

Journée des doctorants du LIG

Appréhender l'hétérogénéité à (très) large échelle

Raphaël Bleuse (raphael.bleuse@imag.fr)

sous la direction de

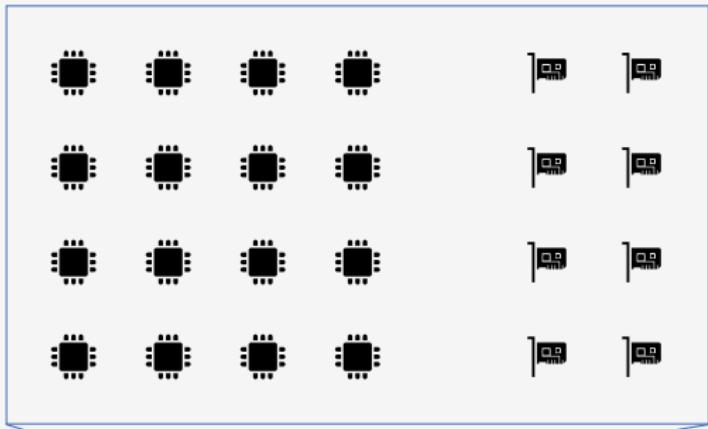
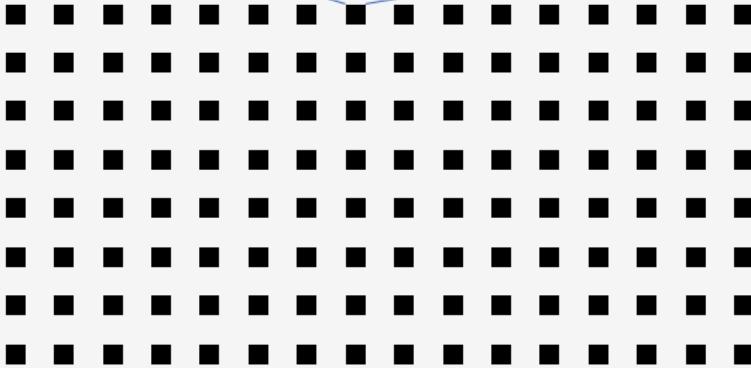
Grégory Mounié (gregory.mounie@imag.fr)

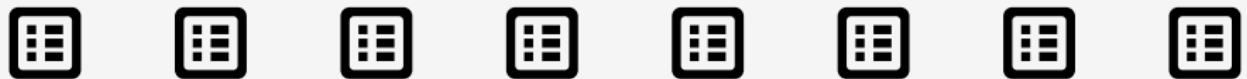
Denis Trystram (denis.trystram@imag.fr)

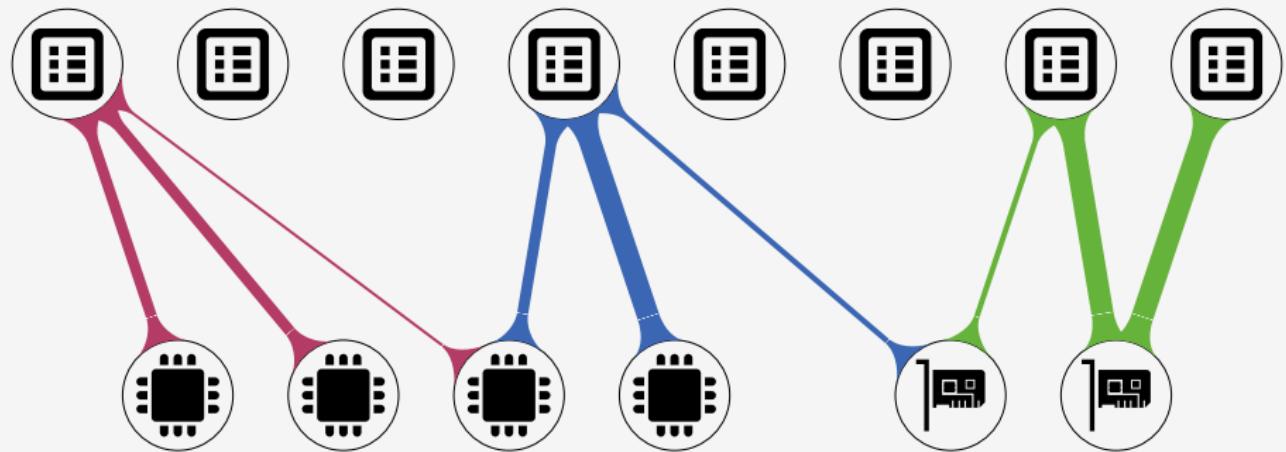


26 mars 2015

10^{18}







A photograph of a large colony of Adelie penguins gathered on a massive, light-blue iceberg. The penguins are white with black heads and dark wings. They are packed closely together, some facing forward, others looking around. The background shows more of the icy landscape and a hint of the ocean.

LET'S GO

MASSIVELY PARALLEL

An approach to reasoning about graph transformations

Jon Hal Brenas, Rachid Echahed, Martin Strecker

LIG, IRIT

26/03/2015

Example

We give a small example:

Pre: $R : \text{Researcher} \wedge L : \text{Lab}$

addR(R Member L);

while ($\exists \tau. \neg(L \text{ T_i } \tau) \wedge R : \exists \text{Pub}.(\exists T. \{\tau\})$) {

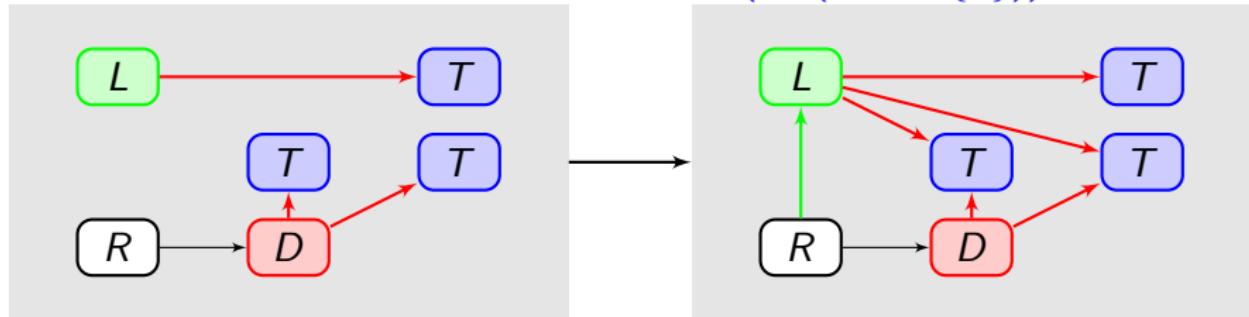
 Inv: $R : \text{Researcher} \wedge L : \text{Lab}$

select τ **with** $\neg(L \text{ T_i } \tau) \wedge R : \exists \text{Pub}(\exists T. \{\tau\})$;

addR ($L \text{ T_i } \tau$)

};

Post: $R : \text{Researcher} \wedge L : \text{Lab} \wedge R : \forall \text{Pub}.(\forall T. (\exists T \text{ T_i } \neg \{L\})) \wedge R \text{ Member } L$



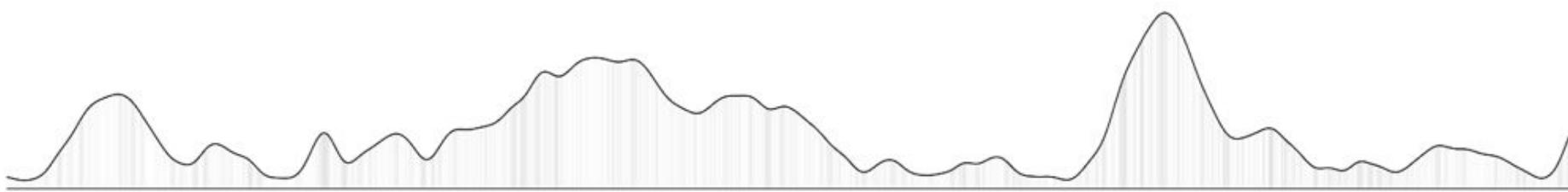
Thanks

Thank you for your attention.

Multi-scale Interaction Techniques for the Interactive Visualization of Traces of Execution

Rémy Dautriche

Supervised by: Renaud Blanch, Alexandre Termier and Miguel Santana



Why?

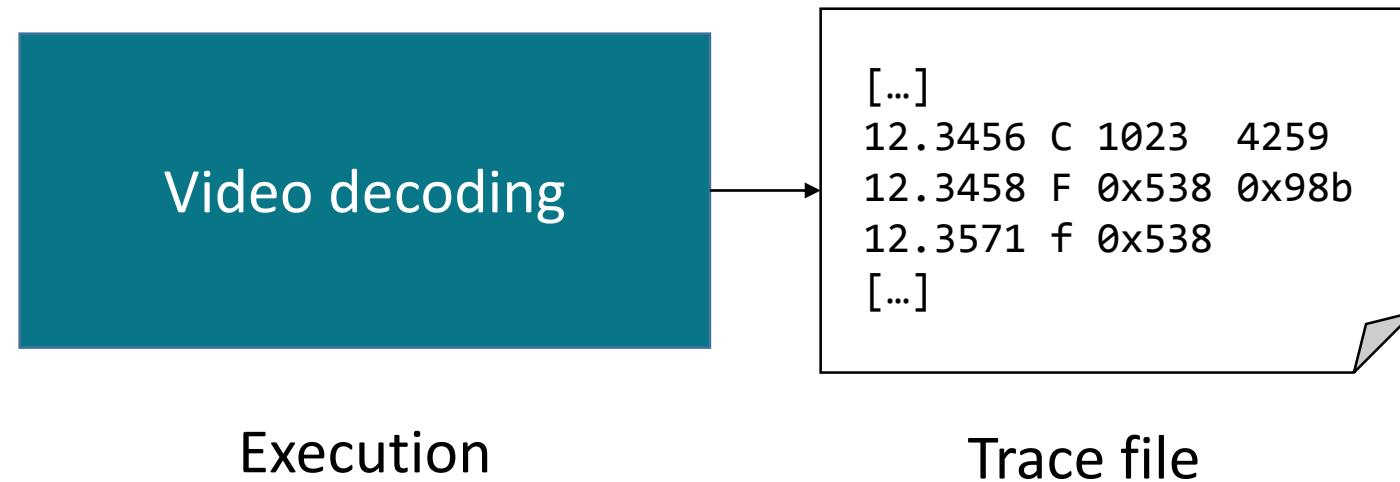
Increasing complexity
of MPSoCs

Increasing complexity
of embedded software:
heterogeneous systems,
parallel architecture

Increasing complexity
of industry standards
(H265, U4K)

