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3reast & Colon Cancers





Sentinel e-health network on grid: developments and challenges

Sentinelle Grid Network

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CONFERENCE





- Context: Cancer screening in France
- The Sentinelle project
 - > Objectives
 - > Architecture
- Specific issues and challenges
 - > 1 Data consistency
 - > 2 Patient identification
 - > 3 Data linkage
- Data linkage results and benchmarking
- Conclusion: further steps and future functionalities





Introduction & context





Screening

"Earlier is better"

3 types of cancers are concerned:

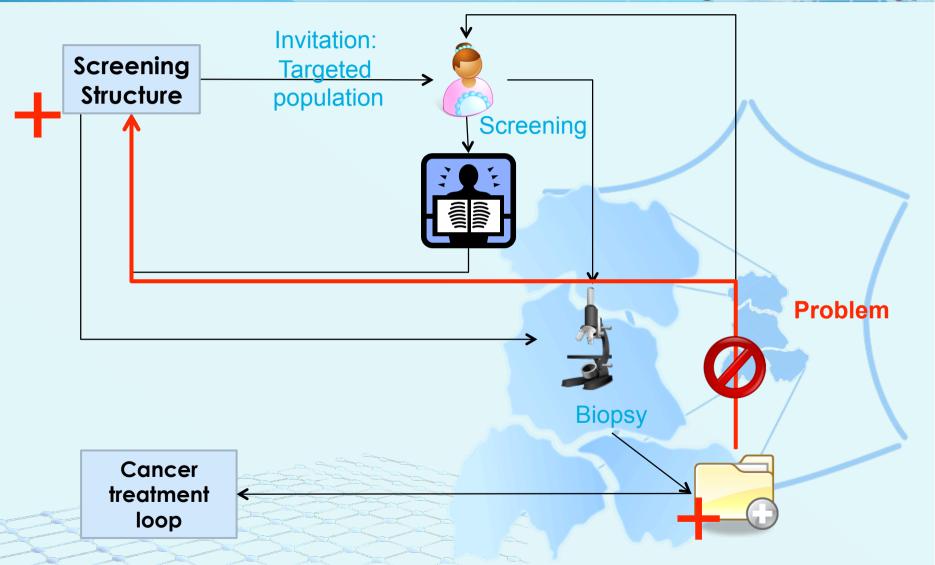
- > Breast, colon and cervical
- > Easy to screen, easy to cure

Cancer screening organizations and pathology laboratories have to exchange data to ensure follow-ups

- No electronic exchanges
- Pathologists refuses to export their data
 - Medical reports are printed/mailed/faxed
 - Re-recorded by the screening organization
- > Data are collected directly besides patients



Cancer screening: methodology



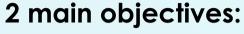


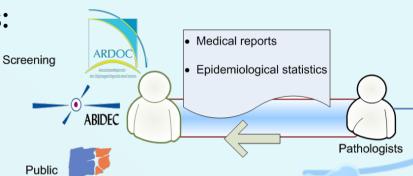


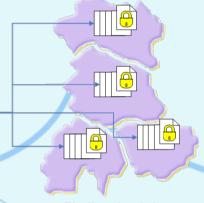
The e-sentinelle project



The e-sentinelle project







Distributed pathology databases

- Improvement of medical data sharing:
 - > Improve and ease collaboration
 - > Speed-up treatments
 - Ease follow-up

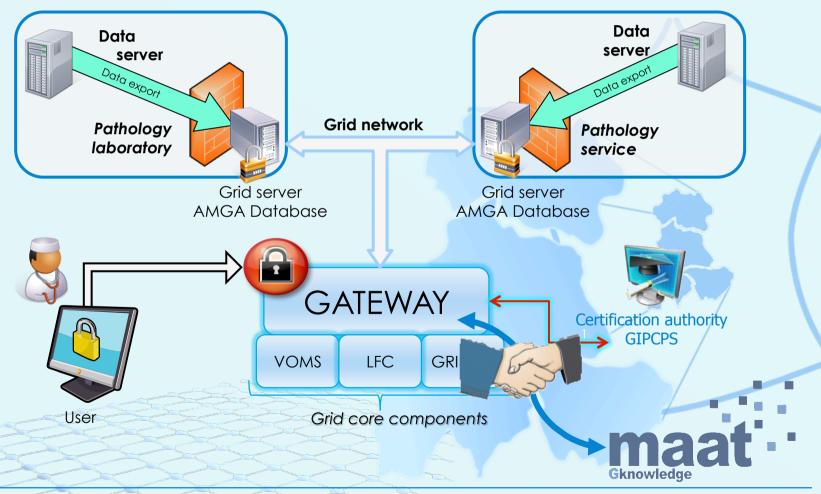
Regional grid network for epidemiology:

