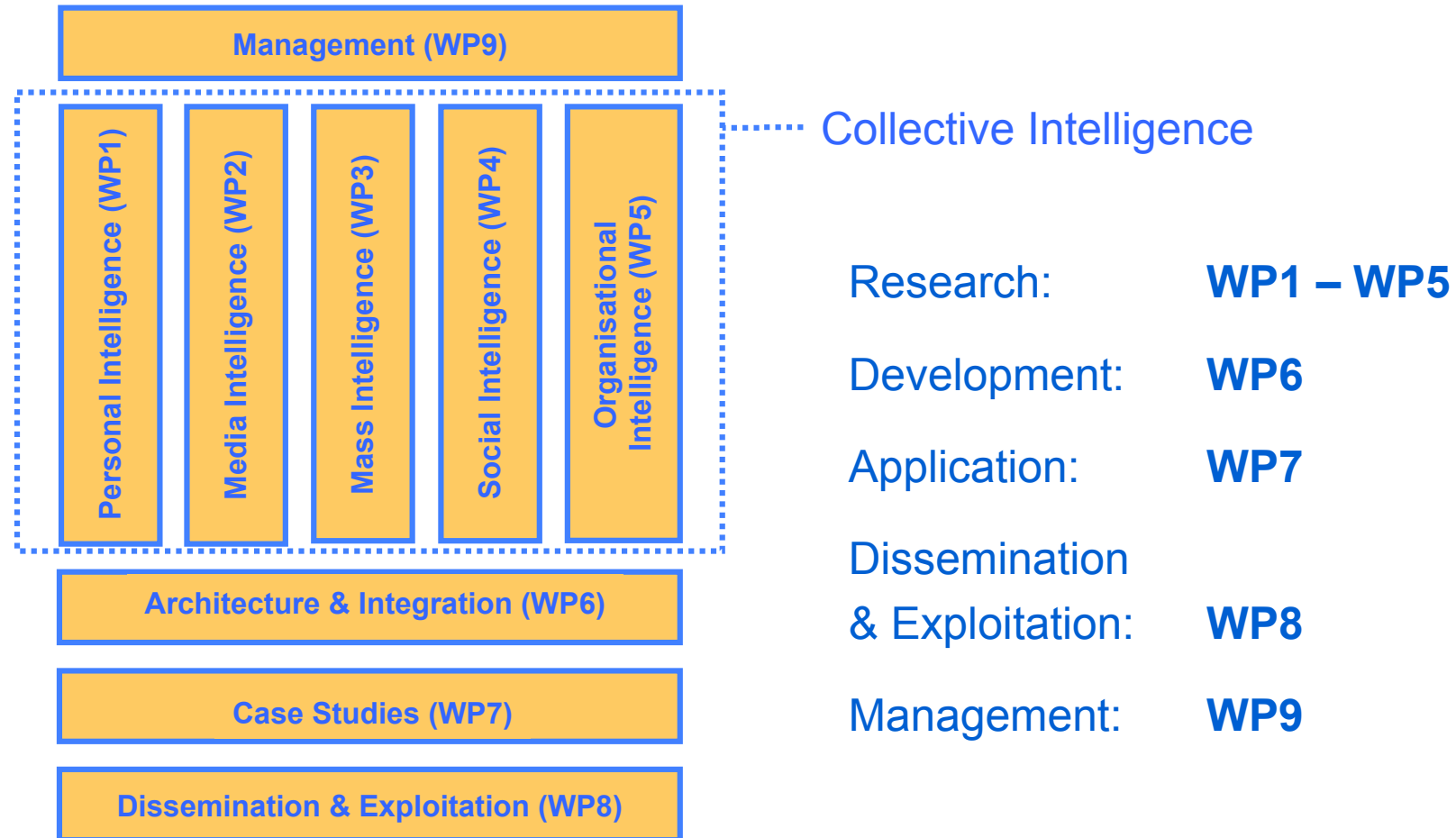


Dopplr helps you share your ... travel ... and exchange tips ... presents this **collective intelligence** - the travel patterns and advice ... as the Social Atlas.

Relevant activities

- Most applications are still harnessing Collective Intelligence
 - Upload applications
- Emphasis is on visualization of results
- Few focus on analysis
- No fusion of modalities – sources
- Unlimited applications
 - Draught detection, through estimation of green levels in flickr photos for fire protection in Spain (MIT Senseable lab)
 - Hollywood stock Exchange – HP Labs

Project work Overview



Content in WeKnowIt

offline → model creation, training

Non-Web 2.0 training data

Standard annotated corpora used for training.

- **Single-modality:** text (Brown corpus), speech (TIMIT database), image (Corel database)
- **Single-source:** prepared by a single person/organization
- **Consistent quality:** absence of spam, malicious or erroneous data
- **Small-moderate volume:** Manually produced

Massive Web 2.0

Massive user generated content and feedback from Web 2.0 applications

- **Multi-modality:** e.g. image + tags, image + geo-location + time
- **Multi-source:** may be generated by different applications, user communities, e.g. delicious, StumbleUpon and reddit are all social bookmarking sites
- **Inconsistent quality:** noise, spam, ambiguity
- **Huge volume:** Massively produced and disseminated