Huge impact on cost and time development

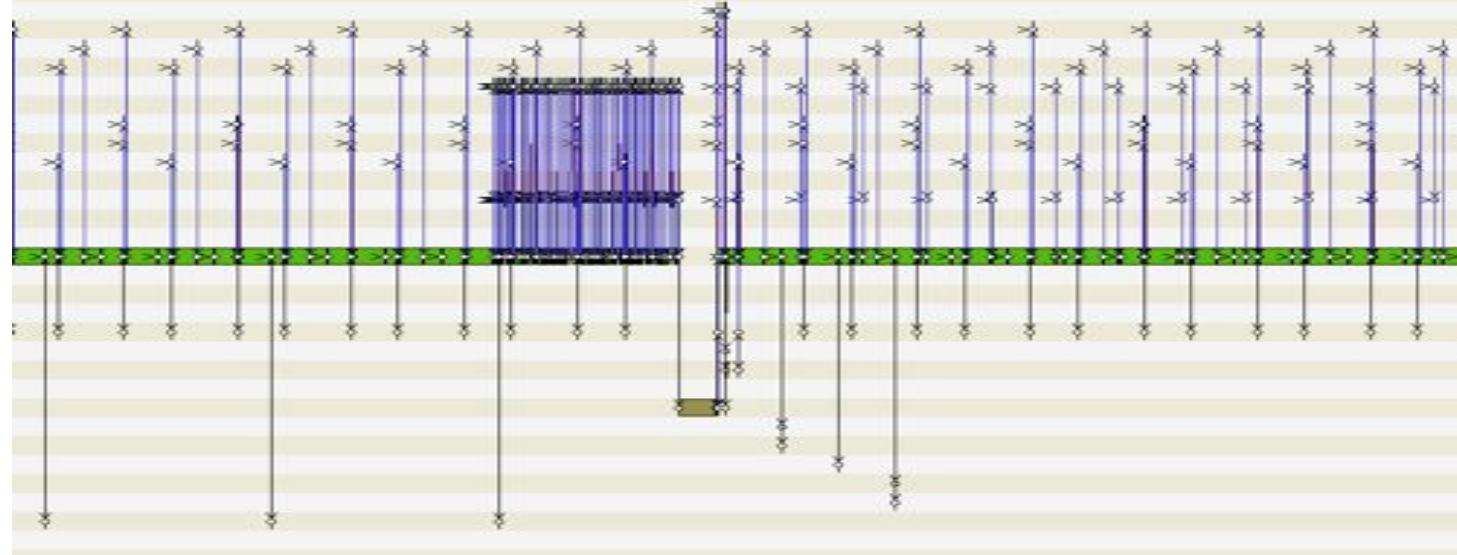
Data: Execution Traces



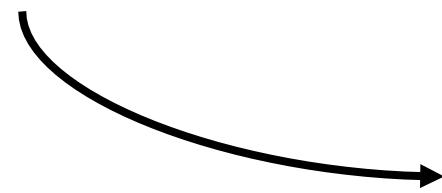
Data: Execution Traces



Challenges

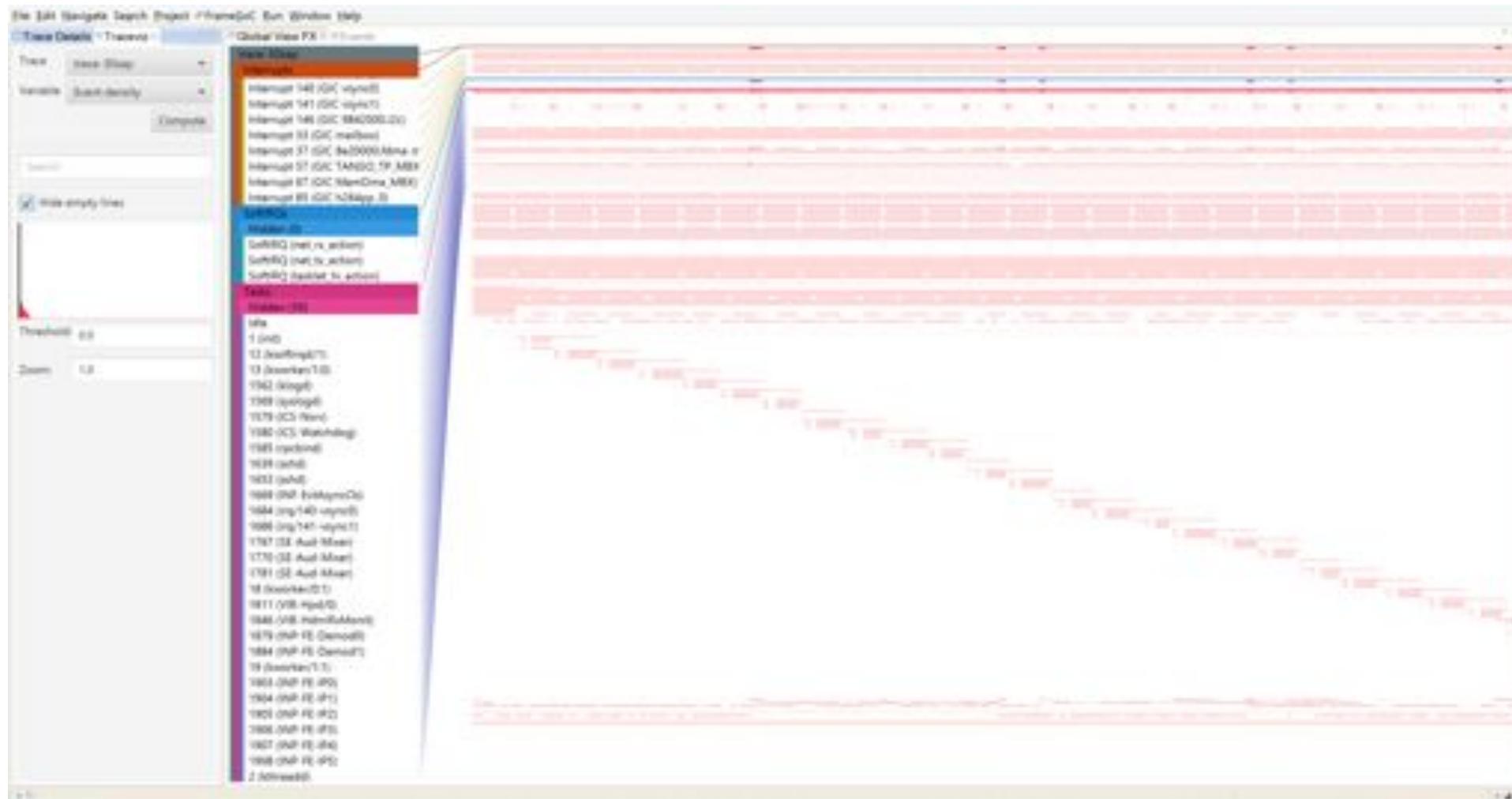


Current tools do not scale

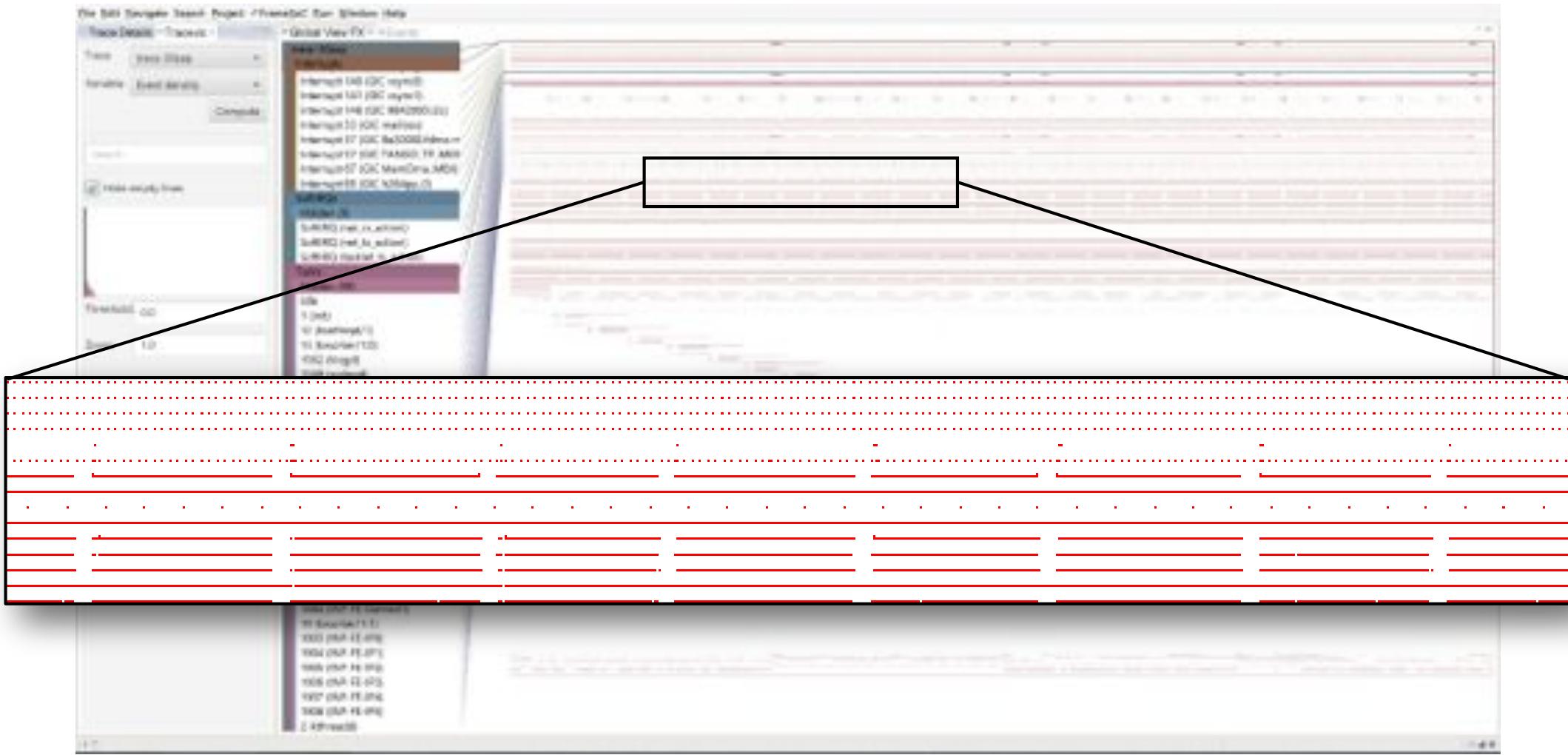


Develop new visualization techniques
to handle this amount of data

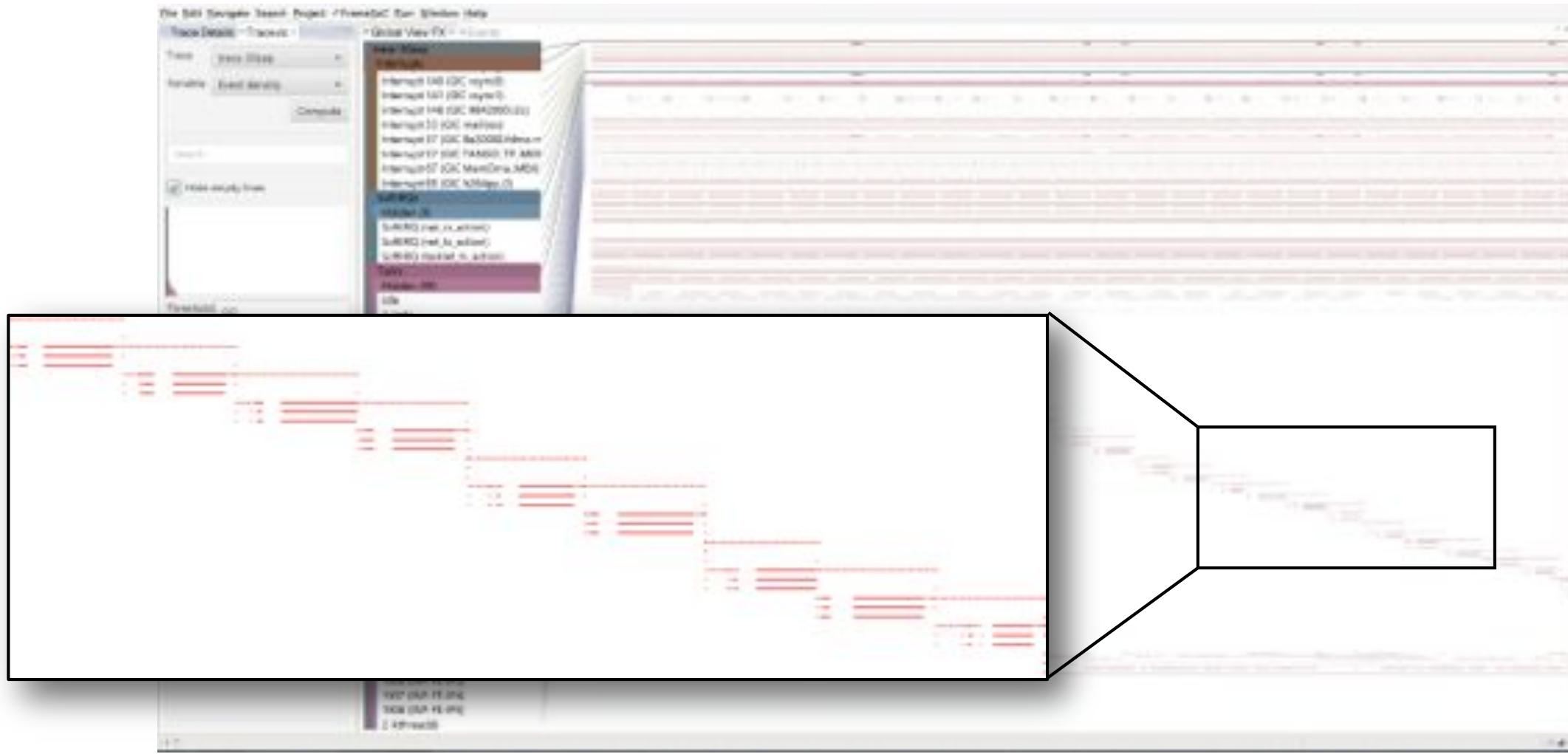
TraceViz: Overview of trace + filtering



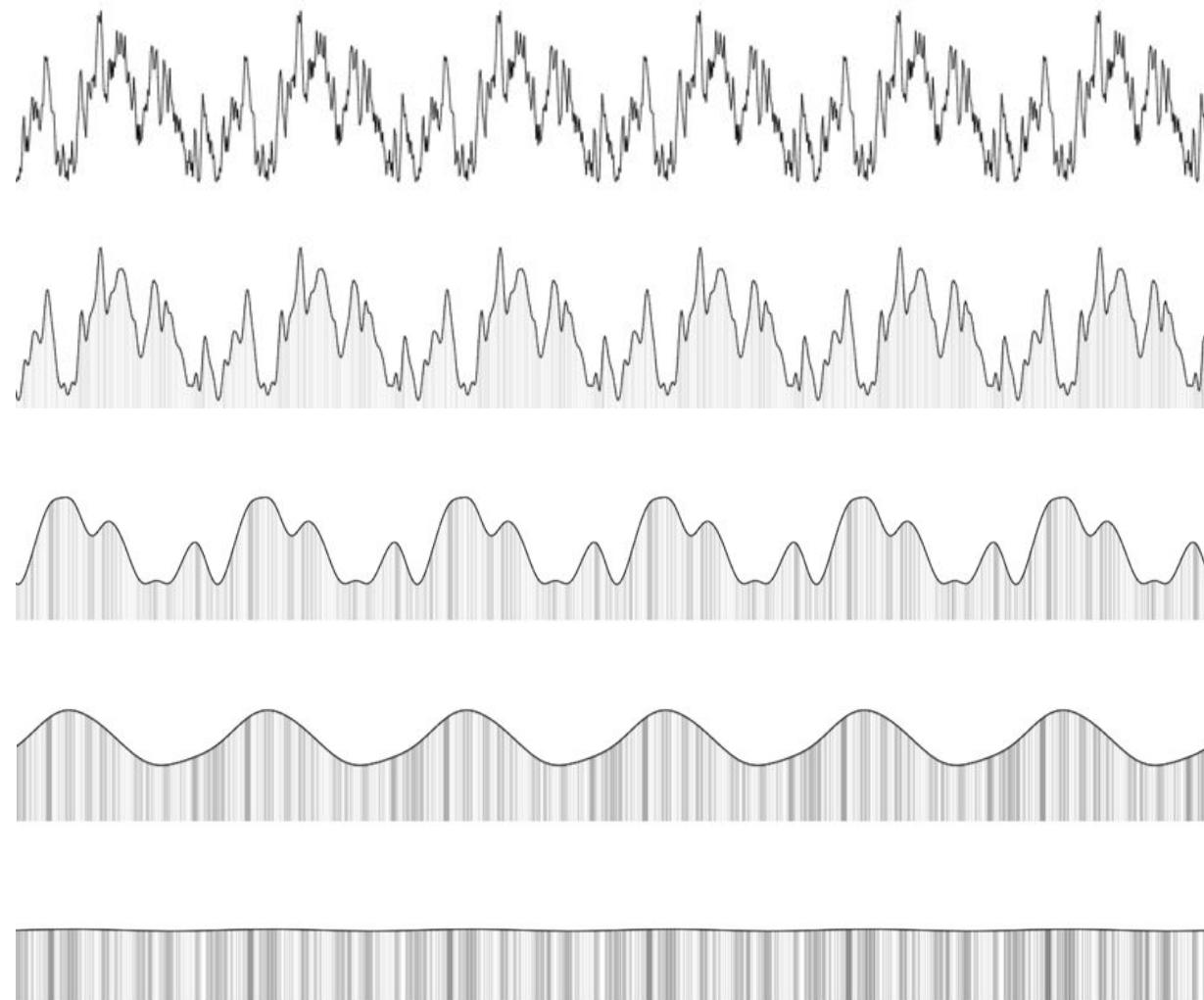
TraceViz: Detect similar and periodic behavior



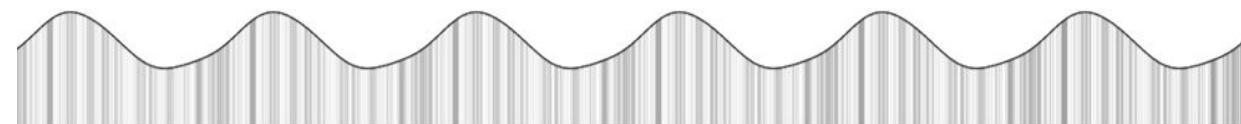
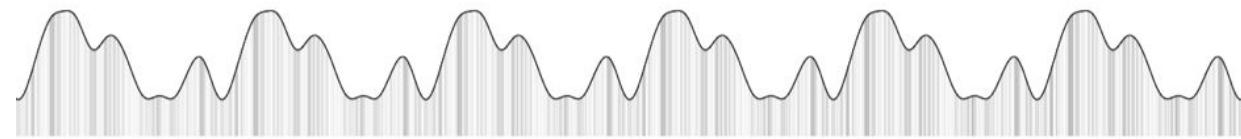
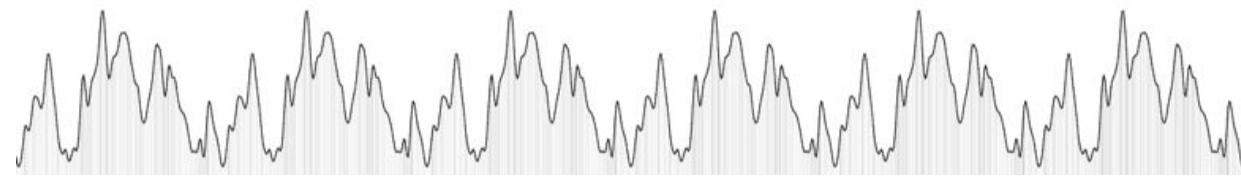
TraceViz: Identify execution patterns



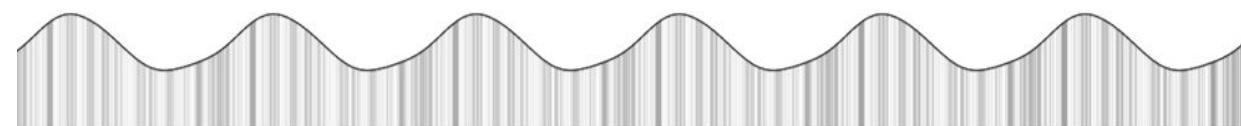
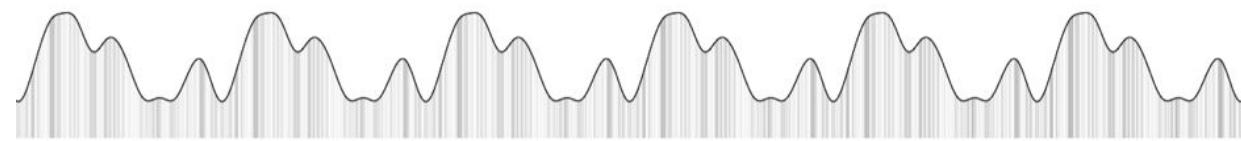
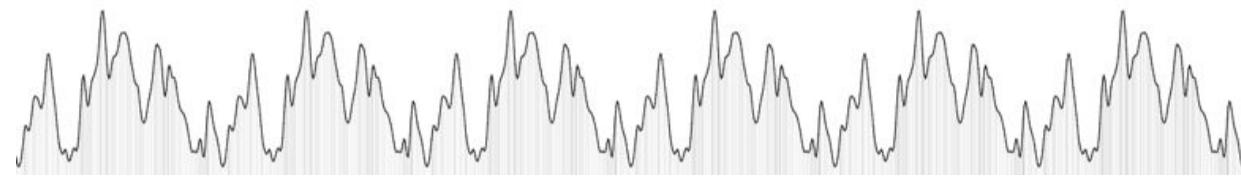
Slick Graphs: Interactive smoothing technique



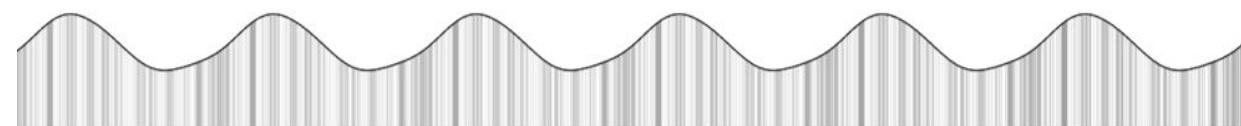
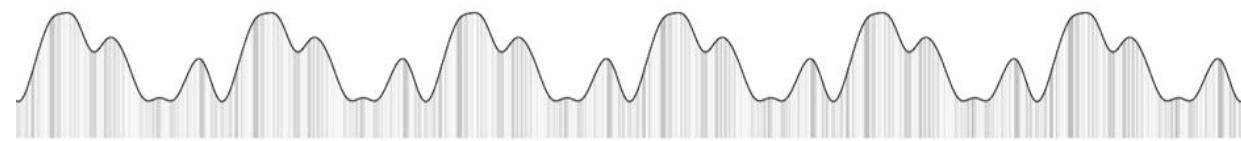
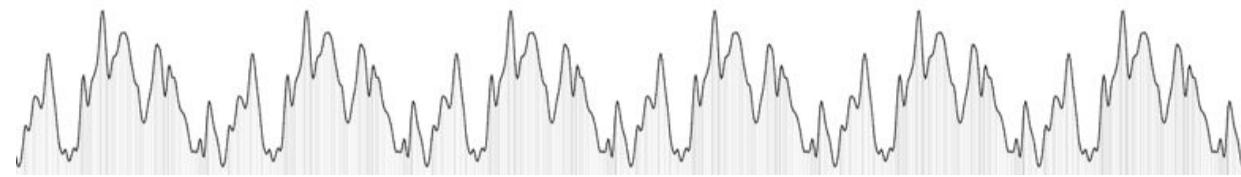
Slick Graphs: Period detection



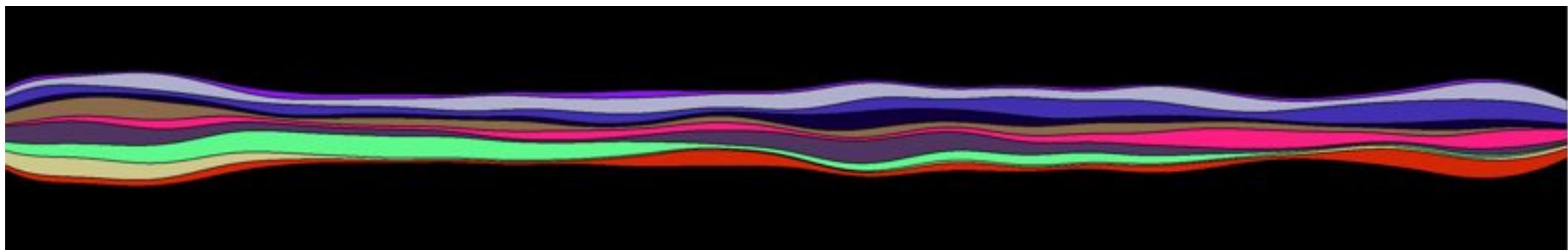
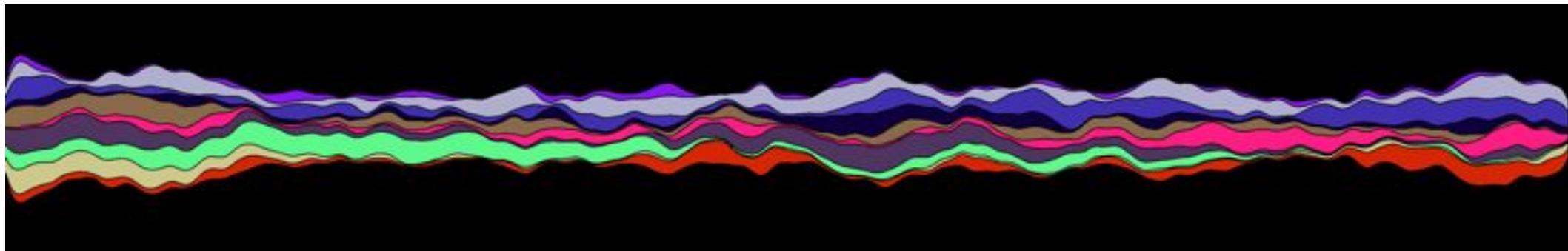
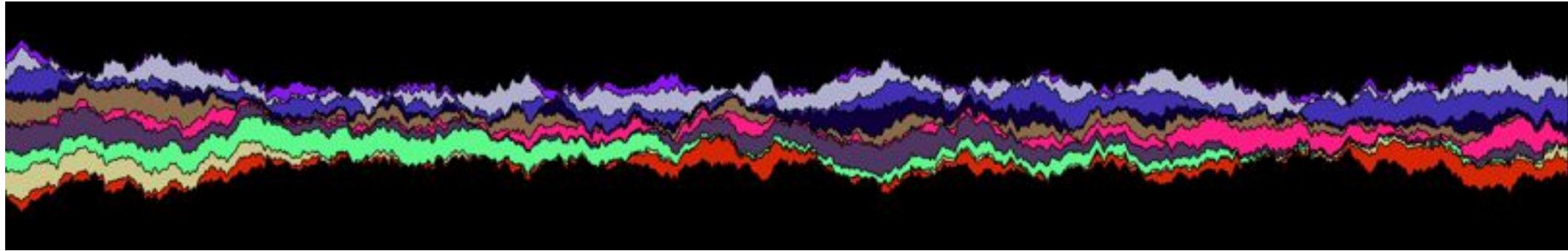
Slick Graphs: No information loss



Slick Graphs: Scalability



Slick Graphs: Integration with existing techniques



What's next?

- Integrating TraceViz and Slick Graphs
 - Industrial deployment at STMicroelectronics
- Enrich visualizations using data mining techniques for
 - Automatic error detections
 - Pattern detections

Thanks !



Journée des doctorants
Laboratoire d'Informatique de Grenoble

Construction de protocoles de soins auto-adaptatifs pour le suivi des maladies

Amira Derradji - 2^{ère} année Thèse CIFRE

Encadrants

Recherche (LIG - SIGMA)
Agnès Front
Christine Verdier



Industriel (Arcan Systems)
Vincent Bouzon



Prise en charge médicale à domicile



Le protocole de soins ...