- Large-scale epidemiology analysis
- > Disease monitoring and alert specification (public health)
- Health requirements meets grid requirements





• A lightweight grid architecture:





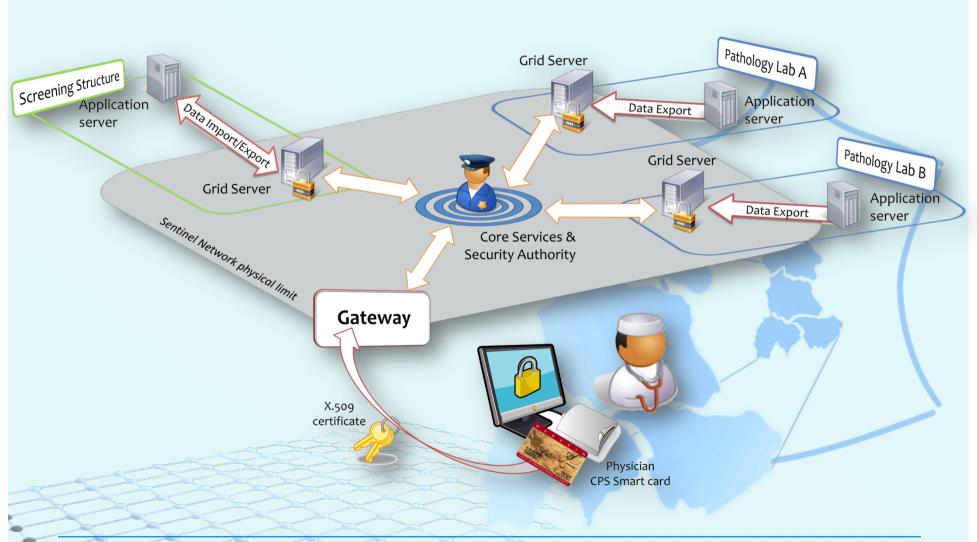
Privacy and security requirements

• Query medical databases on demand:

- No massive extraction of databases
 - Respectful of data ownership
- Respectful of patient privacy
 - Compliant with data processing laws
 - Use of cryptographic algorithms
- > Guarantying all security requirements
 - Using grid security layers
 - Strong authentication methods



Project architecture







Specific issues and challenges

- 1 Data consistency
- 2 Patient identification
 - 3 Data linkage



1 - Data consistency

 Consistency over distributed databases is fundamental in order to be able to query them

Several steps are needed:

Data extraction

 Mechanisms (automatic or semiautomatic)

Data integration in the grid network

- Export parsing and integration
- Adjust rights and ownership according to local policies



• Semantic annotation on fields

Dataset standardisation

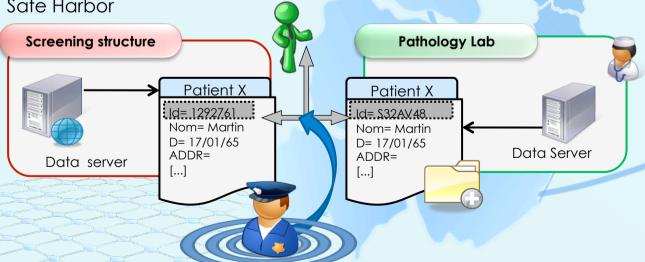
• Fields specification



2 - Patient identification

• Problem:

- > Identification system is strictly regulated in France:
 - NSS number usage prohibited
 - Not always significant
- > Security constraints:
 - Medical data transfer over a network is regulated by data privacy laws:
 - National laws (French Informatique & Libertés, UK Data protection act, German BFDI...)
 - EU directive 95/46/EC
 - US-EU Safe Harbor