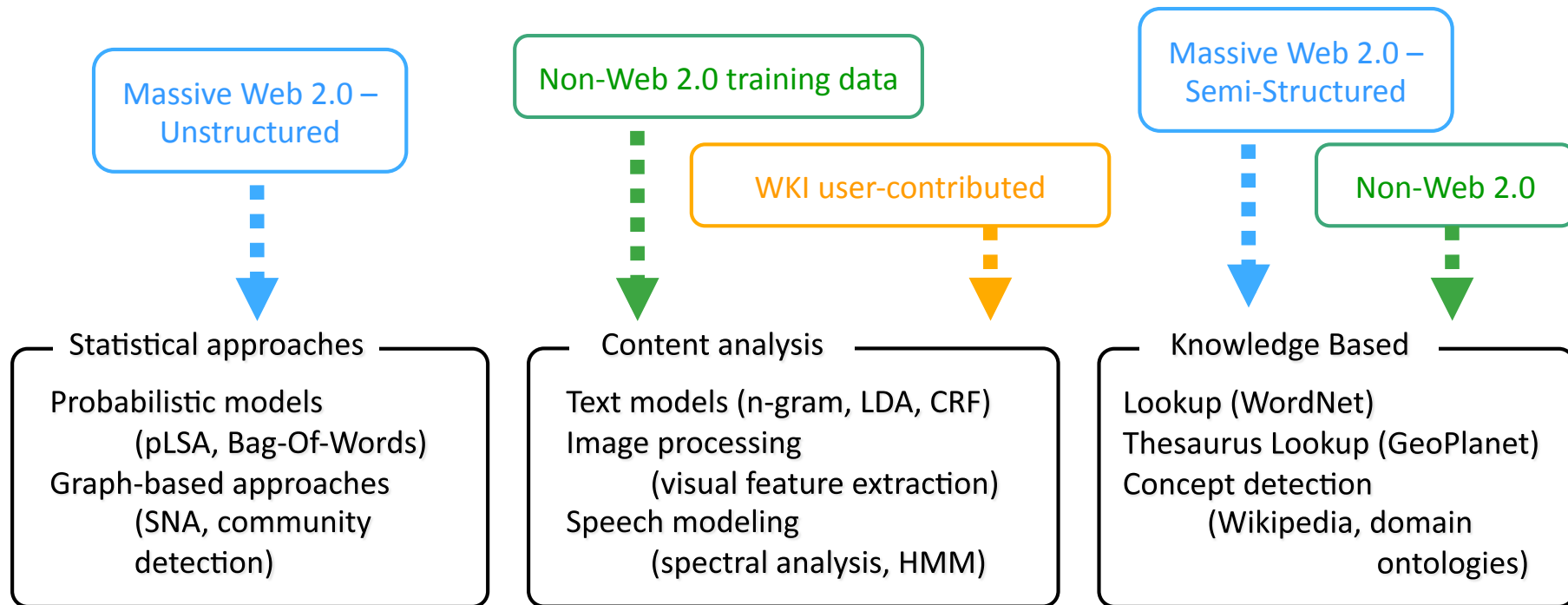
online → user profiling, method invocation

WKI user-contributed

Online content and user actions by WeKnowIt users. It is mainly used for triggering WeKnowIt services and for providing context to them, e.g. user profile, input content to be used as example for querying, etc.

Technical approach

Variety of approaches depending on content-metadata input.



Massive → Collective Intelligence

Multi-Modal (Fusion) → Combined CI

Inputs

Massive Web 2.0
(publicly available)

WIKI user-contributed

Non-Web 2.0 training data

Processing

Locations

- WP1: Get recommendations
- WP2: Visual analysis
- WP2: Text annotation
- WP3: POI recommendation
- WP3: POI clustering
- WP3: Search place POI
- WP5: csxPOIs

Topics

- WP2: Tag normalization
- WP2: Tag processing
- WP3: Text classification
- WP3: Hybrid image clustering
- WP3: Local tag community detector

Social connections

- WP4: Emergency alert service
- WP4: Community analysis tool

Events

- WP2: Speech search
- WP2: Semantic photo query
- WP5: Log merger
- WP5: Semaplorer(++)

Entities

- WP3: Named entity detection
- WP3: Entity facet extraction - ranking

Persistence

Representation

- WP1: CURIO
- WP1: VERACITY
- WP5: Event model F + M30
- WP6: Common data model

Storage

- WP2: Speech Indexing
- WP6: Data Storage

Access

- WP1: Account Manager
- WP1: Login
- WP4: Community administration platform
- WP5: Group Management

Application

GUI

- WP1: Manage Item
- WP1: Comment
- WP1: Tag
- WP1: Users messaging
- WP1: Search Knowledge Base
- WP6: System Integration
- WP3: Lexical Spam Detector

ER

- Mobile app
- Desktop proto
- Post ER tool

CSG

- Travel preparation
- Mobile guidance
- Post-travel logging



WP1: Personal Intelligence

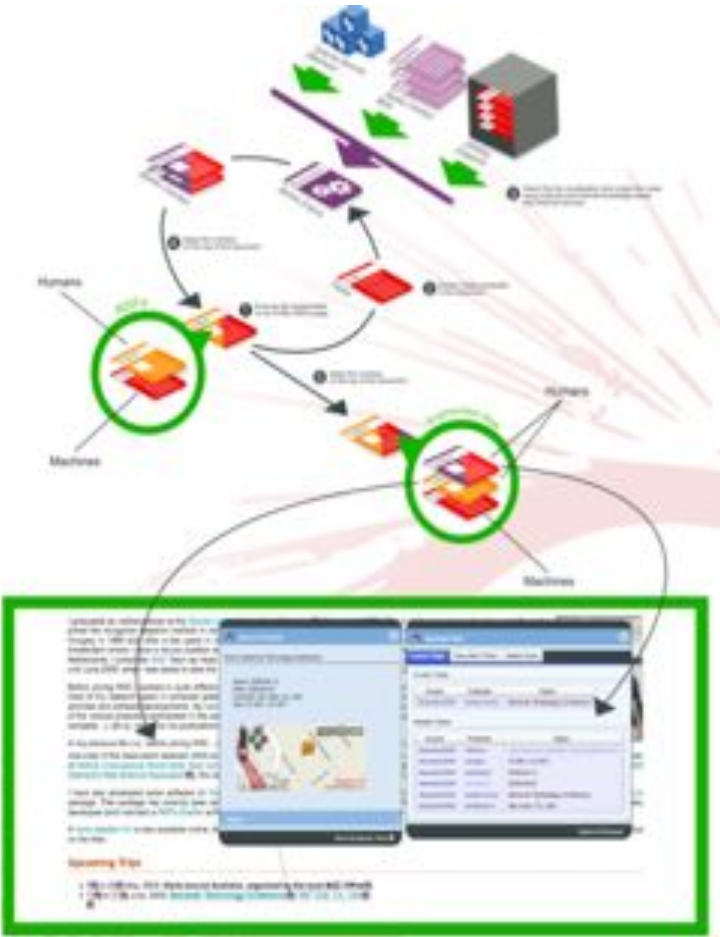


Access - Touchpad

CURIO Interaction Ontology Verosity (WP1/WP5)



Sparks: Semantically Aware Visualization Framework Context – Attention Streams



WP2: Media Intelligence

Visual analysis – Localization - Clustering



Rome, Lazio, Italy

Results for 'Rome' (3)

- Places (2)
- Points of Interest
- Events (0)