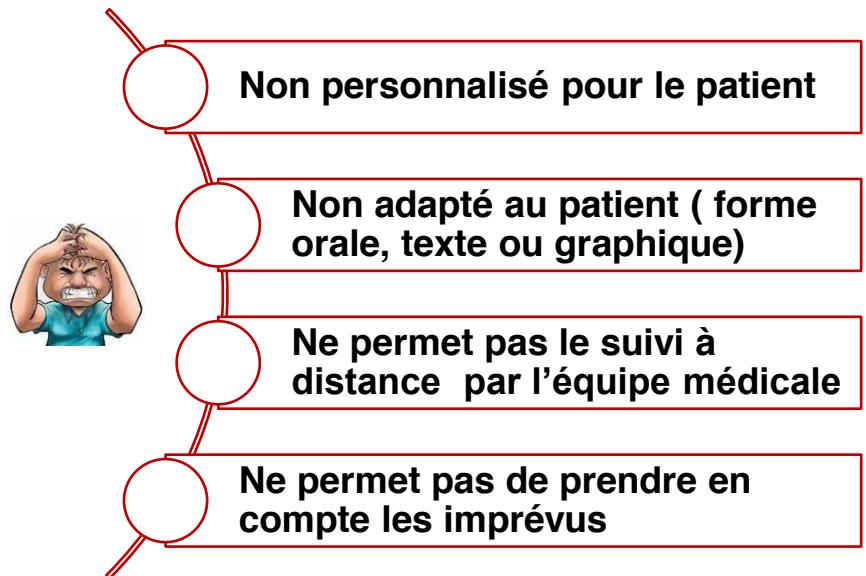
HYPOGLYCEMIE : Taux inférieur à 0,6 g/l voir 0,8g si hyperglycémies prolongées antérieures



- ↳ Variable d'un sujet à autre, toujours les mêmes pour un même patient.
- ↳ Irritabilité, vision floue, fatigue, comportement inhabituel, trouble parole, somnolence, palpitations, pâleur, sueurs, tremblements, faim, céphalées.

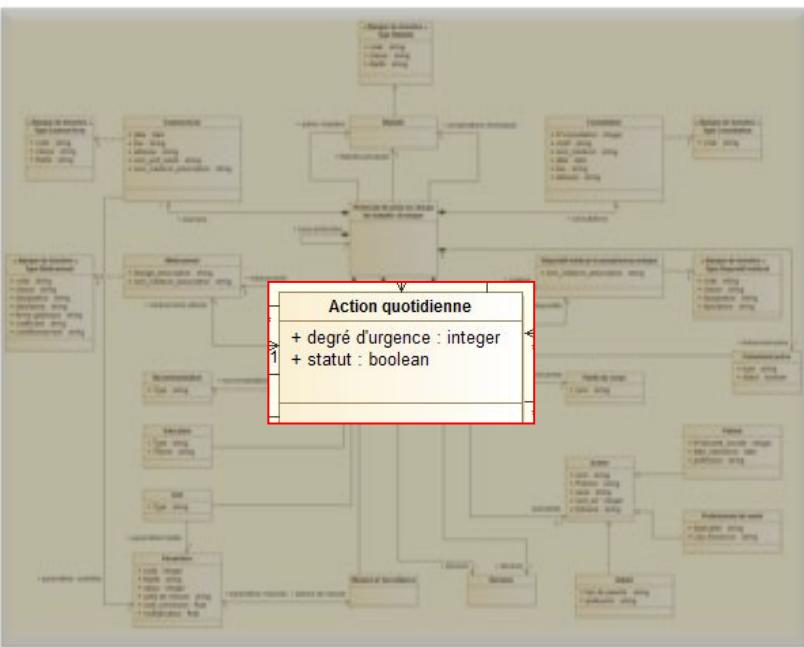
Actions

- ↳ Arrêt de l'activité et s'asseoir.
- ↳ Se resucrer avec 3 sucres soit 15 g de sucre ou équivalents jus de fruit, 1 càS de confiture ou de miel (éviter biscuits et chocolat).
- ↳ Tout de suite après faire glycémie dans les 15 mn suivant resucrage.
- ↳ Contrôler de nouveau la glycémie 30 mn plus tard si reste bas, reprendre 15 g de sucres et un sucre lent.
- ↳ Si inconscient glucagon en IM.
- ↳ Rechercher la cause de hypo.
 - quantité féculent insuffisant repas souté
 - activité physique plus importante
 - stress infection
 - médicaments
 - erreur dans traitement etc
- ↳ Prévoir ultérieurement une adaptation du traitement ou alimentation en cas d'activité plus importante.



- ✓ Workflow Management System (WFMS)
- ✓ Approches d'adaptation

Langage orienté patient



Adapter le protocole
de soins



Améliorer l'IHM patient
Reconnaissance de la parole



Merci de votre attention

Vers un outil auteur pour des EIAH destinés à l'apprentissage de méthodes de résolution

Awa Diattara, équipes : MeTAH - LIG & TWEAK - LIRIS

OBJECTIF

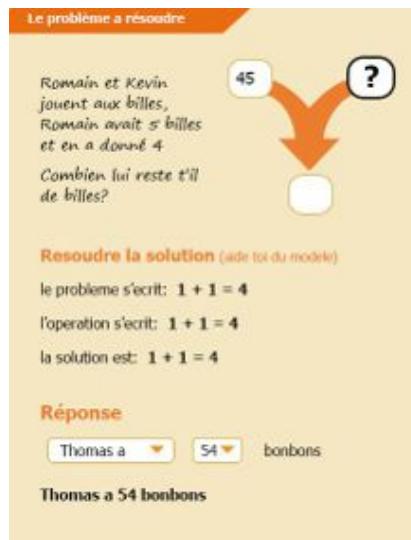
Permettre à un enseignant de n'importe quel domaine de pouvoir concevoir lui-même un EIAH de type AMBRE

POINTS CLES

- Assistance à l'utilisateur
- Généralisation à partir d'exemple

QUESTIONS DE RECHERCHE

1. Acquisition des connaissances suffisantes pour faire du raisonnement?
2. Comment concilier l'acquisition des connaissances avec la conception de l'interface de l'EIAH?



Exemple de l'EIAH AMBRE-add

Ambre
Apprentissage de Méthodes Basé sur le Raisonnement à partir de l'Expérience

The large image shows the Ambre software interface with various screens displaying problems and solutions, and a navigation menu on the left.

Contact : awa.diattara@imag.fr



DESIGN OF A BAYESIAN MACHINE

Marvin Faix

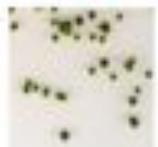
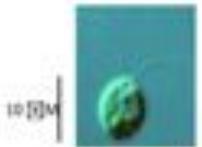
PhD Student day's 26-03-15

Thesis director : Emmanuel Mazer, LIG-INRIA

Thesis co-director : Laurent Fesquet, TIMA

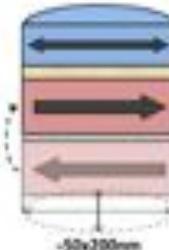
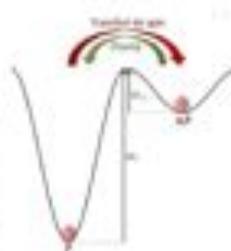
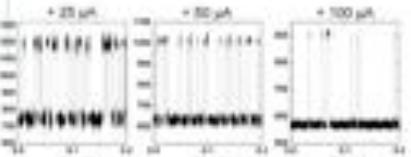
BAMBI PROJECT

« bio study chlamydomonas »



MTJ NANO COMPONENTS :

- intrinsically stochastic nano component
- tunable
- coding in time



Soft evidence

$$P(M|D) = \alpha f$$

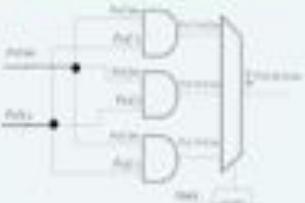
$$\hat{P}(D) \text{ with evidence input}$$

$$P(M) \prod_{i=1}^n P(D_i) + \frac{1}{Z(D)} \sum_{i=1}^n P(D_i) - \sum_{i=1}^n P(D_i) \sum_j P(M|D_j)$$

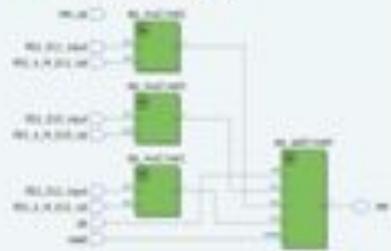
$$P = \begin{cases} P_0 & \left(\begin{array}{l} Y_0 = P_0, P_1, W \\ D_0 = (P_0, M_0, D_0) = P_0 P_1 P_2 P_3 P_4 P_5 P_6 P_7 \end{array} \right) \\ P_1 & \left(\begin{array}{l} Y_1 = P_1, P_2, W \\ D_1 = (P_1, M_1, D_1) = P_1 P_2 P_3 P_4 P_5 P_6 P_7 \end{array} \right) \\ P_2 & \left(\begin{array}{l} Y_2 = P_2, P_3, W \\ D_2 = (P_2, M_2, D_2) = P_2 P_3 P_4 P_5 P_6 P_7 \end{array} \right) \\ P_3 & \left(\begin{array}{l} Y_3 = P_3, P_4, W \\ D_3 = (P_3, M_3, D_3) = P_3 P_4 P_5 P_6 P_7 \end{array} \right) \\ P_4 & \left(\begin{array}{l} Y_4 = P_4, P_5, W \\ D_4 = (P_4, M_4, D_4) = P_4 P_5 P_6 P_7 \end{array} \right) \\ P_5 & \left(\begin{array}{l} Y_5 = P_5, P_6, W \\ D_5 = (P_5, M_5, D_5) = P_5 P_6 P_7 \end{array} \right) \\ P_6 & \left(\begin{array}{l} Y_6 = P_6, P_7, W \\ D_6 = (P_6, M_6, D_6) = P_6 P_7 \end{array} \right) \\ P_7 & \left(\begin{array}{l} Y_7 = P_7, W \\ D_7 = (P_7, M_7, D_7) = P_7 \end{array} \right) \end{cases}$$

Bayesian model description

high level circuit simulation



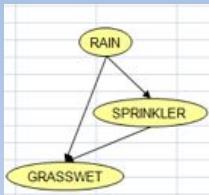
mapping on FPGA



- How to represent a probabilistic data ?
- What is a probabilistic chip ?

Algorithmic

- Bayes Inference Problem



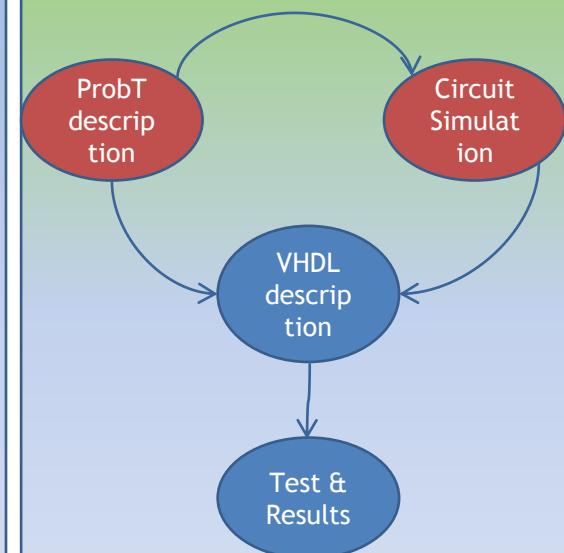
$$\sum_{s \in \{0,1\}} P(\text{RAIN} = 1) P(\text{SPRINKLER} = s | \text{RAIN} = 1) P(\text{GRASSWET} = 1 | \text{SPRINKLER} = s, \text{RAIN} = 1)$$

$$\sum_{r \in \{0,1\}} \sum_{s \in \{0,1\}} P(\text{RAIN} = r) P(\text{SPRINKLER} = s | \text{RAIN} = r) P(\text{GRASSWET} = 1 | \text{SPRINKLER} = s, \text{RAIN} = r)$$

- Soft Inference:

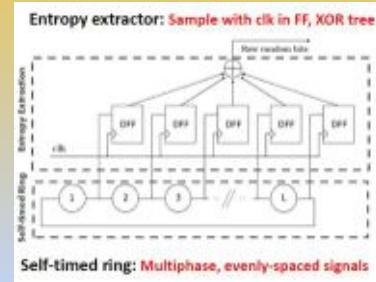
$$P(S | P(D_1), \dots, P(D_n))$$

Software/ compiler

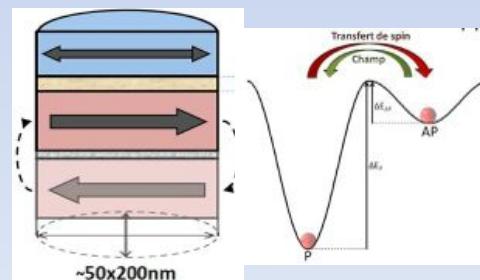


Innovative Hardware

- CMOS approach: TRNG



- Nano device approach



Efficient algorithms to mine per-item patterns in user-generated data

Martin Kirchgessner

SLIDE team

advised by Sihem Amer-Yahia & Vincent Leroy



martin.kirchgessner@imag.fr

Journées des doctorants du LIG - Mars 2015

Efficient algorithms to mine per-item patterns in **user-generated data**

ie., any data organized as “(plenty) lists of items”

Efficient algorithms to mine per-item patterns in user-generated data

ie., any data organized as “(plenty) lists of items”

- ▶ Supermarket receipts (DataLyse project)
- ▶ Playlists (the poster's examples)
- ▶ Web history
- ▶ ...

Efficient algorithms to mine per-item patterns in user-generated data

1. eggs, sugar, flour, butter, nuts
2. flour, sugar, eggs, milk
3. paper, flour, eggs, sugar
4. flour, eggs, sugar
5. butter, bread, cheese
6. flour, sugar, eggs, bread, carrots
7. salad, bread, sugar, flour, eggs

...

Efficient algorithms to mine per-item **patterns** in user-generated data

1. eggs, sugar, flour, butter, nuts
 2. flour, sugar, eggs, milk
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 7. salad, bread, sugar, flour, eggs
- ...

Efficient algorithms to mine per-item **patterns** in user-generated data

1. eggs, sugar, flour, butter, nuts
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 4. flour, eggs, sugar
 5. butter, bread, cheese
 6. flour, sugar, eggs, bread, carrots
 7. salad, bread, sugar, flour, eggs
- ...

But we have a billion receipts !

Efficient algorithms to mine **per-item patterns** in user-generated data

We propose a way to:

- ▶ organize the result set
- ▶ reduce the result set

Efficient algorithms to mine per-item patterns in user-generated data

ie. “parallelized”

- ▶ on multi-core
- ▶ on Hadoop clusters

Find me in front of:

Efficient algorithms to mine per-item patterns in user-generated data

Martin Kirchgesner, advised by Sihem Amer-Yahia & Vincent Leroy
Preliminary works with Alexandre Termier & Marie-Christine Rousset



We adapt pattern mining to user-generated data, at Web scale.

- Many systems record per-user (or per-action) lists of items: retail tickets, playlists, browser logs...
- Patterns (here, sets of artists) allow an analyst to:
 - summarize the data by highlighting co-occurrences
 - link each pattern back to the concerned users, for further analysis

Frequent Itemset Mining can find that {
294,720 people listen to Radiohead
61,487 to {The Beatles, The Rolling Stones}
73,666 to {Coldplay, Muse} ... and millions of such sets }

Lesser-known artists are also part of interesting patterns

CHALLENGES

FIM can't find {
The 19,882 people who all listen to {Franz Ferdinand, Radiohead, Interpol} ?
1,995 hipsters listening to {Vitalic, Hot Chip} ?
1,994 (metal) fans of {Children of Bodom, In Flames, Arch Enemy, Eternal Tears of Sorrow} ?
200 (likely French) listeners of {Tyo, Les Wriggles, Fabulous Trobadors} ?
10 eclectic users who know {Sido, B-Tight, Tony D, G-Hot, Fler, Sido, Alpa Gun} ?

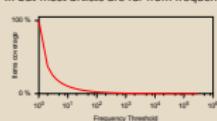
EXAMPLE : LastFM

Listing the 50 favorite artists...

... of 1.2M users

>1M DISTINCT ARTISTS

... but most artists are far from frequent



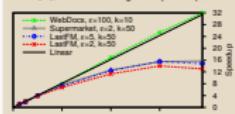
BIG DATA SYSTEMS

Modern systems are heavily parallel. Our algorithms leverage such architectures, to analyze datasets at all scales.

On multi-core

TopPI's solution space is a tree.

By assigning a branch to each available core [2], we achieve good speedups :



Examples presented here are extracted

We propose **TopPI**

The first algorithm and implementation to find Per-item patterns

READABILITY ASSUMPTIONS

- It's OK to study 2 or 3 dozens of patterns
- The analyst usually jumps in the results via a key item
"What's related to The Beatles?"

A NEW PROBLEM STATEMENT

Given transactions over a set of items I and an integer k , return, $\forall i \in I$, the k most frequent closed itemsets containing i .