2 - Patient identification

• Challenge:

- Offer a decentralised identification system able to merge distributed patient identities
- > Respect privacy
- > Comply with third party identification sources

• A new identifier is added to patients for each database:

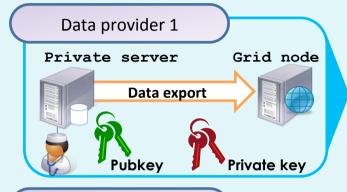
- uuid-based: a 128bit long (as much as possibilities as ipv6 addresses)
- > Randomly generated ex: 2fffb5aa-07f0-47bf-bb1e-90fda3bada14
 - Statistically unique and anonymous

Identification system

Consists in merging distributed identities through a centralised server

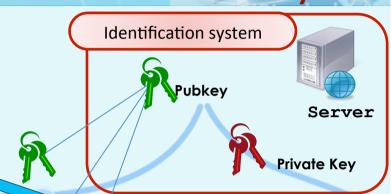


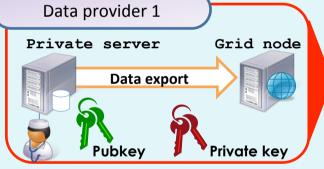
2 - Identification system



Patient X ID=**12002110**RSCAId= **a5b**[...]

RSCAId= a5b[...]c43a N= Martin D= 17/01/65 ADR= Street X [...]



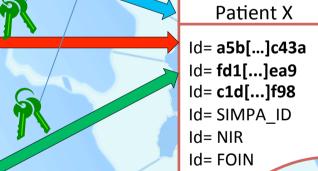


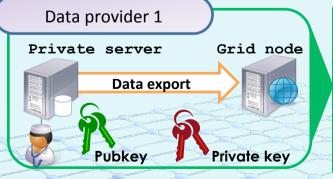
Patient X

Id=1D72C1 RSCAId= fd1[...]ea9 N= Martin D= 17/01/65 ADR= Street X

[...]

[...]





ID=**0192298**RSCAId= **c1d[...]f98**N= Martin D= 17/01/65 ADR= Street X

Patient X





2 - Patient identification solution

• Advantages of the solution:

- > Scalability: patient identity can easily evolve
- Privacy: no common identifier exists in distributed databases which remain independent
- Identity management: in case of double identification or two patients with the same number, merging and separating identifiers is straightforward
- > Interconnectivity: easy to add another identification source

• Drawbacks:

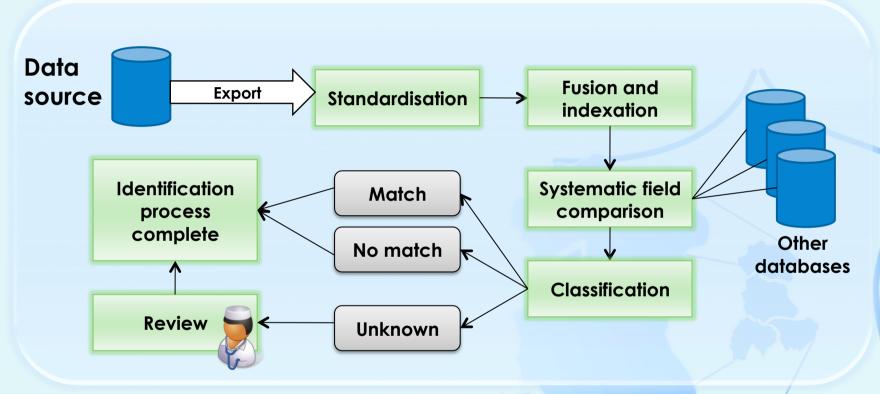
- SPOF in the central server
- Lot of encryptions/decryptions
- 1st phase of identification is costly
- > Need a good data linkage process



- No common identifier exists => patients have to be compared on their own characteristics :
 - > Name / Surname /Sex
 - > Birthdate
 - > Address
- The data linkage process must be as efficient as possible:
 - > Good automatic decision
 - Restricting double-counting (False negatives or type I errors)
 - Avoiding false matching (False positives or type II errors)
- Data linkage ≠ strcmp in C++
- Reasons: Look alike patients
- Need high level comparison algorithms:
 - Measure of similarity between two strings