Rome, Lazio, Italy

- Colosseum
- Trevi Fountain
- Spanish Steps
- Roman Forum
- Trastevere

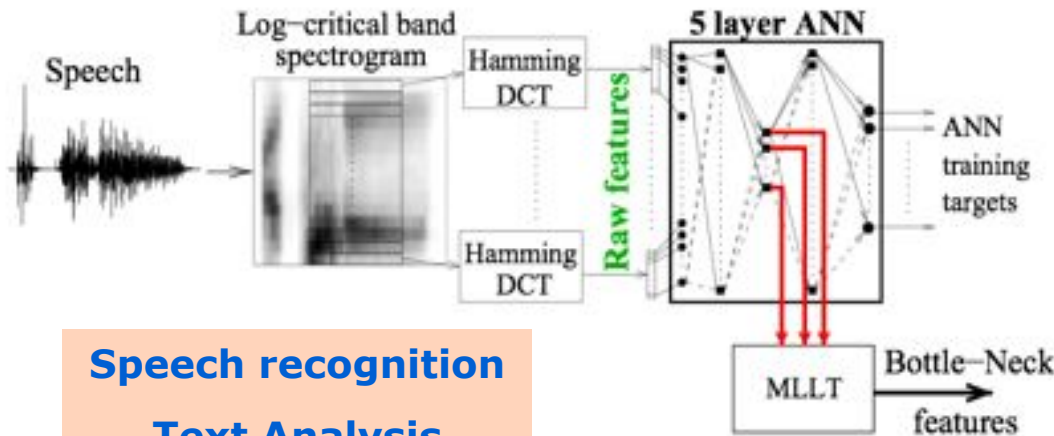
Also try

- Paris
- Venice
- Vienna
- Madrid
- Brussels

Points of Interest

- Colosseum
- Trevi Fountain
- Spanish Steps
- Roman Forum
- Trastevere
- Piazza Navona
- Colosseo (Rome Metro)
- Arch of Constantine
- Santa Costanza
- Stadio Olimpico (stadium)
- Castel Sant'Angelo

Map showing the location of Rome, Lazio, Italy.



Speech recognition
Text Analysis

Ranked entities

Combined CI Services

WP3: Mass Intelligence

Clustering – Community Detection

visual



tag



visual + tag



microsoft 0.139		girls 0.110		deportivo 0.077
teched 0.136		show 0.108		sports 0.076
autumn 0.132	graffiti 0.199	sexy 0.108	3gsm 0.408	camp 0.068
student 0.132	streetart 0.199	faceb 0.108	fira 0.375	soccer 0.065
partners 0.132	art 0.199	erotic 0.107	fair 0.273	lacoru 0.055
msp 0.132	street 0.199	model 0.107	trade 0.058	futbol 0.054
nov 0.125	mtn 0.012	festival 0.107	montjuic 0.051	liga 0.052
water 0.031	happy 0.012	faceb0 0.107	fountains 0.037	deportes 0.041
bycicle 0.017	blameles 0.011	tattoo 0.105	amazing 0.029	barca 0.038
food 0.010	montala 0.009	fishnet 0.019	show 0.027	depo 0.036
	parr 0.007		night 0.012	camprou 0.036
	spraint 0.003		water 0.011	

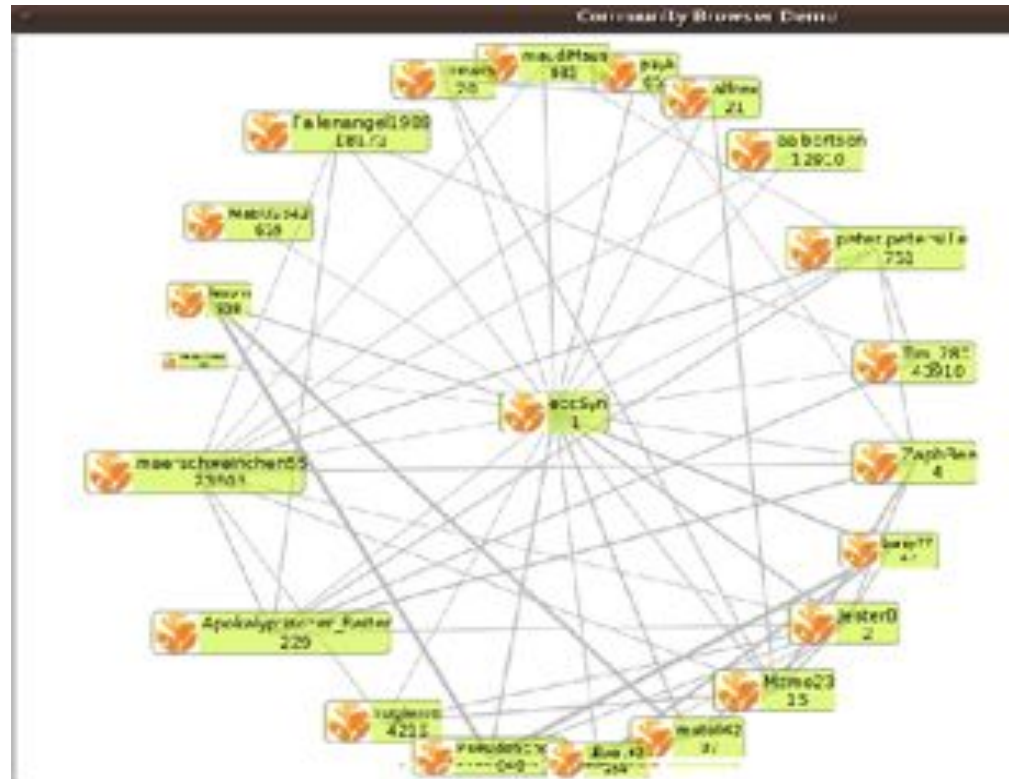
Event detection

CF	Normalized CF	Structural Similarity
Heineken Music Hall (neighbors: 172)		
Le Zenith (11)	Plaza d. Toros de Valencia (0.25)	Le Zenith (0.6201)
Forest National (7)	Le Zenith (0.1078)	Halle Tony Garnier (0.5244)
Werchter (5)	Belgrade Fair-Hall (0.0909)	Principe Felipe Arena (0.4878)
Lotto Arena (5)	Fonix Hall (0.0667)	Rockhal (0.4672)
Oslo Spektrum (4)	Samsung Arena (0.0625)	Werchter (0.4590)
Cardiff International Arena (4)	Paradiso (Amsterdam) (0.0526)	Coliseu dos Recreios (0.4590)
Skandinavium (4)	Bang Your Head! (0.0455)	Coliseu de Porto (0.4497)
Melkweg (4)	Hala Rondo (0.0455)	Killesbergpark (0.4462)
Degree filtered: France, Paris, Switzerland, Brussels, Madrid, Czech Republic, Lisbon, etc.	Degree filtered: France, Switzerland, Paris, Brussels	Degree filtered: -
Madame Tussauds (neighbors: 346)		
Alton Towers (10)	Rock Circus (0.2857)	Historic House Trust (0.3413)
Thorpe Park (9)	National Wax Museum (Ireland) (0.1429)	Hudson River Maritime Museum (0.3409)
London Eye (8)	The Amsterdam Dungeon (0.1)	Mabee House (0.3366)
Chessington World of Adventures (7)	Fort Decker (0.0833)	Johnson Hall State Historic Site (0.3328)
Baker Street (5)	Glaspalast (Munich) (0.0714)	Empire State Railway Museum (0.3298)
Legoland (4)	Johnson Hall State Historic Site (0.0714)	
Natural History Museum (4)		
Degree filtered: New York, Victoria and Albert Museum, Buckingham Palace, Westminster Abbey, London	Degree filtered: -	Degree filtered: -

PoI recommendation

WP4: Social Intelligence

Emergency Alert Service



Community administration platform
Community analysis tool
Community browser

Emergency Alert
for
Michael Ovelgoenne, +49 967 6543210
reported by
Andreas Sonnenbichler +49 123 4567890
at 2010-10-03 16:45h CET

An emergency situation was reported by Andreas Sonnenbichler. Although Police was notified, the EAS calculated you to be closest to Andreas. Andreas is in need - please hurry and help him! Andreas transmitted his current geo-position (B). Please use the map to find him. Your position is marked with (A). This is not a joke!