We get, for each artist a listened to at least 2 times, k sets of artists including a

MINING, ALL AT ONCE

Mining {a,b,c} is much more profitable if it's a top-pattern for a, b and c

SLIDE

Pattern Discovery and Exploitation"

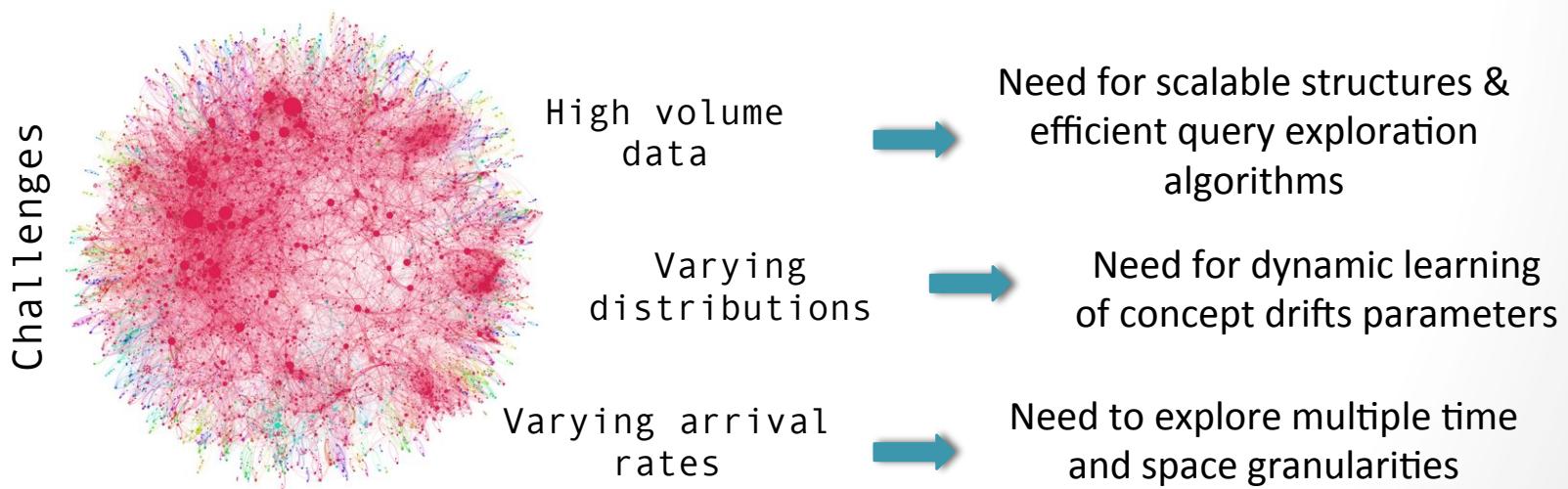
Spatio-Temporal Concept Drift Exploration

Sofia Kleisarchaki

Advisors: Sihem Amer-Yahia, Ahlame Douzal-Chouakria,
Vassilis Chistophides

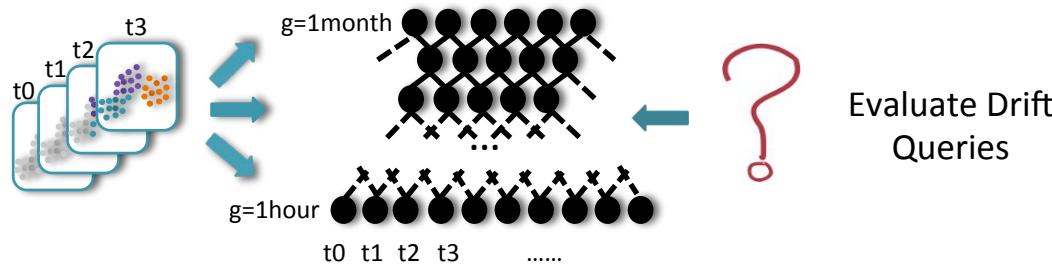
Context

- Imagine the ‘Universe of Data’
 - Consists of data forming *concepts evolving in spacetime*
- Analysts as Observers & Explorers
 - Observe the universe to *reveal concept drifts*
 - Explore the observable space to *query time and space dynamics* of concept drifts

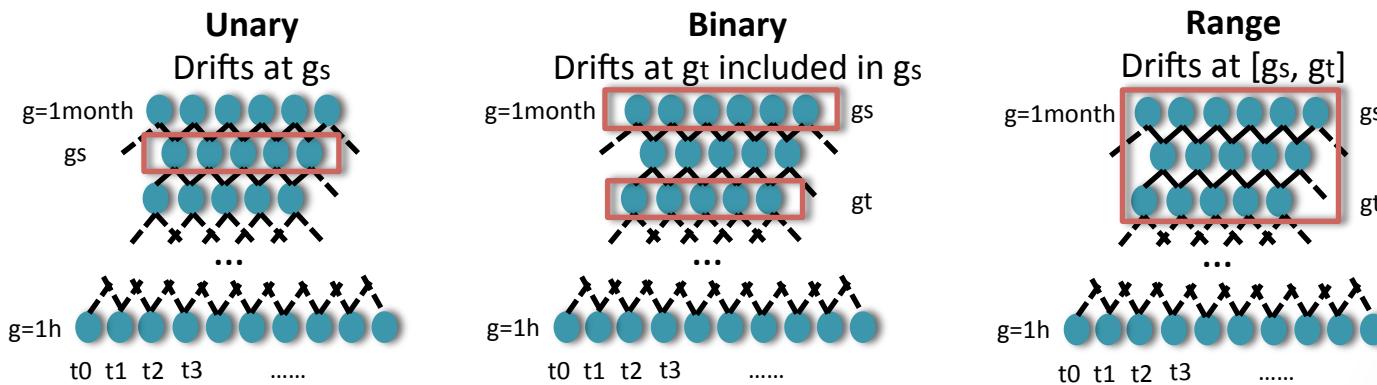


Approach & Contributions

- Provide a scalable drift index maintaining concepts in multiple time granularities

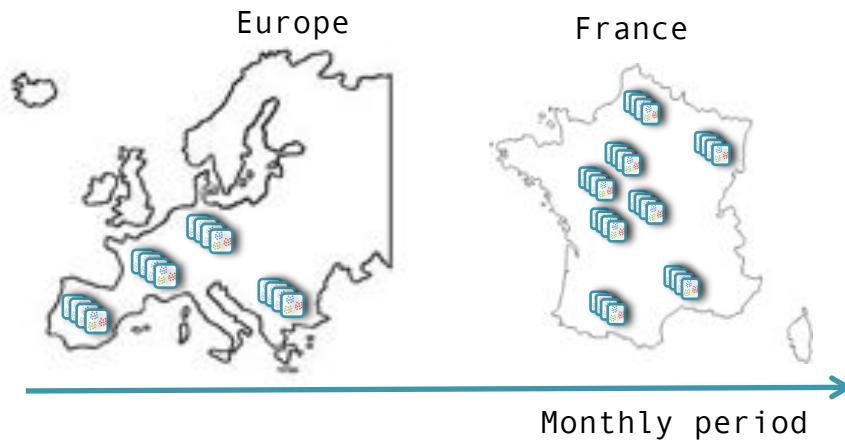


- Formalize flexible drift queries and evaluate efficiently



Future Work

- Explore the spatial dimension of geo-annotated data
 - Drifts occur at several geographic levels (e.g., Europe, France) and time granularities (e.g., day, month)



- Provide a flexible approach for querying spatio-temporal concept drifts



AMIA

Analyse de données, Modélisation et Apprentissage automatique

Développement d'un robot attentionné pour l'aide aux personnes en situation de fragilité



Quentin Labourey

Directeurs :
Olivier Aycard – LIG
Denis Pellerin – GIPSA-lab



Pourquoi ?

- ✓ Surveillance de personnes fragiles par des capteurs dans l'infrastructure (caméras, micros...) est mal acceptée
 - Capteurs trop intrusifs, intégration coûteuse et complexe
- ✓ Développer un robot compagnon attentif pour la surveillance d'un ensemble de personnes en situation de fragilité
 - Robot bien visible équipé d'une tête
- ✓ Robot compagnon attentif capable de
 - Percevoir son environnement
 - Analyser des situations complexes
 - Focaliser son attention
 - Naviguer dans un environnement dynamique



AMA

Analyse de données, Modélisation et Apprentissage automatique

Comment ? Meet Qbo !

Capteurs : Caméras & Micros
Actionneurs : Tête et base

Qbo = Système de perception active



Mes travaux portent sur :

- La perception (Détection de locuteurs, classification de sons, détection de visages...)
- La navigation « intelligente » (Comment construire son itinéraire pour la surveillance ?)



LIG

Analyse de données, Modélisation et Apprentissage automatique

Rendez-vous au poster !



Platform of Contextualization for the Personal Cloud

LIG PhD day – 26 March 2015

Anh Dung LE - SIGMA

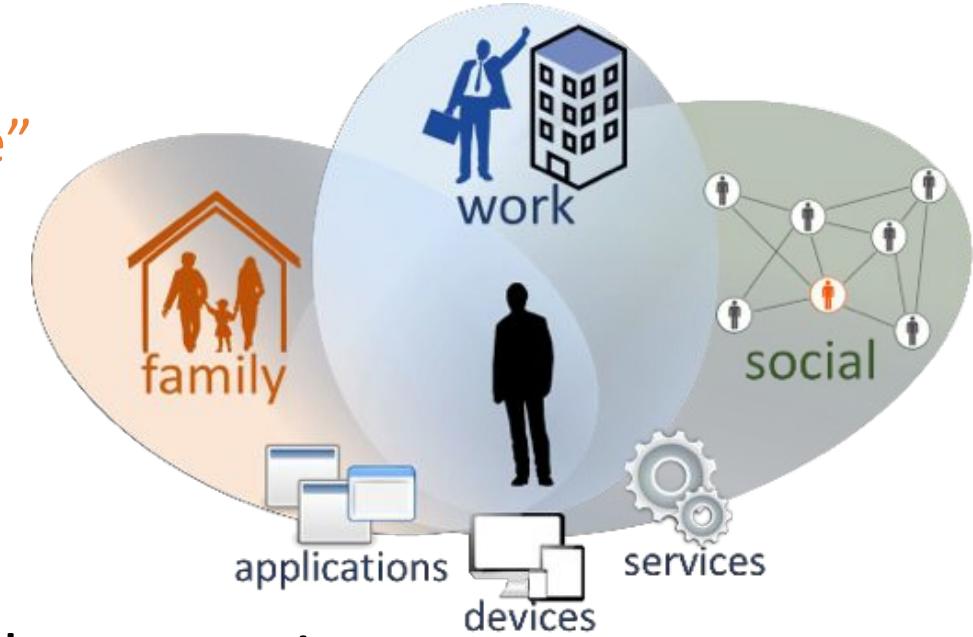


Personal Cloud and Contextualization

● Personal Cloud

“A place where I interact in safe with my digital life”

- Facilitate the management, the execution and the interaction of the user with his digital universe
- Personalized user experience

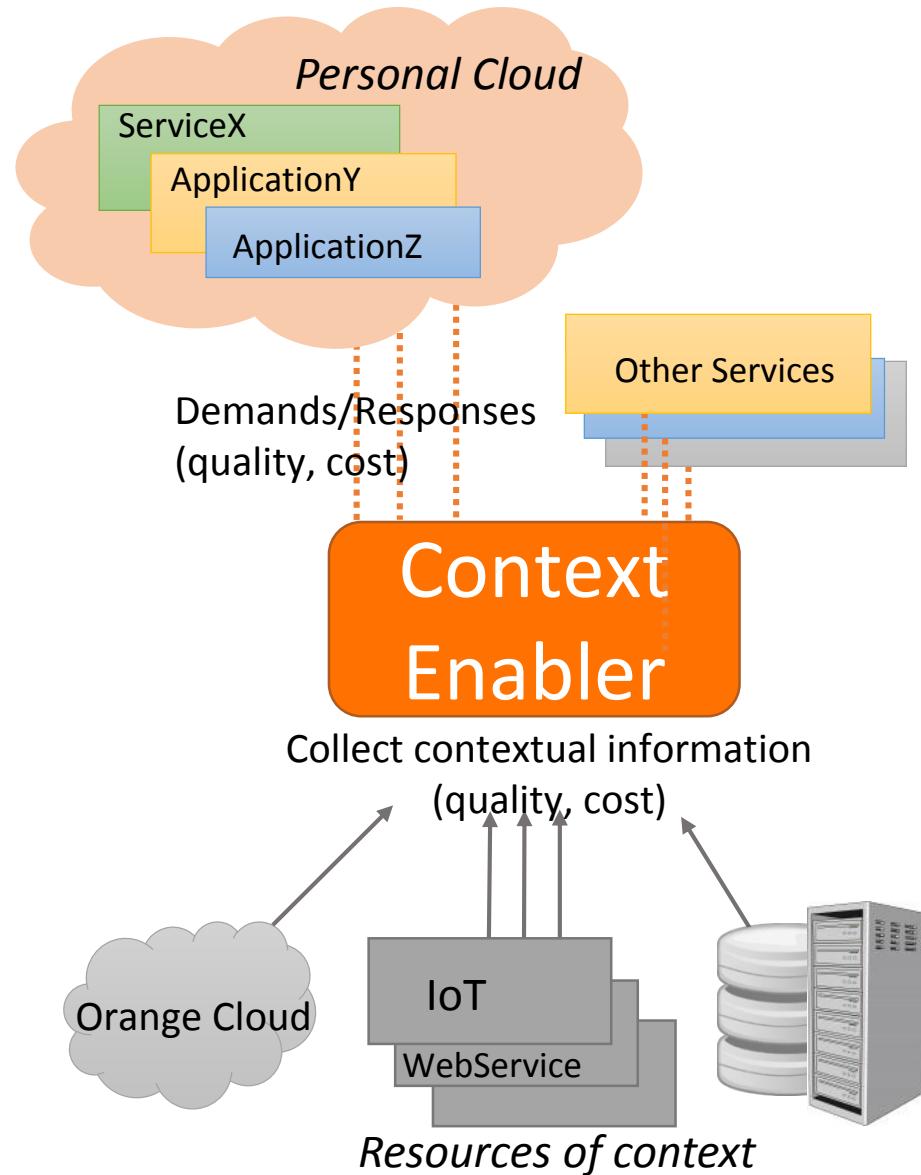


● How ? → *Context-awareness* to personalize the execution of application/service with:

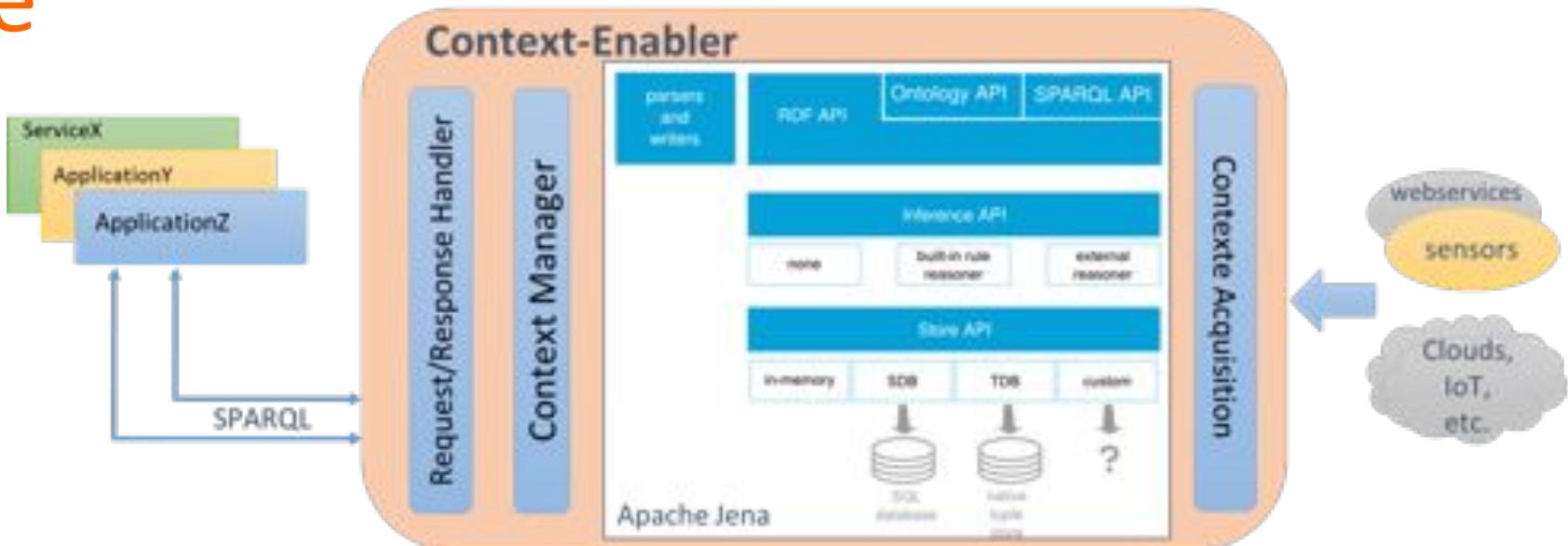
- Functional context: location, temperature, time, etc.
- Execution context: resources, data, user's profile, etc.

Feeding the Personal Cloud with a quality and cost-aware Context-Enabler

- Applications demand context with quality and cost constraints
- Semantic description, quality and cost information to classify and to make choices of context resources
- Continuous responses by adapting dynamically its context provisioning chain



Prototype



Context representation: Ontology OWL/RDF
RDF framework: Apache Jena
Rules of reasoning: SWRL

Query language: SPARQL, C-SPARQL
Communication/Messaging : RabbitMQ

Research perspectives

- Definition and implementation of the Quality and Cost of context aspects
- Representation and manipulation of semantically equivalent resources
- Language to query and to handle complex-context information
- Techniques to allow the dynamic adaptation of the Context-Enabler

Thank you 😊

Scheduling with contiguity and locality

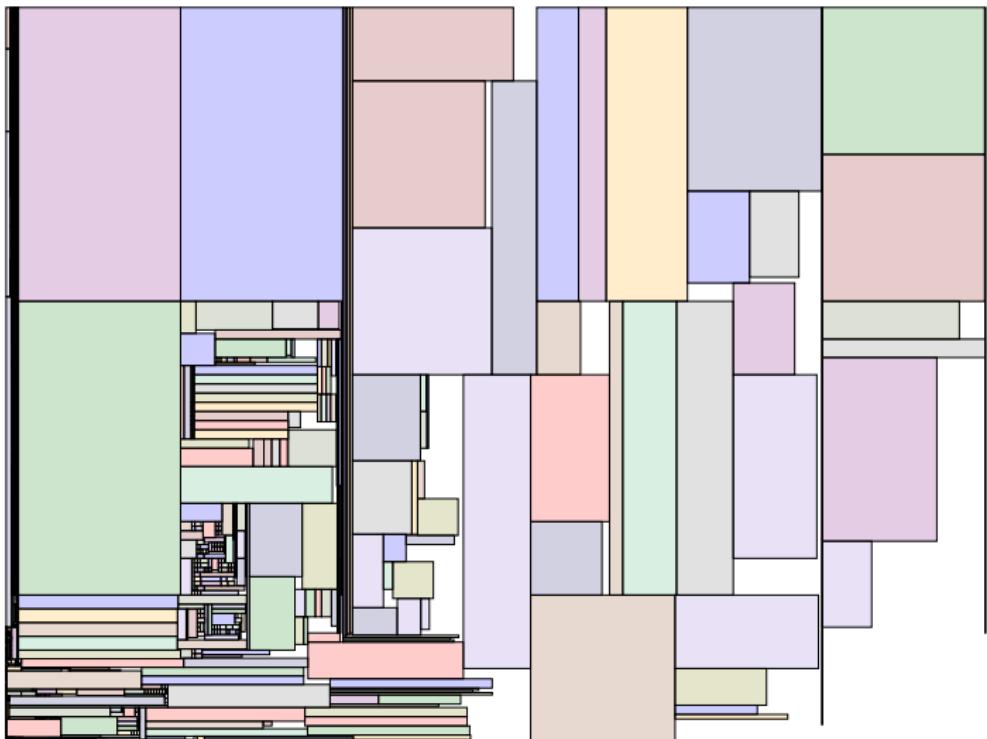
Fernando Mendonca
supervised by
Denis Trystram Frédéric Wagner