3 - Identification workflow



- If score > high_thres_identities are merged
- If score < low thres a new patient is created</p>
- Tricky patient: manual intervention is needed



3 - Data linkage algorithms

2 families:

- String similarity methods:
 - Jaro-Winkler Algorithm

$$O(\sigma_1, \sigma_2)$$

- Calculation of a sum of exact characters matches and permutations.
 - Very efficient for names
- Rupert / Robert = 0.45
- Martha / Matrha = 0.85

- Phonetic-based methods
 - Soundex algorithm
 - Calculates a phonetic codification of strings according to a transformation table:

Class	1	2	3	4	5	6
Letters	BFPV	CGJKQSXZ	DT	L	ΜN	R

- > Highly language sensitive
- Efficient for spelling and phonetic errors
- Rupert / Robert = 1.0
- Martha / Matrha = 0.52



3 - Benchmarking methods

• Experiment configuration:

- > Two subsets of 10000 unique records with 12% of joint records
- > Incorporation of random bias in the datasets:
 - Remove/add 0-4 letters
 - Permutations of 0-2 letters with keyboards neighbours (50%) or random letters (50%)

• Parameters:

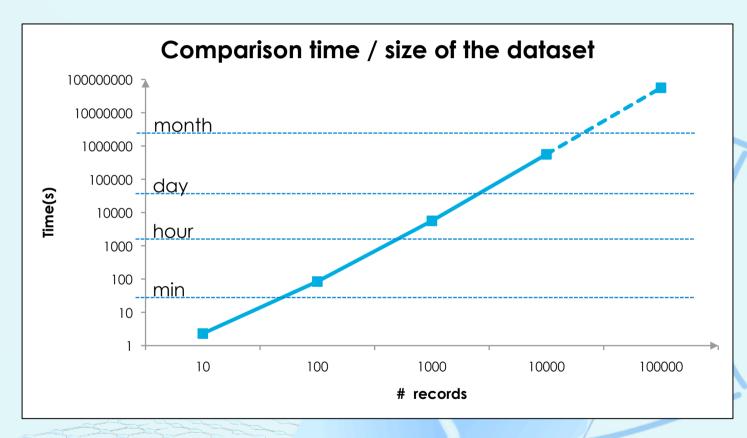
 \rightarrow Threshold = 0.8

Results:

Field – Method	TP	FN	FP	Result	Accuracy
Last Name – Jaro-Winkler	11.53	1.21	0.06	96.08	90.08
Last Name – Soundex-US	9.33	1.14	0.11	77.75	88.19
First Name – Jaro-Winkler	13.11	2.21	0.09	109.25	85.07
First Name – Soundex-US	10.37	1.93	0.13	86.42	83.43
Address – Jaro-Winkler	9.82	1.72	0.11	81.83	84.29
Address – Soundex-US	7.41	1.72	0.19	61.75	79.51



• Comparison times for 1 dataset and 1 CPU:



Algorithm complexity -> 0(n²)



Breast Cancer in Auvergne: population of 70000 people

- > 4.900.000.000 N-to-N comparisons
- > 1month of CPU time

Need heuristics to speed-up the calculation:

- > Geographic restriction (zip-code)
- > Criteria filtering (birthdate, sex...)
- > Stop the process if one key-field failed (name totally different)

• What else?

- > Pre-calculation of phonetic codes
- Use grid computational power





Discussion & Conclusion





- Better integration of the Sentinelle platform inside the medical software
- Extension of the network to other pathologists/screening structures
- Extension to other types of cancers, other pathologies
- Adding medical data (Mammographies)





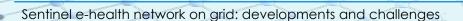
Complex statistical analysis

Use grid CPU power for cancer cluster investigation (epidemiology)

Define and monitor epidemiological alarms

Ease and enhance screening

- > Multidisciplinary collaborations
- > Real time access
- Clean duplicates in databases







 The e-sentinelle initiative started in 2009 and will enter in production phase soon

Fully grid compliant network

 National funding for the next 3 years in order to validate the RSCA





Thank you

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