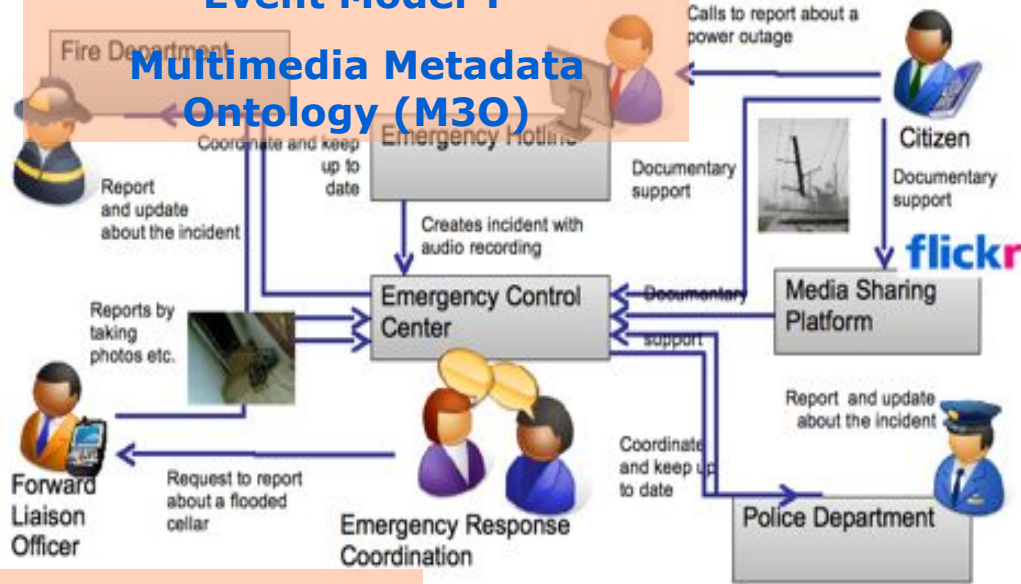


Emergency Alert System (EAS) is a mobile safety service provided by the **ruhrh Institute of Technology**. You have been notified as the EAS system calculated you to be part of Andreas's social network.
Contact information: Karlsruhe Institute of Technology, Institute of Information Systems and Management, Prof. Dr. Geyer-Schulz, Phone: +49-721-608-8402

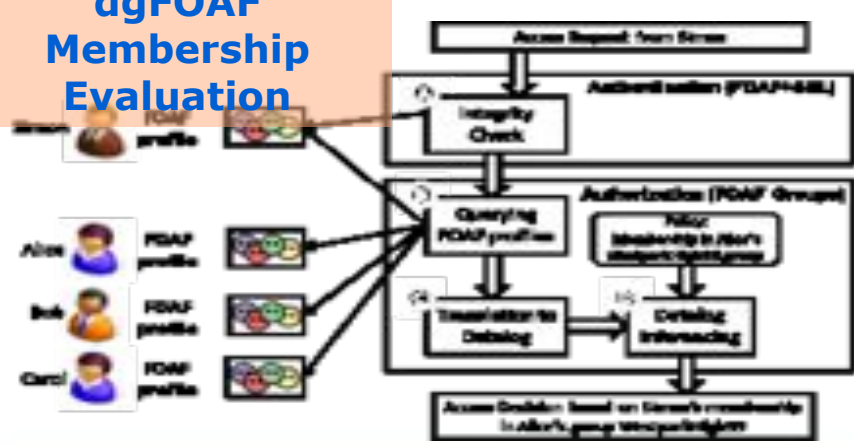
WP5: Organisational Intelligence

Event Model-F

Multimedia Metadata Ontology (M3O)