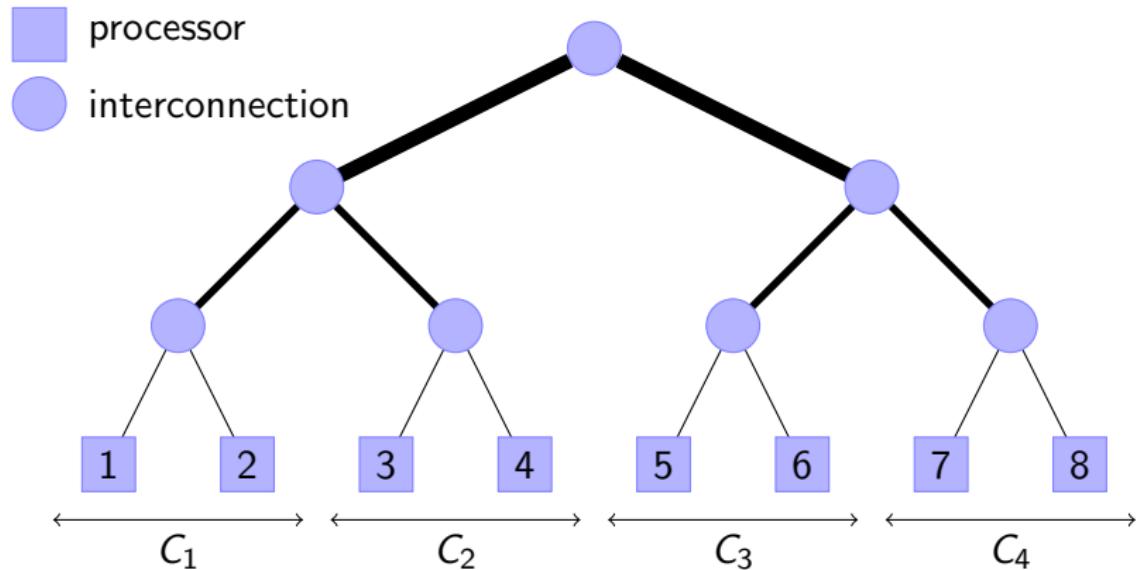
M O A I S

March 25, 2015

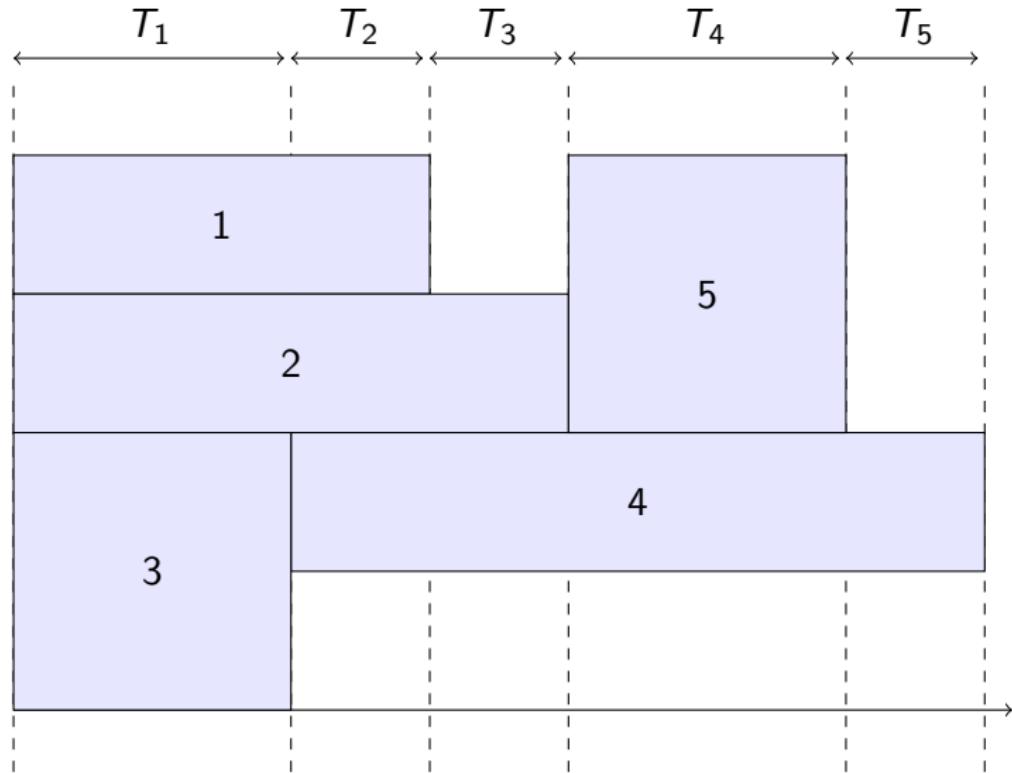
Batch schedulers

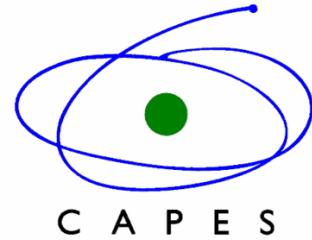


Platform



Algorithm





Privacy-aware Personal Information Discovery

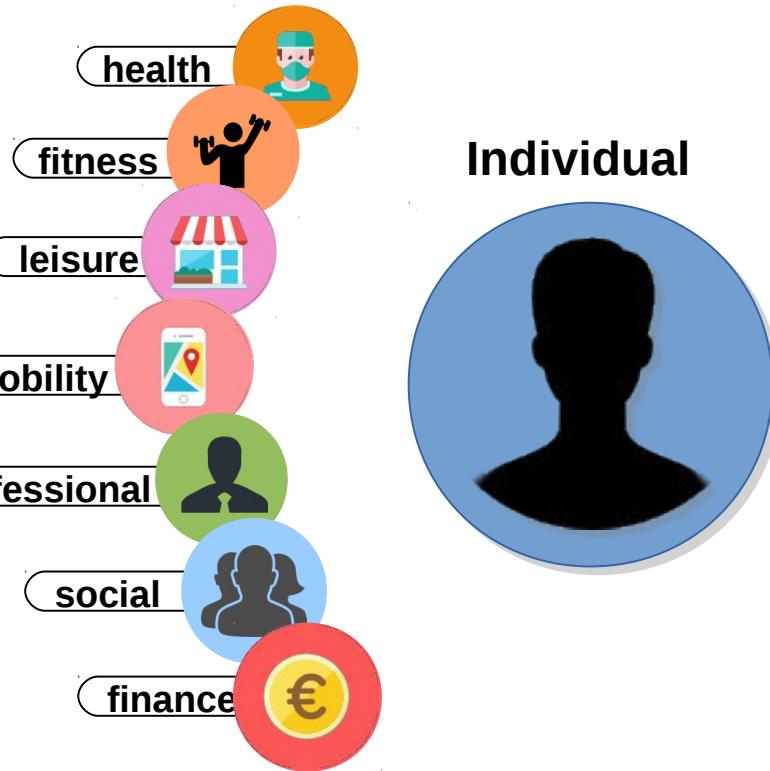
Thiago Moreira da Costa

Supervisor: Hervé Martin

How to control personal privacy In the Big Data Era?

Big Data

Privacy Threats



Personal Privacy



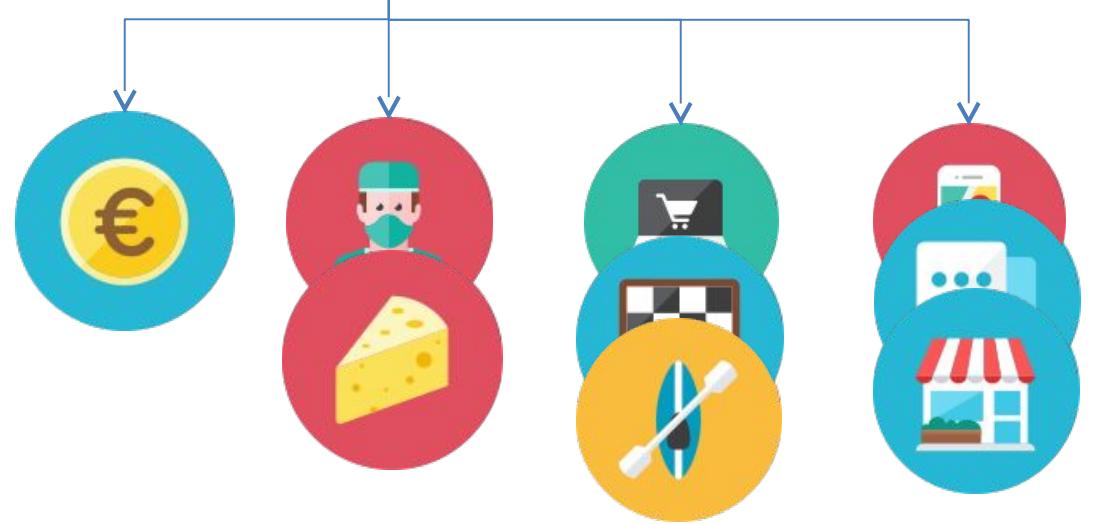
- What to share?**
- To whom?**
- When to share?**



Life Aspects



**Privacy
policy**



Finance

Health

Leisure

Social

How to control personal privacy In the Big Data Era?

Privacy Preservation
Privacy Preservation Platform

Personal Privacy
Life Aspects + Privacy Policy

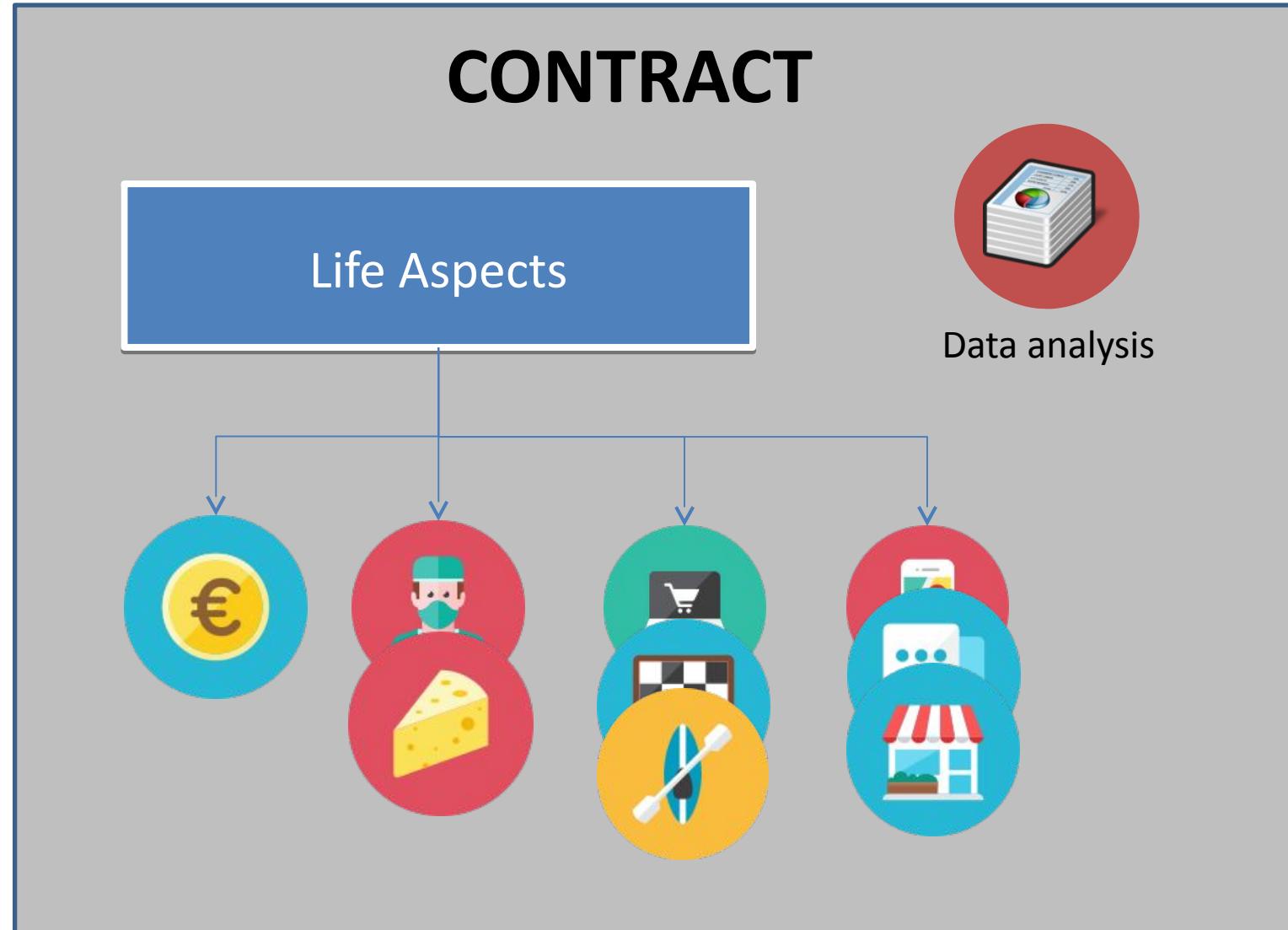
Big Data
Life Logging + Data Mining

**Personal Information
Discovery
Contract**

Personal Privacy



Data Producer



Data Consumer

Thank You!



Human-Robot Motion: an Attention-Based Navigation Approach

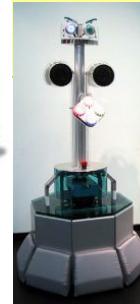
Rémi Paulin

LIG – INRIA, équipe PRIMA

Supervisors : Thierry Fraichard and Patrick Reignier

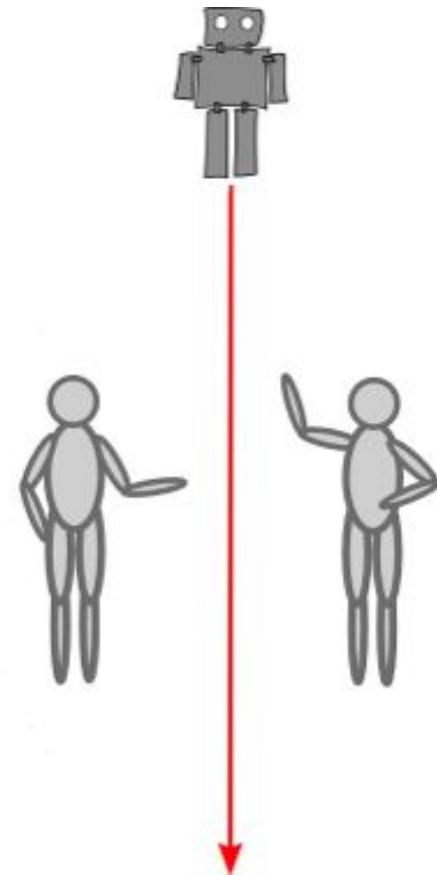
Journée des doctorants LIG 2015
26 Mars 2015

1997



2005

Robots see humans as mobile objects.

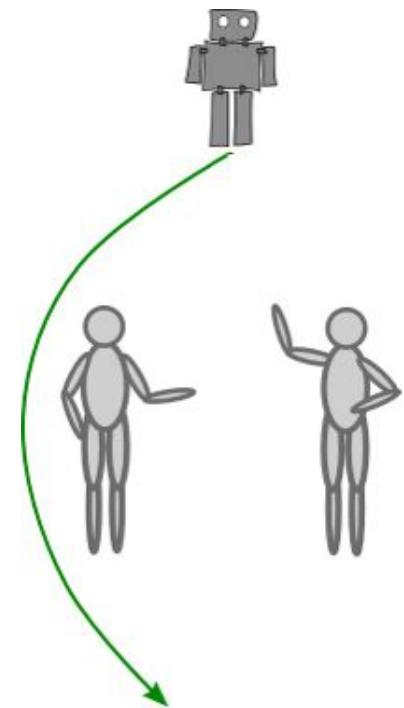
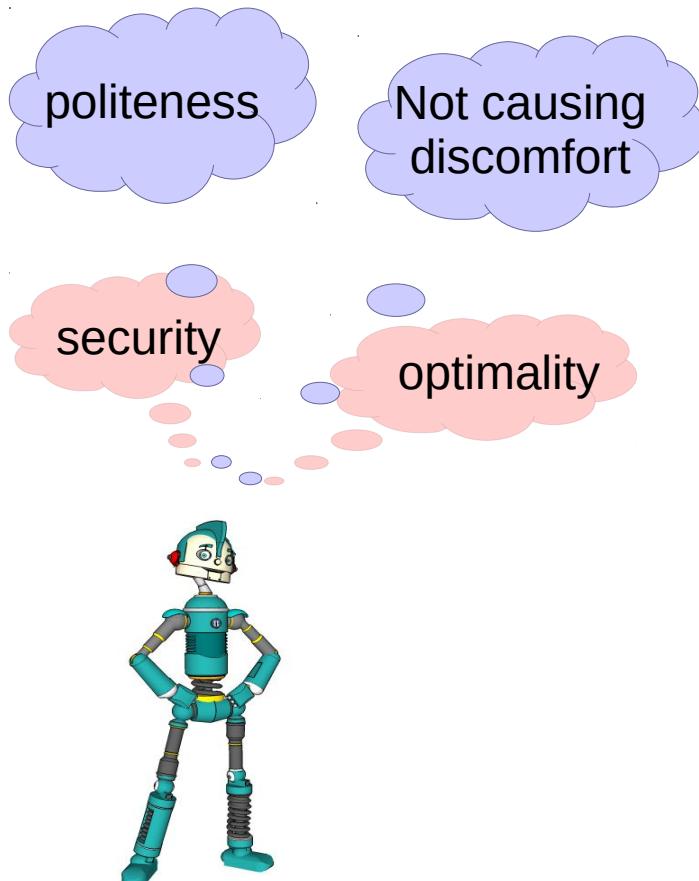


motion safe and optimal
but inappropriate

2005

now

Robots see humans as social entities.