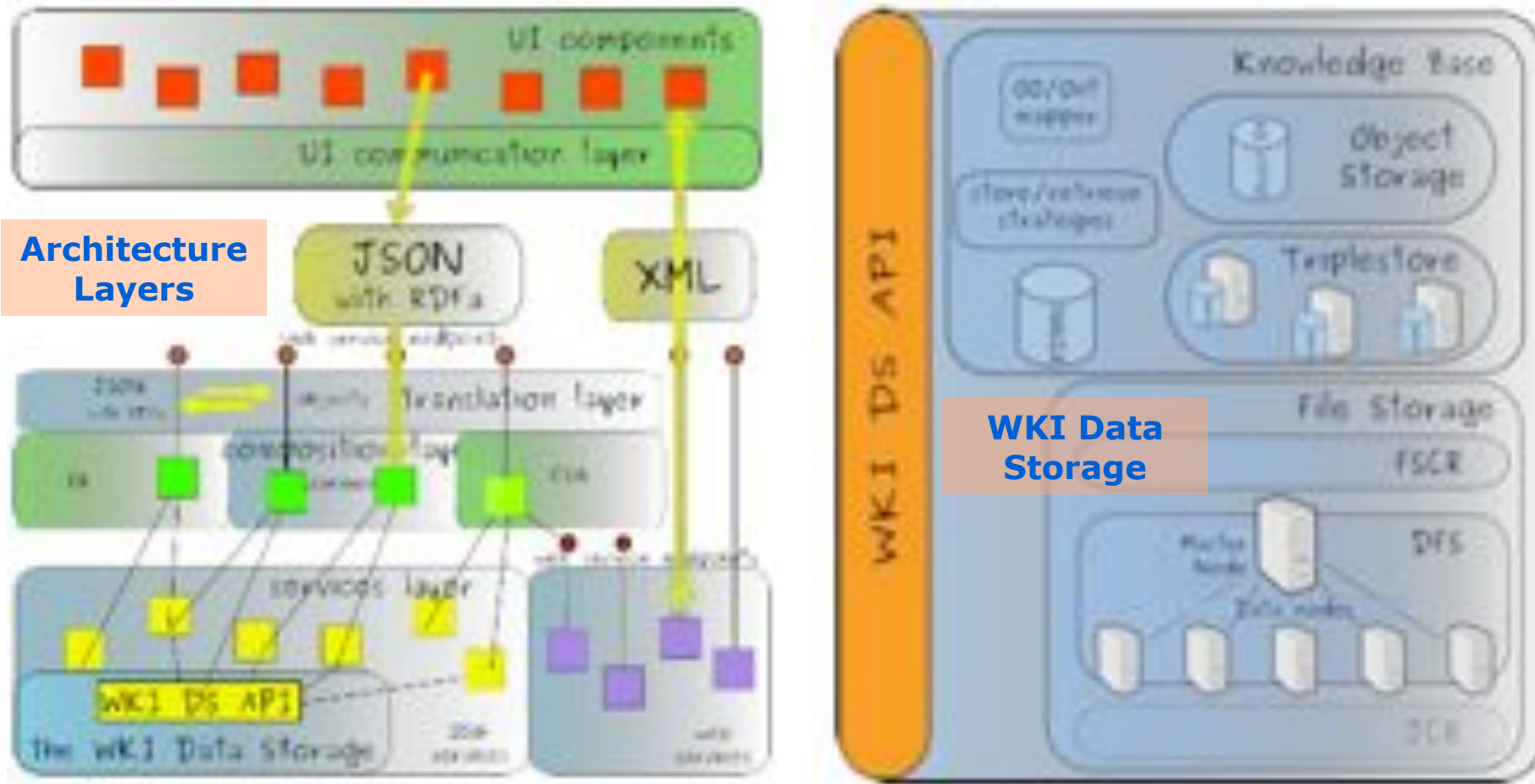
dgFOAF Membership Evaluation



Time	From/To	Action	Message
15 Sep 2006 14:55	From South Yorkshire Police	Informed Philip Horton	Chemical incident at ENPAR, Ecclesfield. Pitric acid - explosive if subjected to heat, friction or shock - found in store. Building evacuated but a large site. Just for info. At this time.
15 Sep 2006 15:56	South Yorkshire Police	Where on site? Liz Bashforth informed	Bomb Disposal evacuating 3-400 - request F.L.O. South Yorkshire Police OIC - RVP to be confirmed. Believed to be houses - 2 rows terrace houses. Pub on Nether Lane - Meadow arm Pub. Nursing Home Nether Lane - is it affected.
15 Sep 2006 16:17	To Graham Smith		As above Pitric Acid Gave my numbers @ Ecclesfield with Gerg Jambor.
15 Sep 2006 16:22	CYPD	PM informed.	Which school - @ Ecclesfield C/T to open up easily.
15 Sep 2006 16:55	Eddie Sherwood	Request police presence in centre until fully set up.	Confirm @ Ecclesfield Sec. WRVS Church has activated. 3 shifts planned - 5x12, 12x6, 6x12.
15 Sep 2006 16:56	To George Hancock		Have enough coaches for 280 - we said that his is now enough to side of @ Ecclesfield police station.
15 Sep 2006 17:05	From Leah Barratt	Offered further support via LB to call her.	Has contacted @ Ecclesfield caretaker. Business Manager. Veronica Hill - PFI team. Can't get in touch with EDEN food rep - only a dep manager.
15 Sep 2006 17:13	To Gerg Jambor		At @ Ecclesfield Police station believes Nightingale Home is not in cordon.
15 Sep 2006 17:20	Angela Rowland via telephone	Asked to go to Longley.	Call out re incident at @ Ecclesfield.
15 Sep 2006	From Graham Smith		At @ Ecclesfield with Gerg Jambor Police have started door knocking now to incident control case. Meadow Farm with 8 issues to Paul M...

Log Merger

WP6: Architecture & Integration



30 services integrated

WeKnowIt Community Detection

Challenges in Social Media network mining

No prior assumptions about structure:

Complex & evolving structure

No possibility for knowing structural features (e.g. number of clusters on a graph) in advance

→ Unsupervised

Scale

Tens of millions of active users frequently contributing loads of content links + metadata (tags, comments, ratings)

→ Efficient - scalable

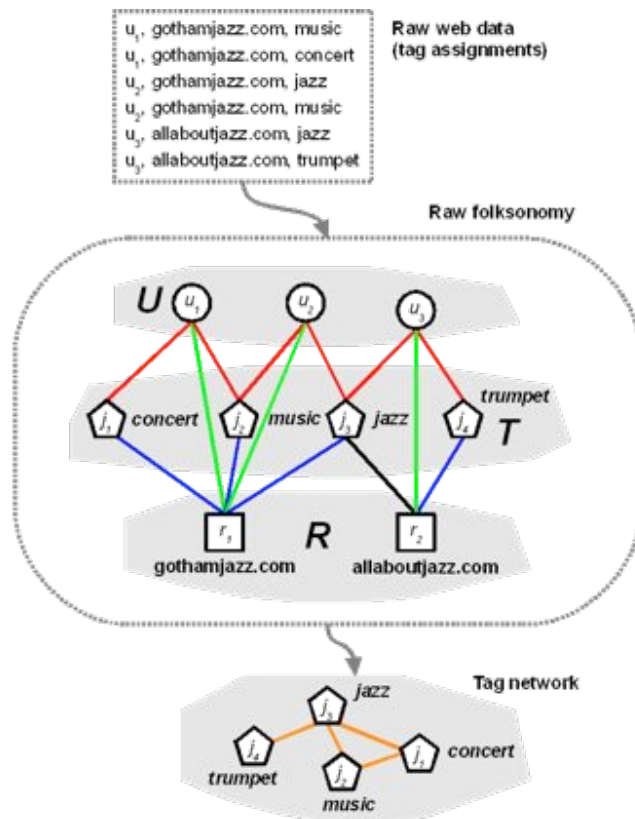
Quality

Spam is very common. Only a portion of user contributions is worth further analysis.

→ Noise resilient

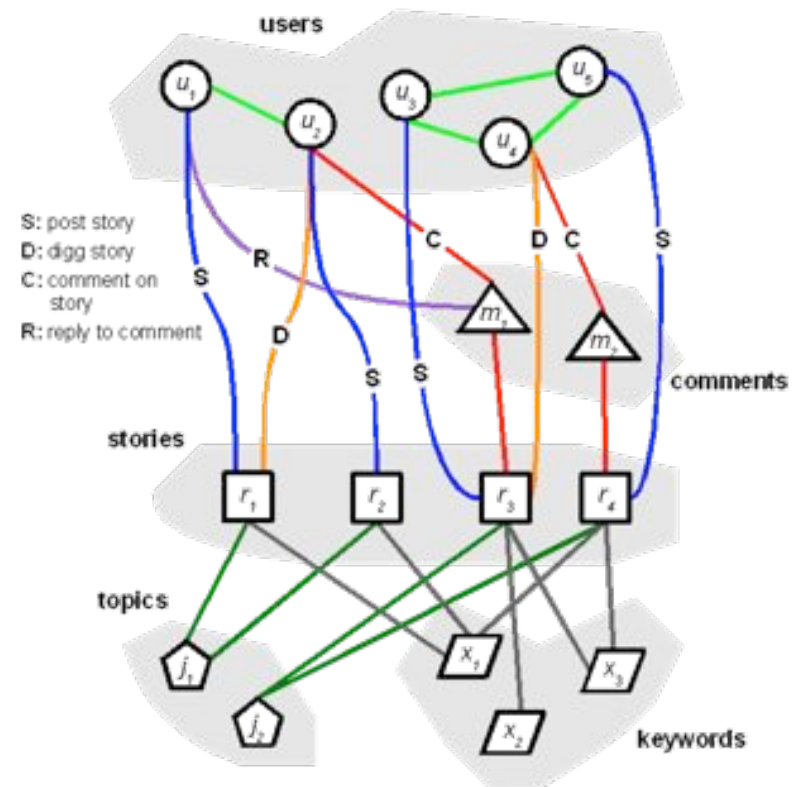
Examples of Social Media networks

Folksonomy (Delicious)



Mika, P. (2005) Ontologies Are Us: A Unified Model of Social Networks and Semantics. Proceedings of the 4th International Semantic Web Conference (ISWC 2005), Springer Berlin / Heidelberg, pp. 522-536

MetaGraph (Digg)



Lin, Y., Sun, J., Castro, P., Konuru, R., Sundaram, H., and Kelliher, A. (2009) MetaFac: community discovery via relational hypergraph factorization. Proceedings of KDD '09, ACM, pp. 527-536

What is a community in a network?

Group of vertices that are more densely connected to each other than to the rest of the network.

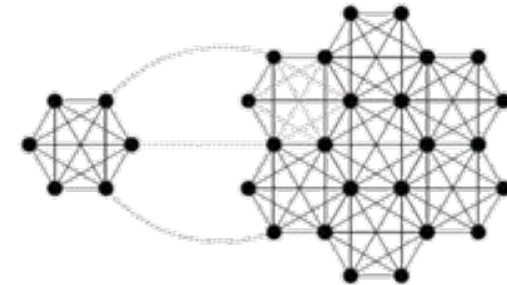
Multiple definitions to quantify communities:

Fortunato S. (2010) Community detection in graphs. Physics Reports 486: 75-174

Global: N-cut, conductance, modularity

Local: Local modularity, (μ, ϵ) -cores

Ad hoc: Label propagation, dynamic synchronization



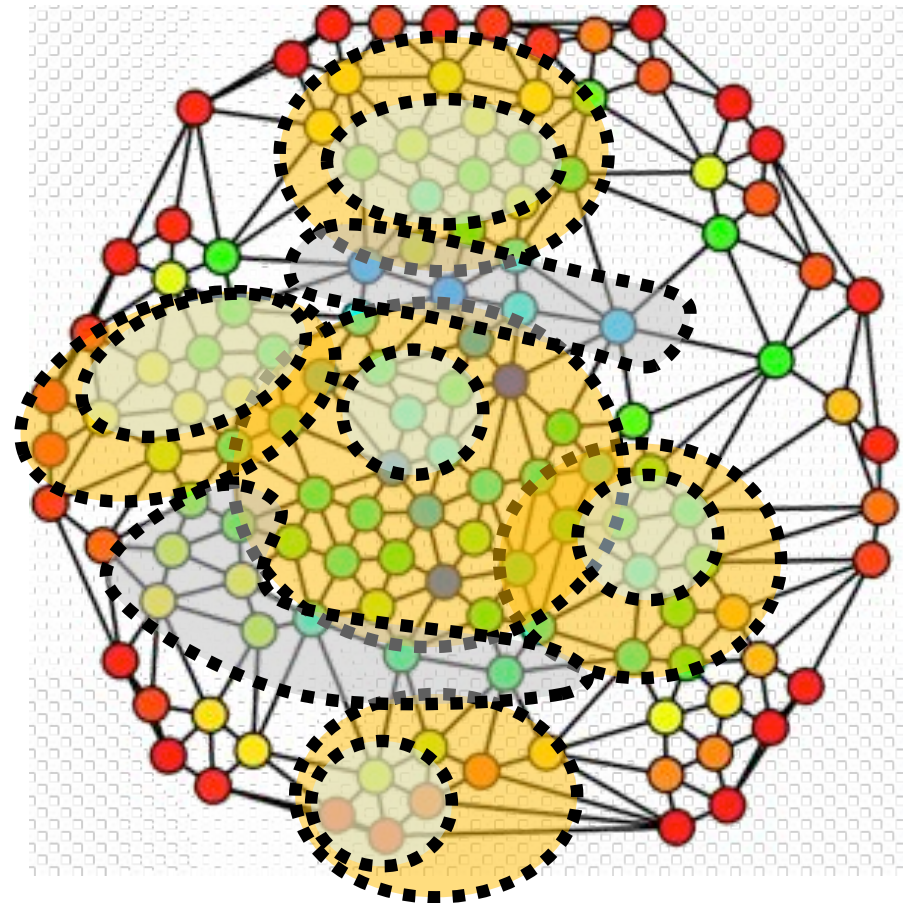
Related to clustering, but: (a) not necessary to know number of communities, (b) computationally more efficient

In WeKnowIt, we focus on local definitions, because of the properties of Social Media networks: efficiency-scalability and noise resilience.

Approach illustration

Two-step process:

- 1st step:
 (μ, ε) – core detection
- 2nd step:
Local expansion
- 3rd step:
Characterization of remaining vertices as *hubs* or *outliers*



Hybrid Photo Clustering

Goal:

- Group large photo collections into clusters based on how much they are related to each other
- Assist browsing and navigation by means of a map-based application
- Detect landmark and event clusters.

Combine both visual features *and* tags

- Two kinds of similarity (visual and tag networks) are complementary to each other
- Many times one photo has missing tags or is hard to interpret visually
- Graph-based approach - superimpose visual and tag graphs
- Use photo cluster features for classification to landmarks/events

Results

- Higher quality clusters by use of both visual and tag similarity instead of only each one of them.
- Clusters can be used for landmark and event detection.
- Integrated in CSG prototype and ClustTour stand-alone demo.