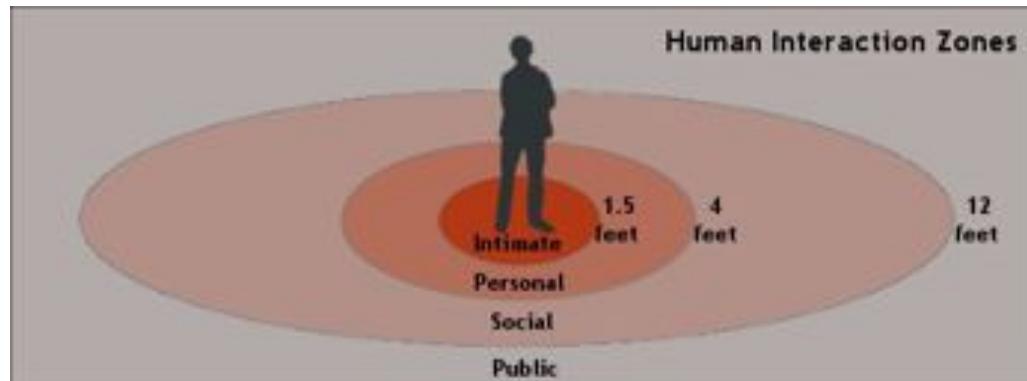


More appropriate motion

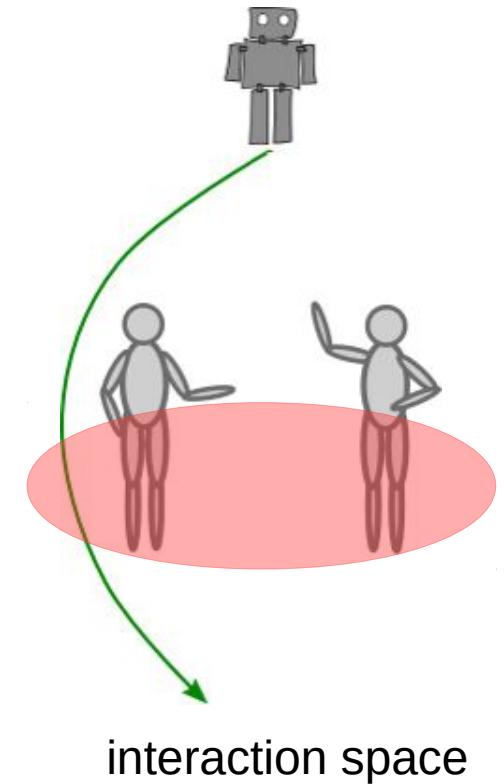
State-of-the-art: social spaces

[Lindner & Eschenbach 11]

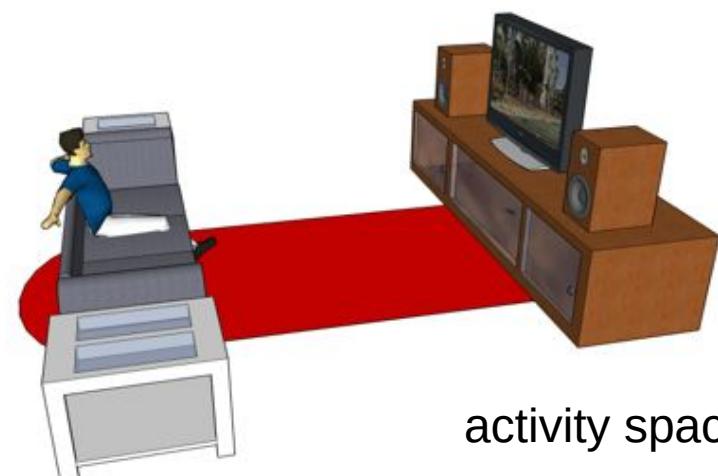
Social space: regions of space where the presence of others causes discomfort.



personal space

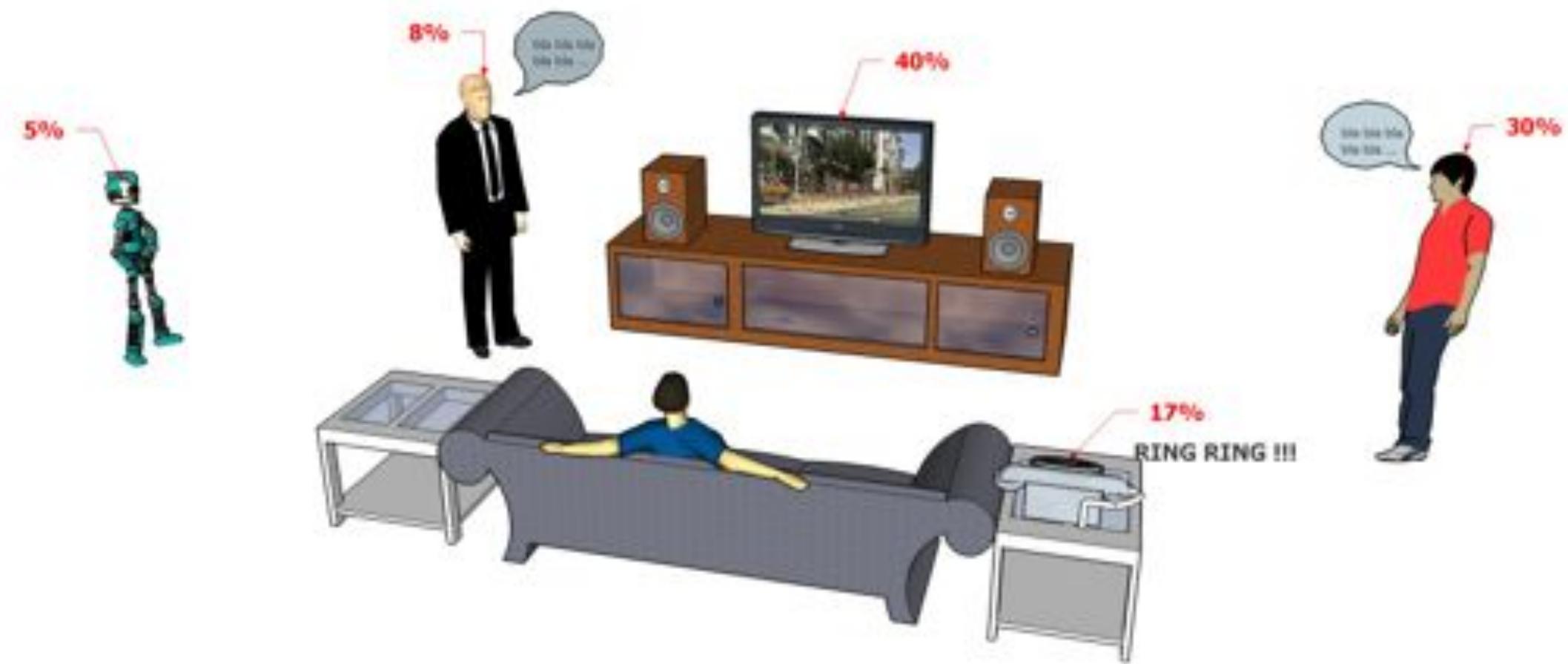


interaction space



activity space

Attention: a new tool for navigation



Attention Model [Maisonasse et al 06]

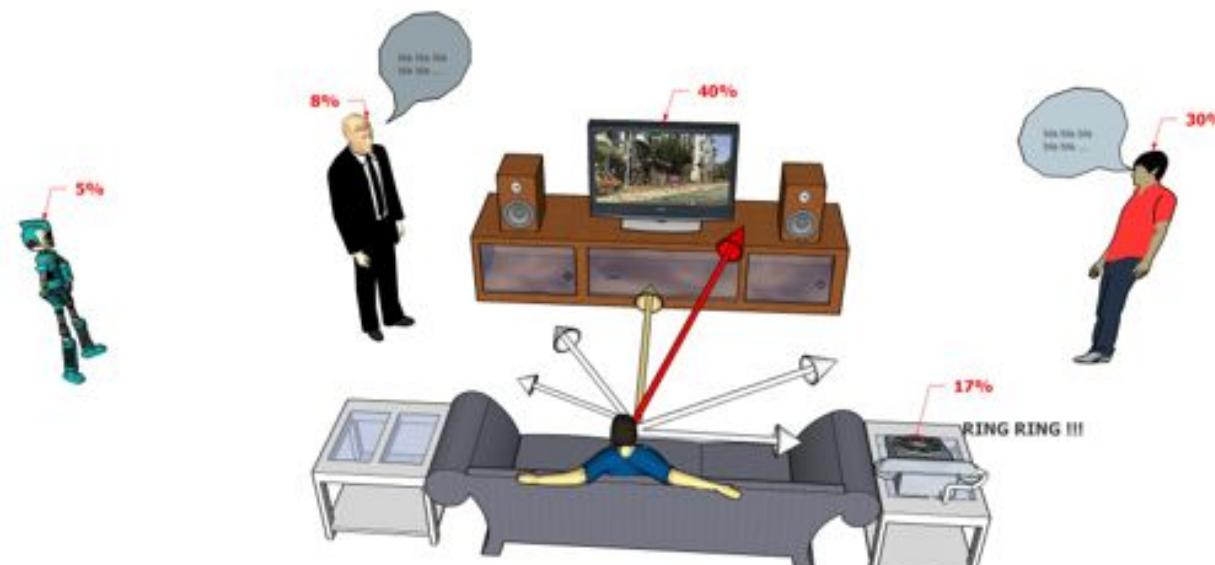
$$\text{Attention} = f(\text{Intention}, \text{Distractions})$$



Individual's activity

Salience of entities

general direction of
attention focus.

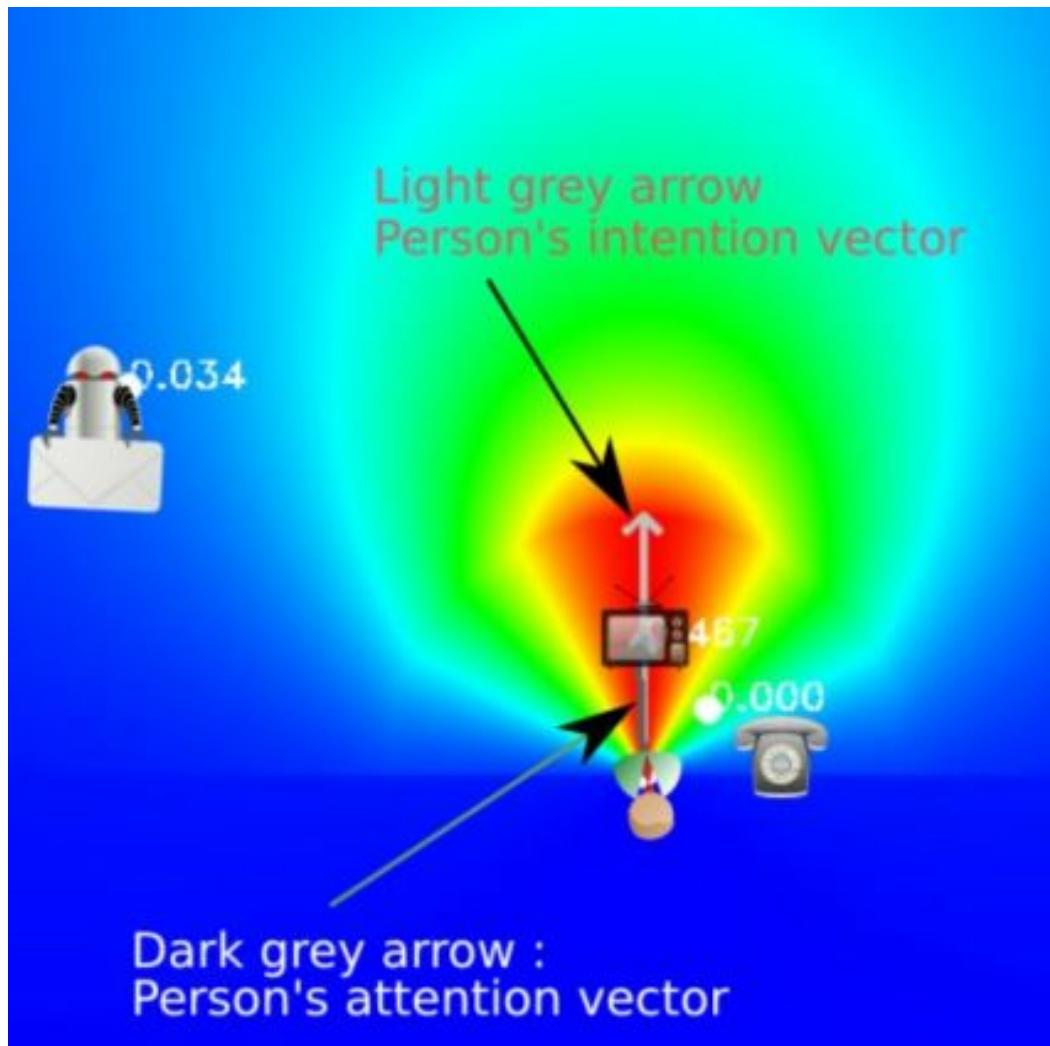


Scenario

		8%	30%	5%	40%	17%
	45%		2%	35%	8%	10%
	60%	3%		20%	2%	15%

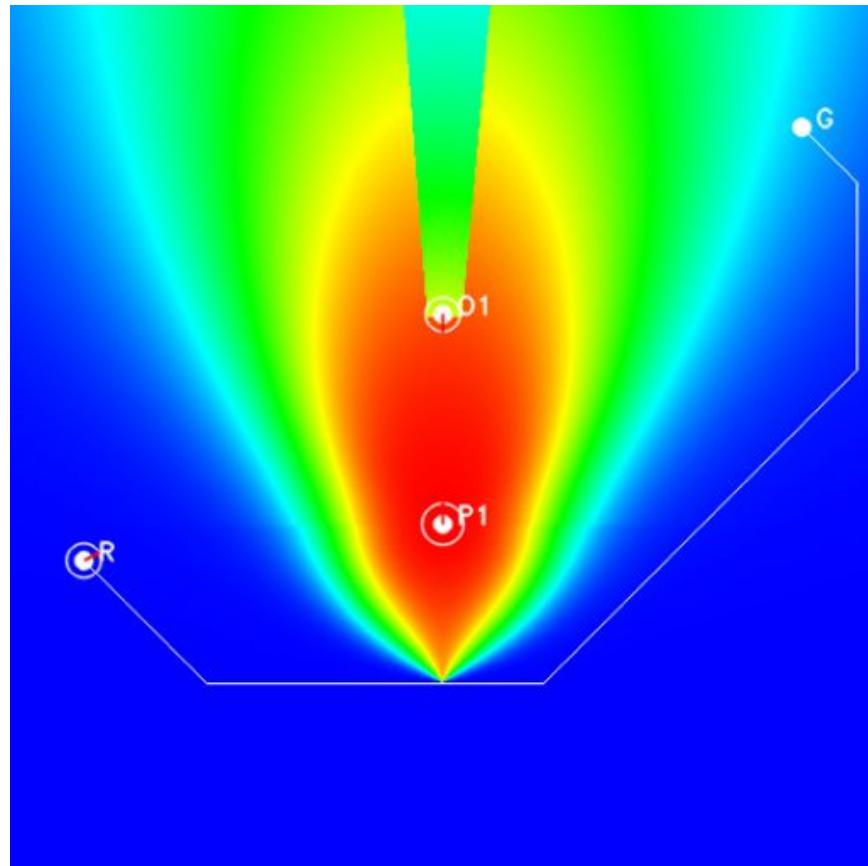
Attention matrix

First results: Attention Field

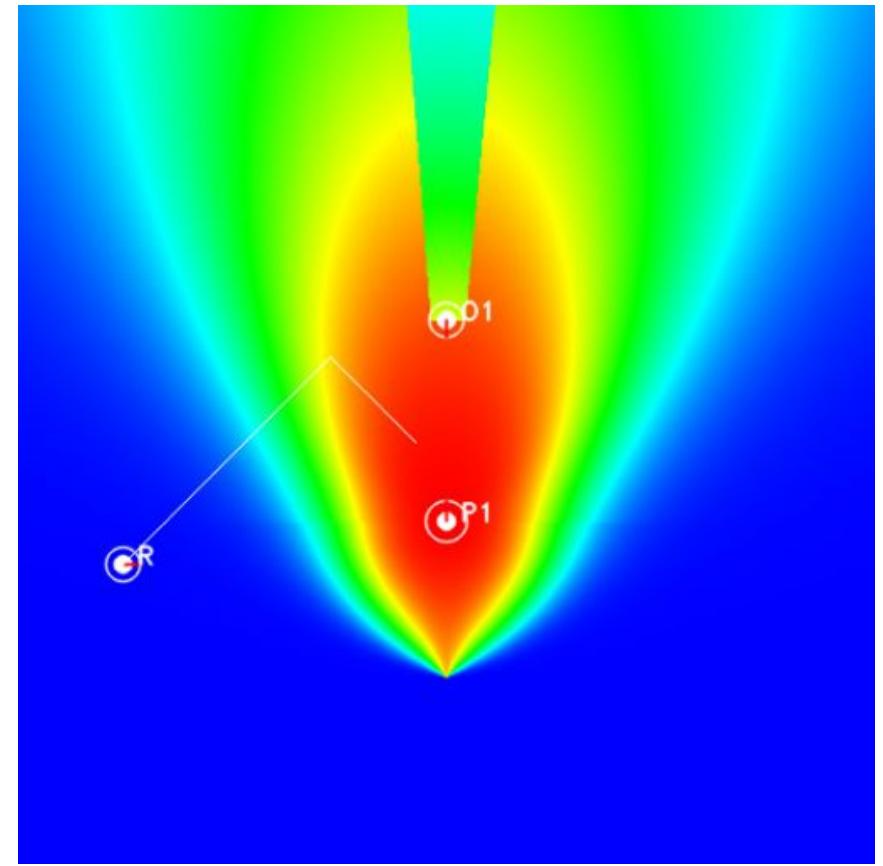


T. Fraichard, R. Paulin, and P. Reignier, *Human-robot motion: An attention-based navigation approach* (ROMAN14)
Nominated for best paper award

Navigation based on the *attention field*

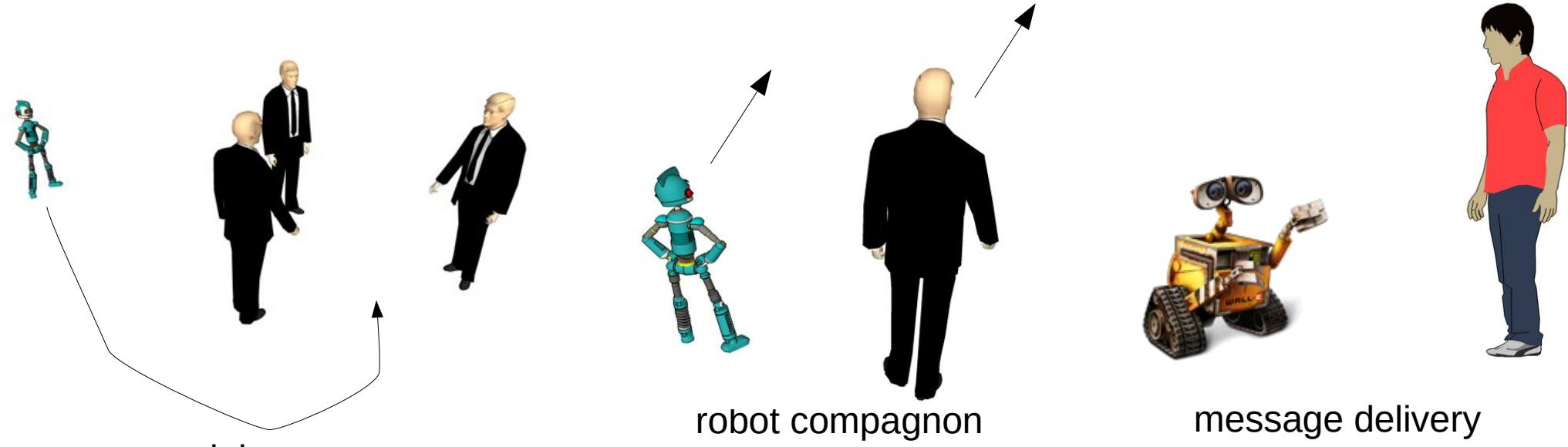


do not disturb



attract attention

Ongoing work



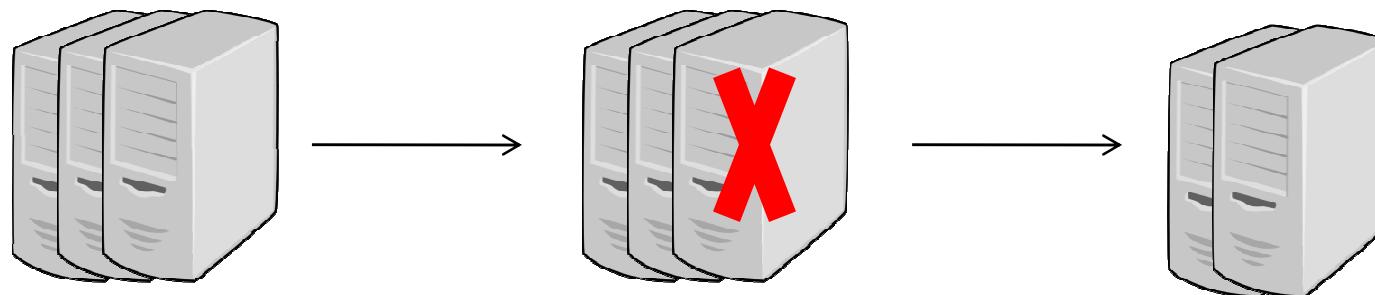
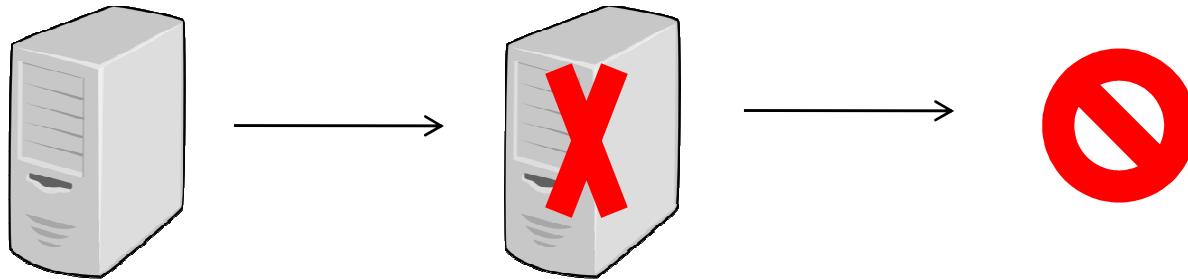
approach validation



D'une tolérance aux fautes byzantines
efficace à une tolérance aux fautes efficace

Lucas.perronne@imag.fr

Tolérance aux fautes via réPLICATION d'état machine



Tolérance aux fautes byzantines

A l'attaque mes cocos !



Colonel « le fidèle » coco

A l'attaque mes cocos !



Général gégé



Colonel « le dissident » coco

Tolérance aux fautes byzantines



Colonel « le fidèle » coco



Dois-je attaquer ?
Dois-je battre en retraite?
A qui dois-je faire confiance?

Gégé a dit “A l'attaque!”

Gégé a dit “RETRAITE !!!”



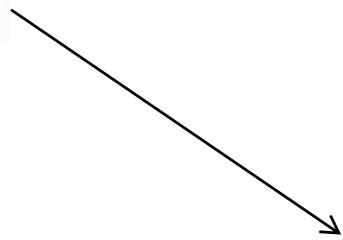
Colonel « le dissident » coco



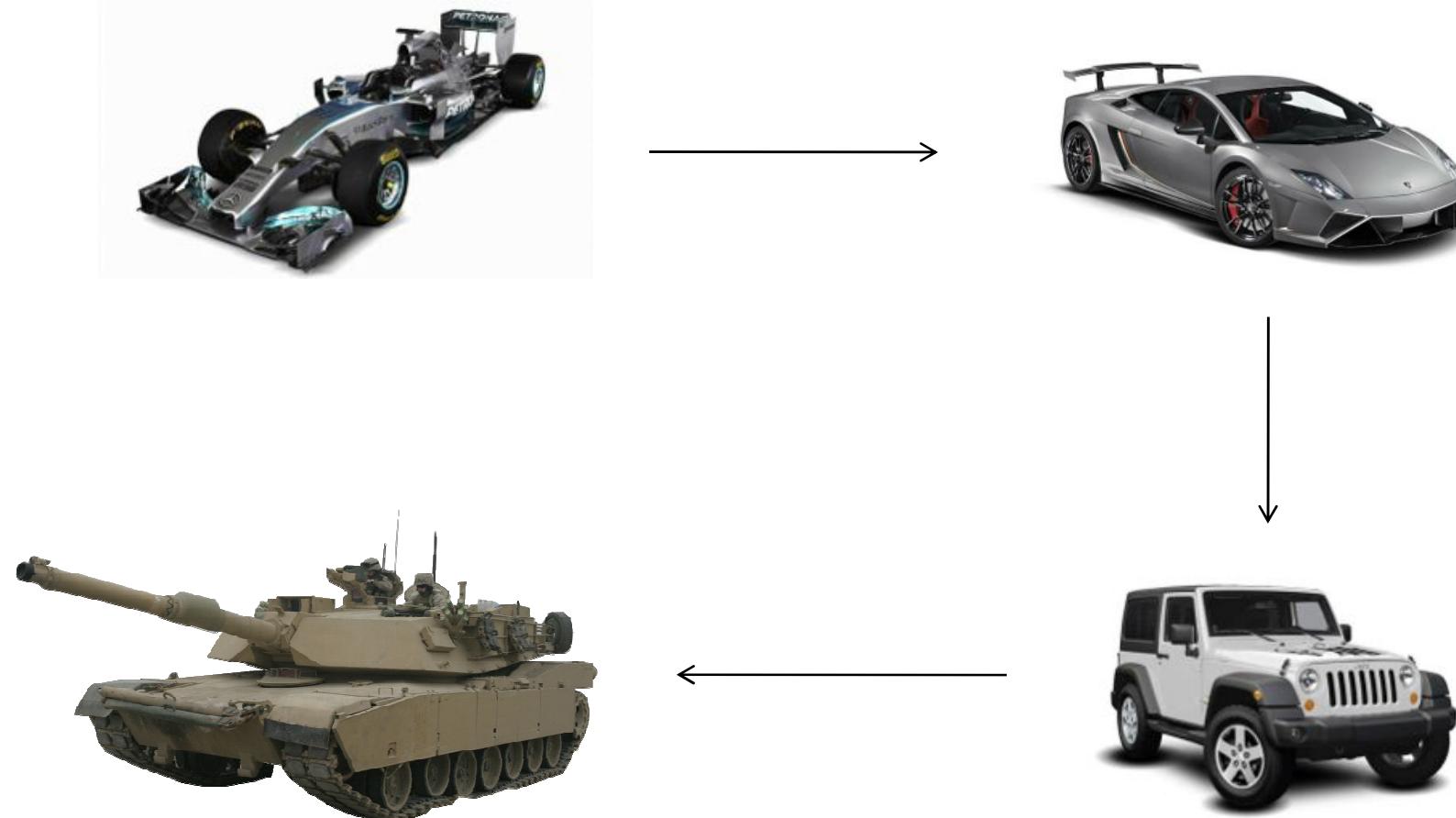
Tolérance aux fautes byzantines : $3f+1$



Tolérance aux fautes byzantines efficace



Tolérance aux fautes ~~byzantines~~ efficace



Journée des doctorants 2015

Validation conjointe en UML et B de la sécurité des SI

Amira RADHOUANI

Directeurs de thèse: Yves LEDRU



Akram IDANI

Narjes BEN RAJEB

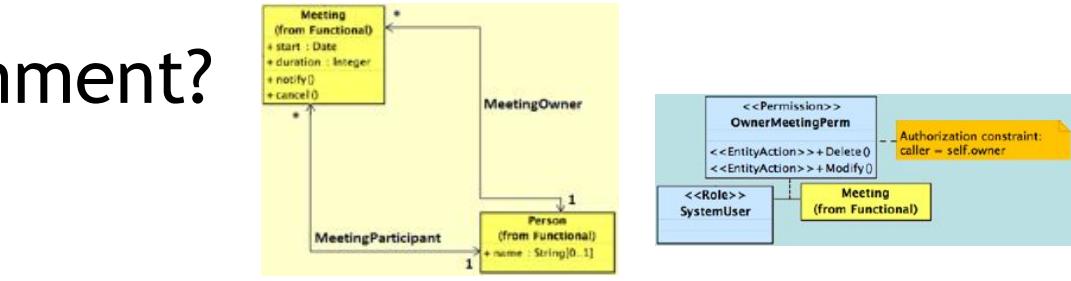


Contexte du travail

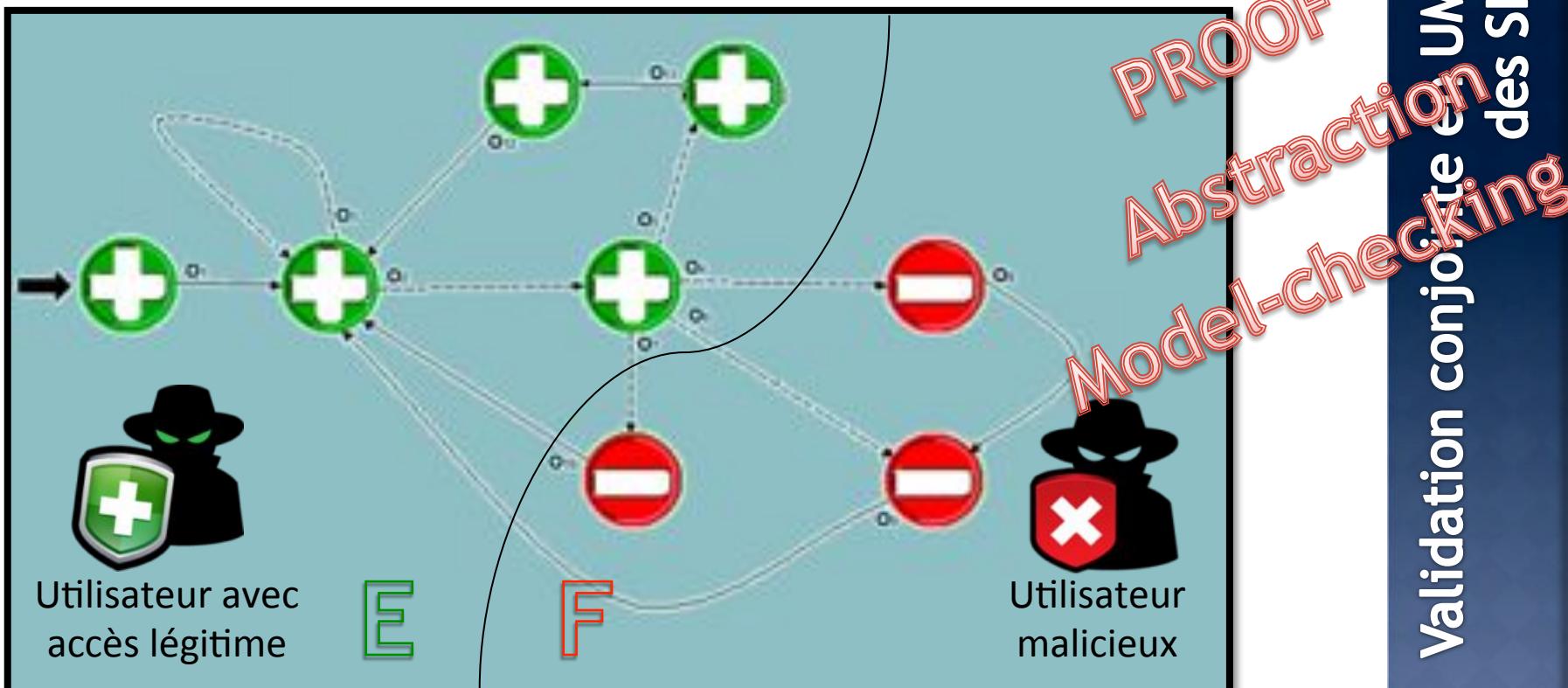


Validation conjointe en UML et B de la sécurité des SI

Validation conjointe en UML et B de la sécurité des SI



Traduction en spécification formelle



A+
@poster session





Peinture numérique 3D pour novices

Elisabeth Rousset / Equipe IIHM



Journée des doctorants 2015
Laboratoire d'Informatique de Grenoble



Un public amateur



123D



123D Make



123D Sculpt



123D Catch



SCULPTRIS

La création de contenu 3D

La création de contenu 3D

➤ Modélisation

La création de contenu 3D

- Modélisation
- Application de texture

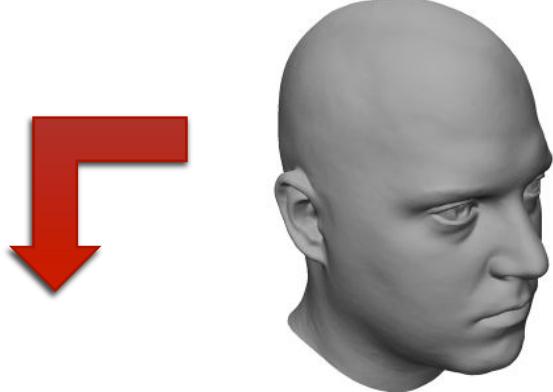
La création de contenu 3D

- Modélisation
- Application de texture
- Animation

La création de contenu 3D

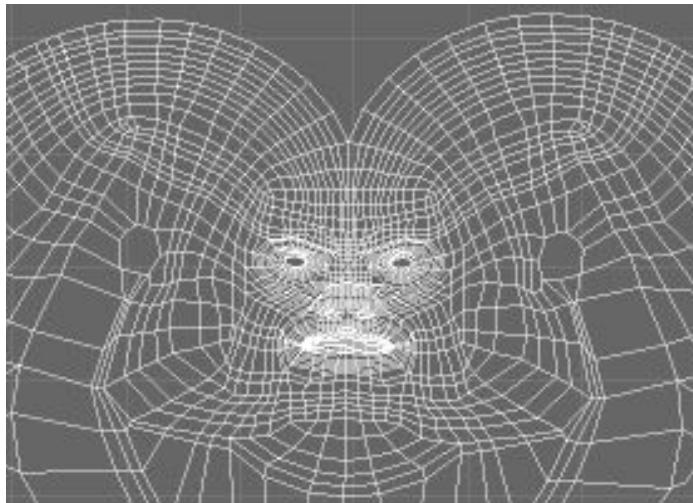
- Modélisation
- Application de texture → Peindre un modèle 3D
- Animation

Application de texture



↗ Dépliage de modèle et peinture 2D

Problème: peindre sur une surface déformée



Application de texture

↗ Peinture 3D par projection

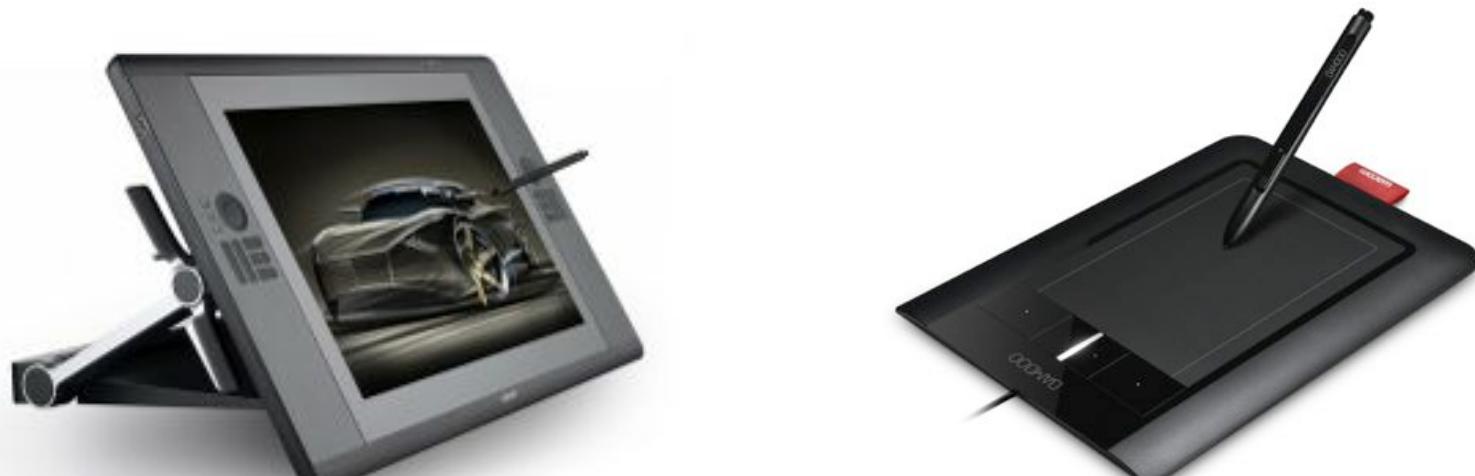
Problème: manipuler le modèle de manière intuitive



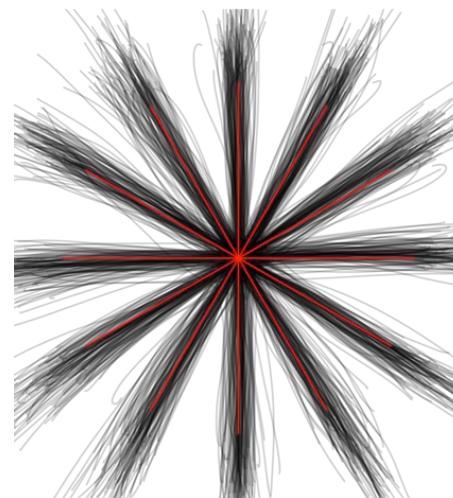
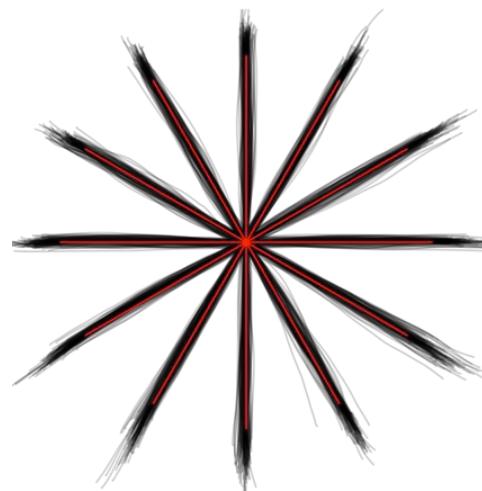
Objectif

- ↗ Rendre la manipulation 3D transparente pour la tâche de peinture
 - ↗ Etudier le matériel adapté
 - ↗ Améliorer l'interaction 3D

Dispositifs de peinture numérique



Dispositifs de peinture numérique





Merci





STEAMER

Géovisualisations pour la représentation des dynamiques spatiales

Application au risque d'inondation impactant le système ferroviaire

Cécile Saint-Marc

Thèse CIFRE SNCF – LIG
Ecole Doctorale ISCE

Discipline : Géomatique





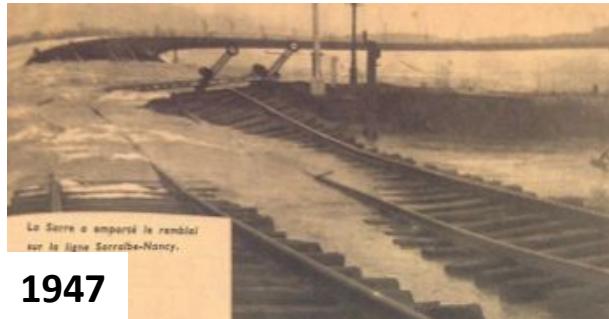
Géomatique



STEAMIEER



Catastrophes naturelles

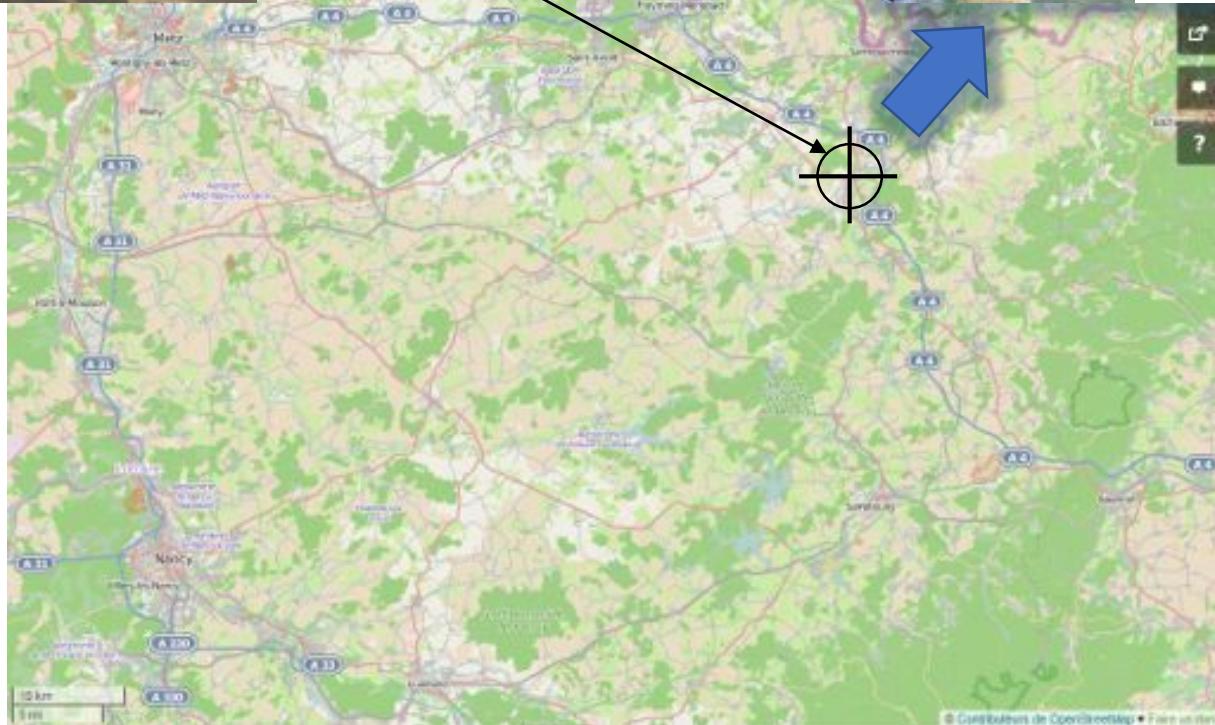


1947

La Serre a emporté le remblai sur la ligne Seraing-Nancy.