Overview of approach

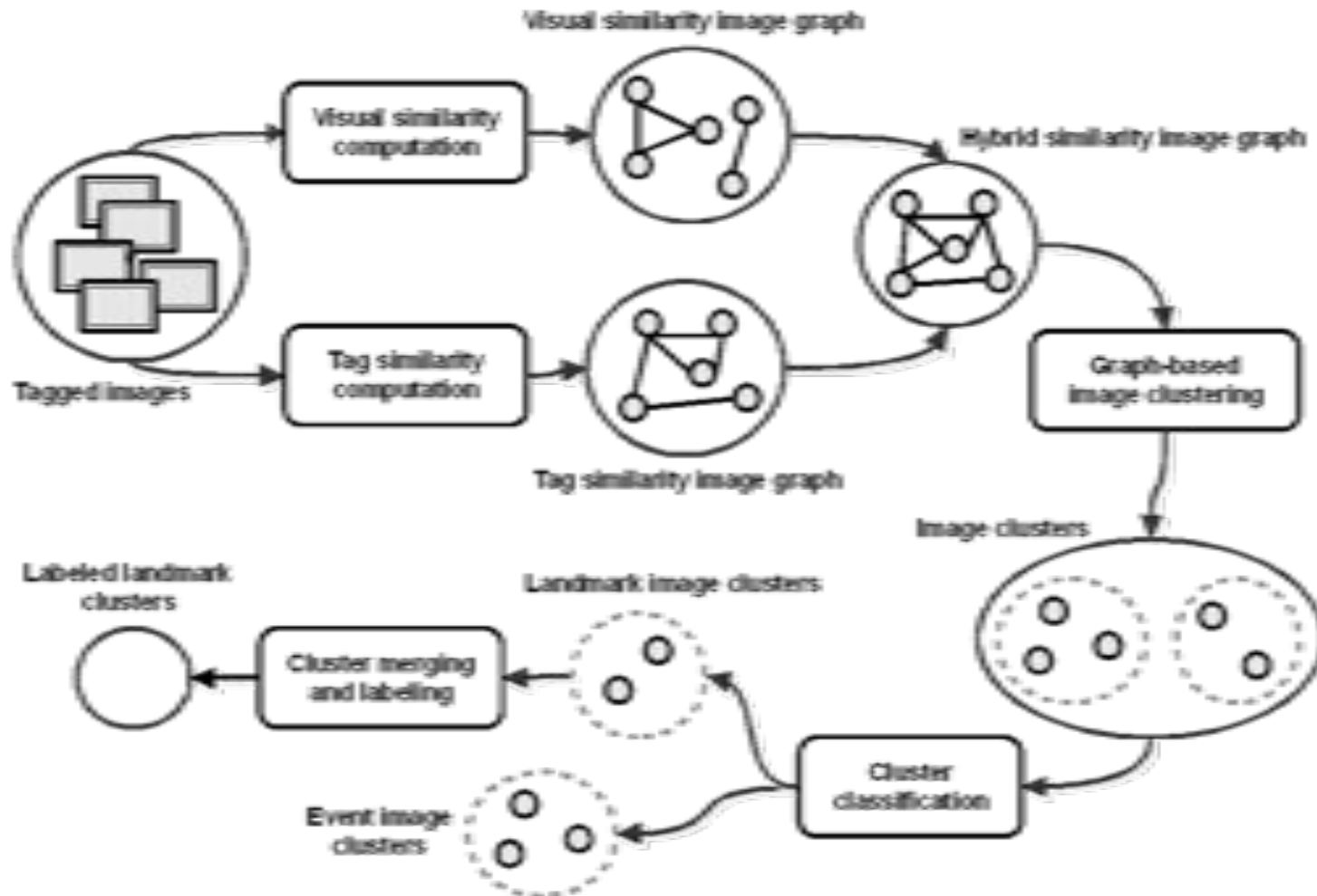


Photo clustering results (1)

User study (involving 20 people)

Users were shown photo clusters and they were asked to judge how relevant the photos of each cluster were related to each other

Each cluster was produced by different notion of similarity (tag-only, visual-only, hybrid). Obviously, users were not aware of this information

Hybrid clusters were found to be of superior quality (highest F-measure)

Algorithm	Precision	Recall	F-measure	κ -statistic
SCAN-VIS	0.980	0.178	0.301	0.925
SCAN-TAG	0.910	0.197	0.323	0.688
SCAN-HYB	0.898	0.246	0.387	0.637
EXP-VIS	0.985	0.178	0.301	0.895
EXP-TAG	0.929	0.201	0.331	0.709

Photo clustering results (2)

Geographic localization of results was also found to be very high. Most clusters correspond to landmarks or events.



EVENTS



Sample results: [Visual] vs. [Tag] vs. [Visual + Tag]

VISUAL



HYBRID



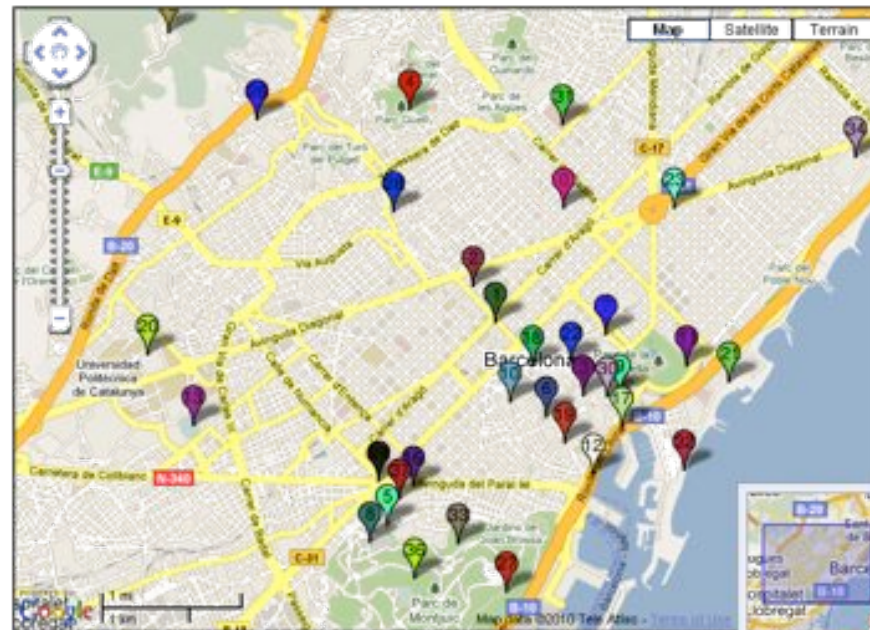
TAG



ClustTour demo: City exploration by means of photo clusters

ClustTour

weknowit 



Landmarks

Tags: sab, bmx, juanma

Tag list

- barris gòtics (3)
- museu (3)
- yellow (3)
- cathedral (2)
- churches (2)
- espanya (2)
- catedral (2)
- casa (2)
- passeig de gracia (2)
- el coll (2)
- mercado (2)
- arte (2)
- verde (2)
- placa (2)
- port vell (2)
- plata (2)
- domènec i montaner (2)
- diagonal (2)
- agua (2)
- apple (2)
- plants (2)
- ass (2)
- sagrada familia (1)
- sagrada (1)
- familia (1)
- catholic (1)
- construction (1)
- spire (1)
- kirche (1)
- casa batlló (1)

Time filter

From 31/10/2002 up to 1/8/2009



weknowit 



co-funded by the European Union

Travel demo

The screenshot shows a Yahoo! search results page for the query "china". At the top, the Yahoo! logo is on the left, and navigation links for "Search", "Places", and "Points of interest" are in the center. A search bar contains the text "china" and a yellow "Search" button is on the right. Below the search bar, the results are categorized under "china:". On the left, there are sub-categories: "Results for 'china' (2)", "Places (1)", "Points of Interest (51)", and "Events (0)". A list of results includes "China", "Terracotta Army", "Forbidden City", "Xinjiang", "Yunnan", "Tiananmen", and "Summer Palace". Below this is an "Also try" section with links to "Xlu0027an", "China", "Beijing", and "Changlu0027an", and a "Show results around my position" button. The main content area features a result for "Terracotta Army". It includes a title, a "Mark as favorite" link, and a map of China with a callout for "Terracotta Army". The text describes the Terracotta Army as the Terra Cotta Warriors and Horses of Qin Shi Huang, discovered in 1974 near Xi'an, Shaanxi province. It mentions that the figures, dating from 210 BCE, vary in height from 183-195 cm (6ft-6ft 5in). The text also notes that the figures include strong warriors, chariots, horses, officials, acrobats, strongmen, and musicians, and that over 8,000 soldiers, 130 chariots with 520 horses, and 150 cavalry horses were found in three pits. To the right of the map is a "Favorites" section with a "Delicious synch" checkbox and a list of items including "Parthenon, Athens, Attiki, Greece". Below the main text is an "Images" section with a grid of 12 small images showing various terracotta figures and the site. At the bottom of the page, there are logos for "weknowit", the European Union flag, and the "SEVENTH FRAMEWORK PROGRAMME" logo.

WKI Grand Travel Challenge

Barcelona, January 21st 2010



WKI Grand Travel Challenge

Barcelona, January 21st 2010





WeKnowIt Grand Travel Challenge (best of)

Group Pool Discussion 18 Members Map Invite Friends

Slideshow Share This

» Add something?



From [colibacari](#)



From [CosteC](#)



From [sacchi](#)



From [brother_joak](#)



From [Pavel_Smirz](#)



From [Pavel_Smirz](#)



From [scbm](#)



From [clabeta](#)



From [Ako_Pap](#)



From [SannaKb](#)



From [WebKnowIt Grand...](#)



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From [YanniKa](#)



From [lovetoy](#)

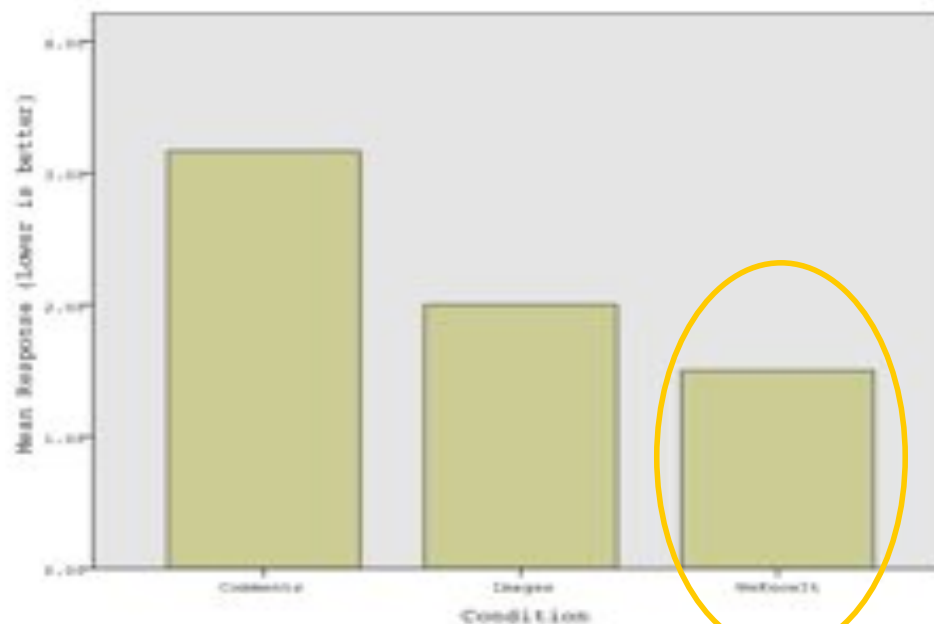


From [bockut.net](#)

Evaluation results

- 5 different evaluation activities
 - 1 ER specific
 - 2 CSG mobile
 - 2 CGS desktop
- 59 users
 - ER personell
 - Citizens
 - Users
 - TID personell
 - WKI consortium

"I found it easy to process all the information available to me"



Experts

"I felt I understood what was going on quickly"

Content - Emergency Response

- Text
 - Sheffield floods forum messages and posts (thousands)
 - Police & fire service logs
 - ABC news tagged articles (~7000 metadata files)
- Image
 - Flickr images + metadata
 - 136 related to June 2007 Sheffield floods
 - ~27K geo-coded photos around the area of Sheffield
 - 1400 ER images (after WKI clustering)
- Speech
 - 1000+ emergency phone calls on Sheffield flooding event
 - 1000+ voice-tagging events by at least 10 users
 - Fused text+speech dataset

Content - Consumer Social Group

- Text
 - Flickr – metadata from geotagged London images (4300+ files)
 - Wikipedia processing
 - GeoPlanet processing
- Image
 - VIRaL (1.2 million geotagged images - 22 European cities)
 - Barcelona meeting dataset - geotagged and tagged (647 images, 1669 tags)
 - 1000 restaurant images
- Social networks
 - Barcelona meeting network of contacts (14 users)

Research Fields and Issues

- Statistical analysis, machine learning, data mining, pattern recognition, social network analysis
- Clustering
- Graph theory
- Image, text, video analysis
- Information extraction
- Fusion techniques
- Trust, security, privacy
- Performance, scalability
 - speed, storage, power, grids, clouds

Conclusions

- Collective Intelligence can be extracted by social media
- New applications and services can be developed
- Fusion of multimodal – multisource info remains a challenge
- Scalability, quality, coverage are important issues