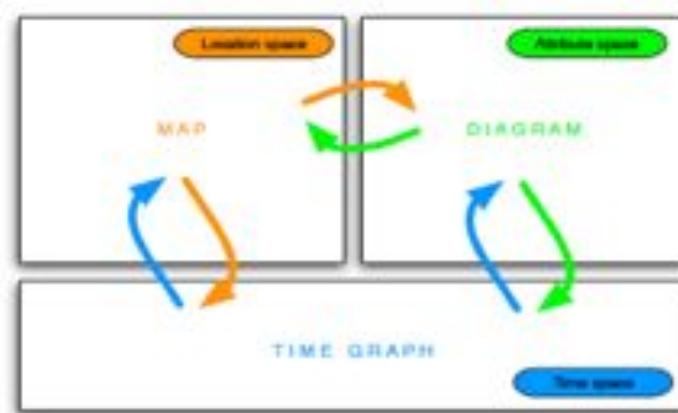
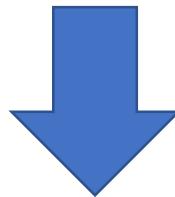
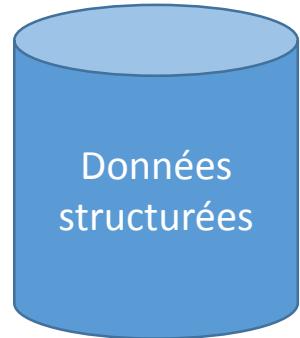
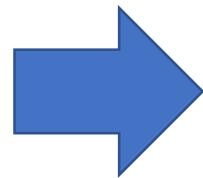
2010



STEAMER



Enjeux



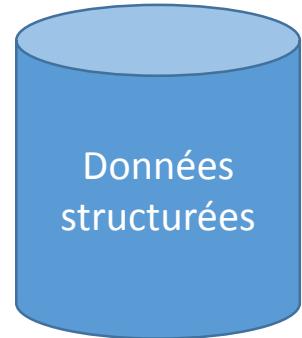
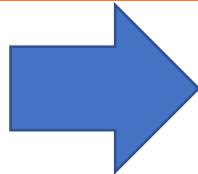
Géovisualisation



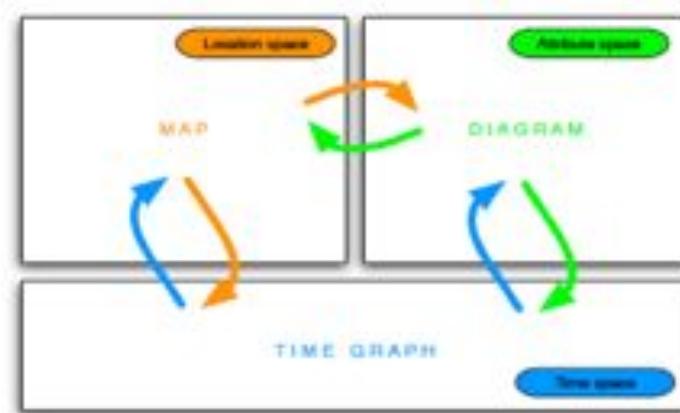
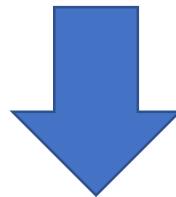
Enjeux



Données hétérogènes,
peu structurées,
imprécises



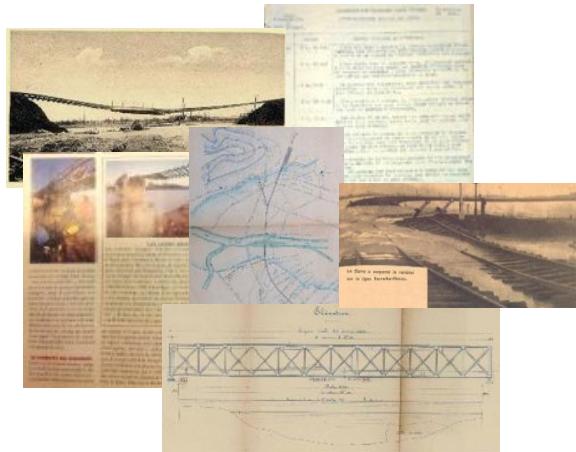
Documents d'archives,
témoignages



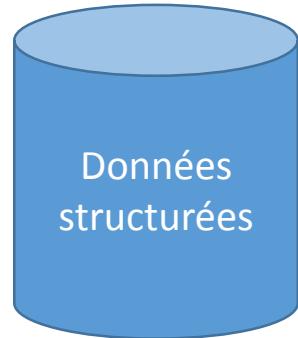
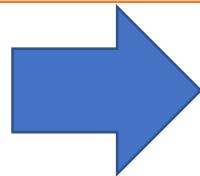
Géovisualisation



Enjeux

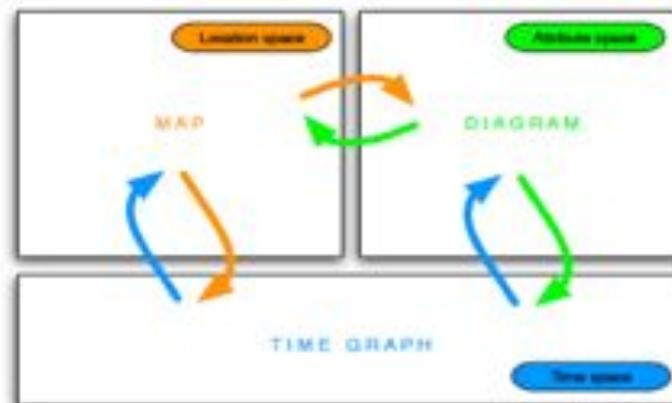
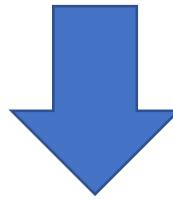


Données hétérogènes,
peu structurées,
imprécises



Documents d'archives,
témoignages

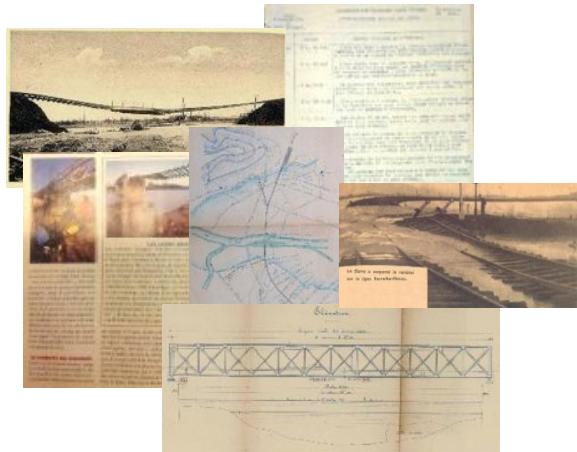
Visualiser la dynamique
spatio-temporelle des
processus pour l'analyser



Géovisualisation

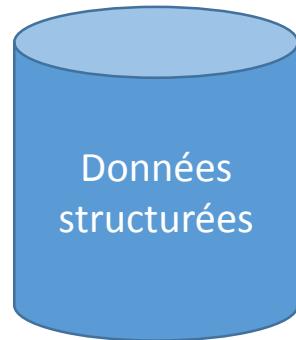
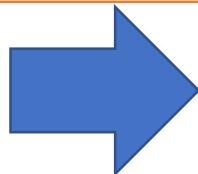


Enjeux



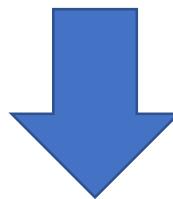
Documents d'archives,
témoignages

Données
hétérogènes,
peu structurées,
imprécises

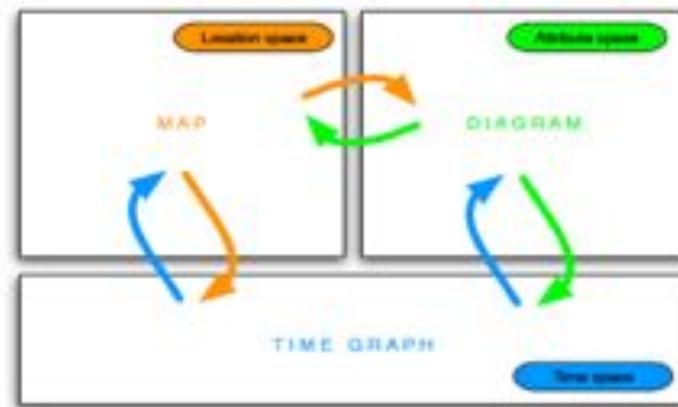


Données
structurées

Visualiser la dynamique
spatio-temporelle des
processus pour l'analyser



Visualiser les relations
entre phénomènes variés



Géovisualisation

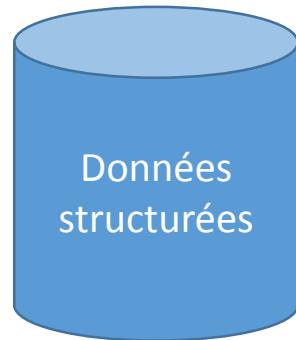
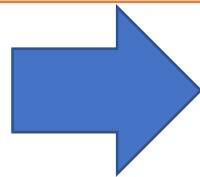


Enjeux

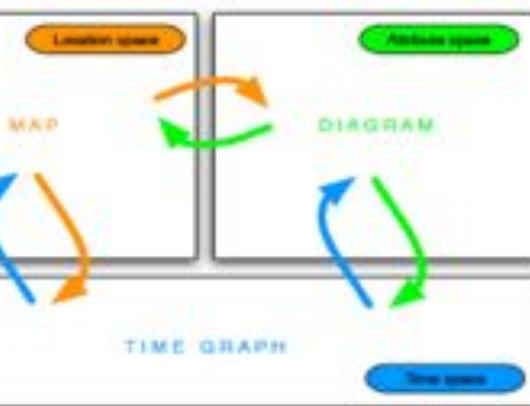
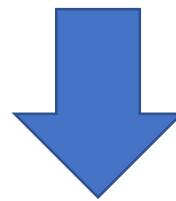


Documents d'archives,
témoignages

Données
hétérogènes,
peu structurées,
imprécises



Visualiser la dynamique
spatio-temporelle des
processus pour l'analyser



Géovisualisation

Visualiser les relations
entre phénomènes variés

Rendre compte des
imperfections



Objectifs

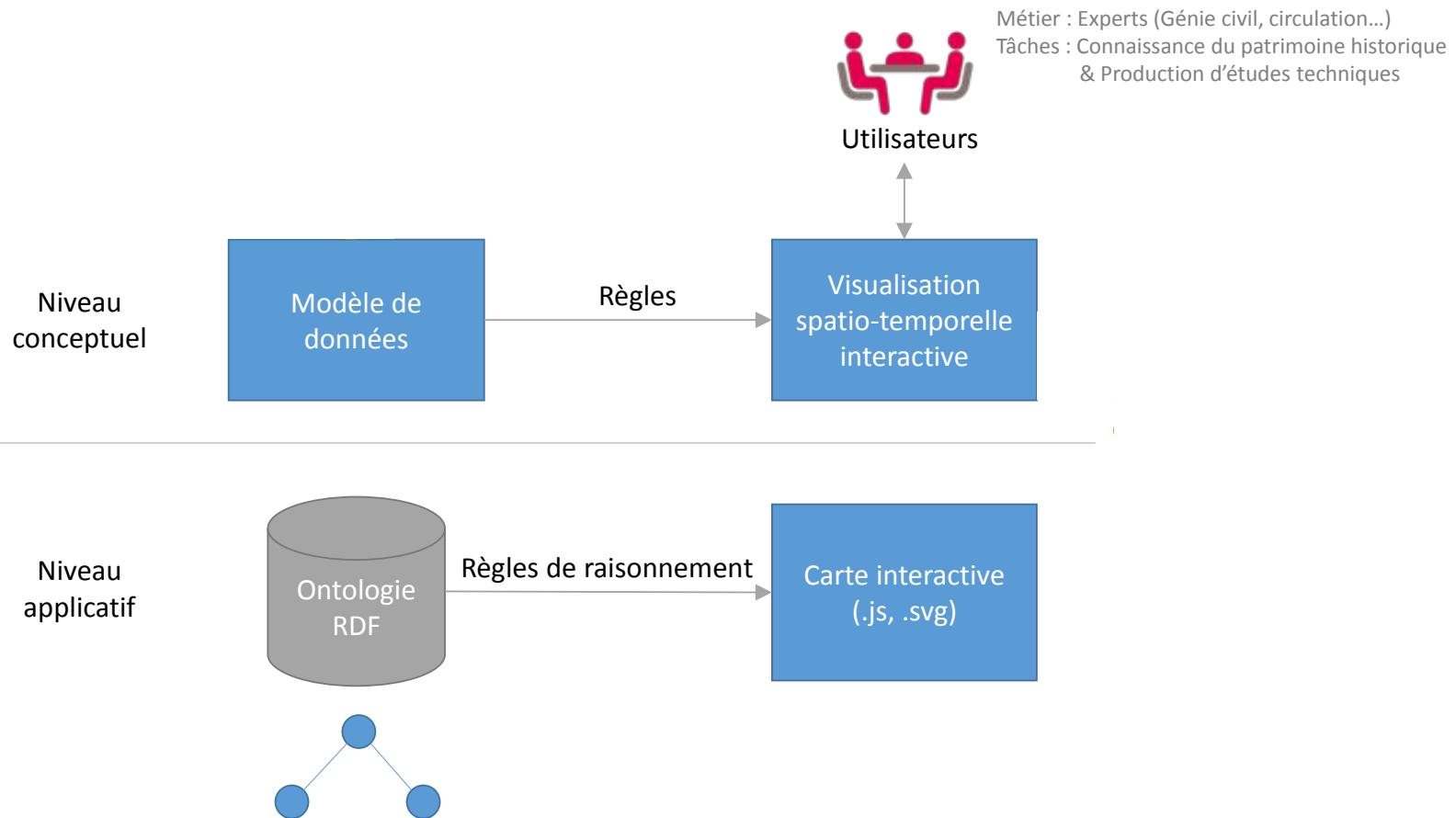
- ☒ **Définir des modes de visualisation spatio-temporelle**, adaptés aux spécificités des données historiques et aux utilisateurs, pour faciliter l'extraction de connaissances sur les dynamiques liant phénomènes spatiaux et système ferroviaire
- ☒ **Systématiser le processus de construction de géovisualisation de dynamiques** en prenant en compte les caractéristiques des données et les besoins des utilisateurs

Cas d'application : inondations majeures ayant impacté le système ferroviaire.

STEAMER



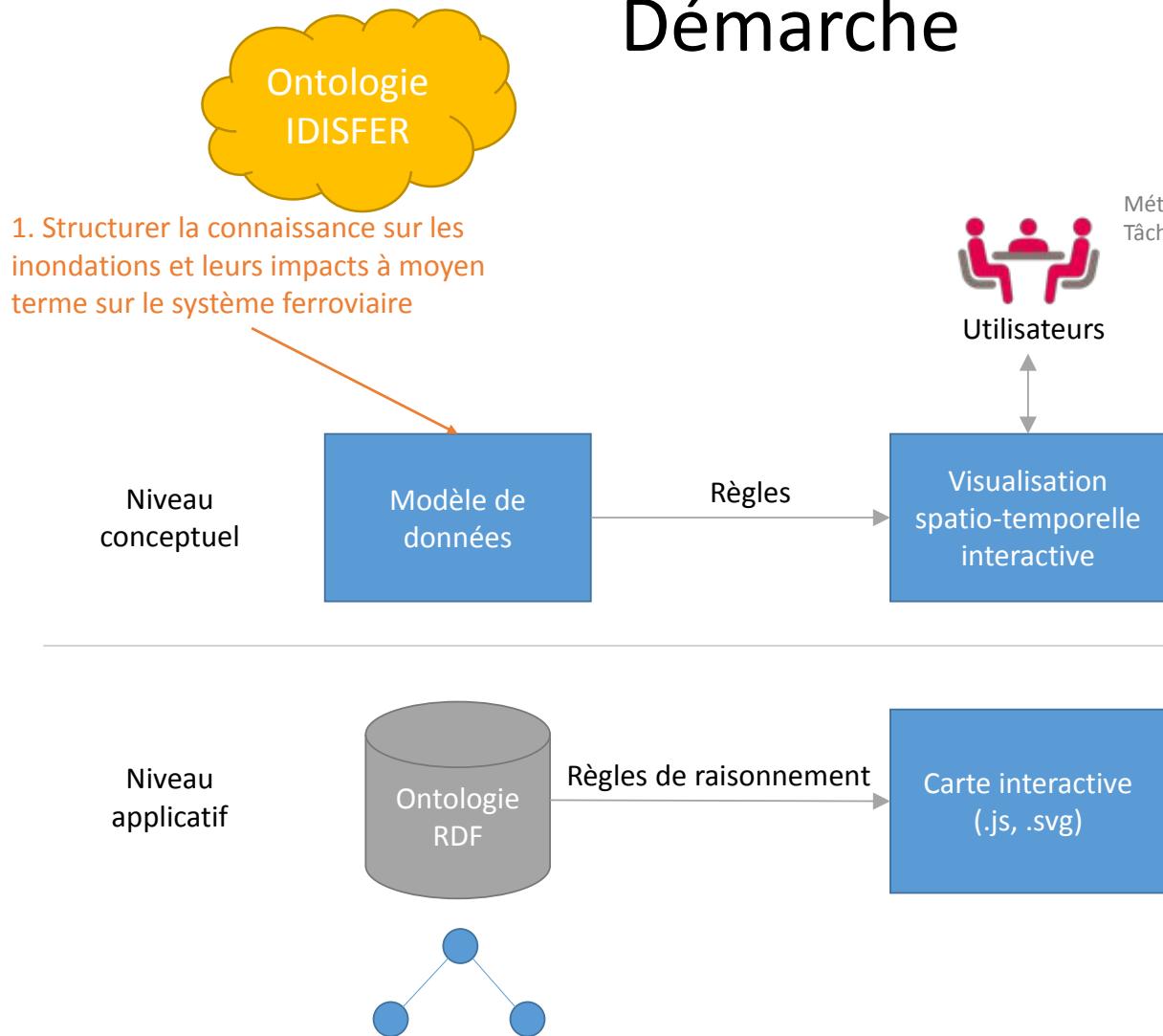
Démarche



STEAMER



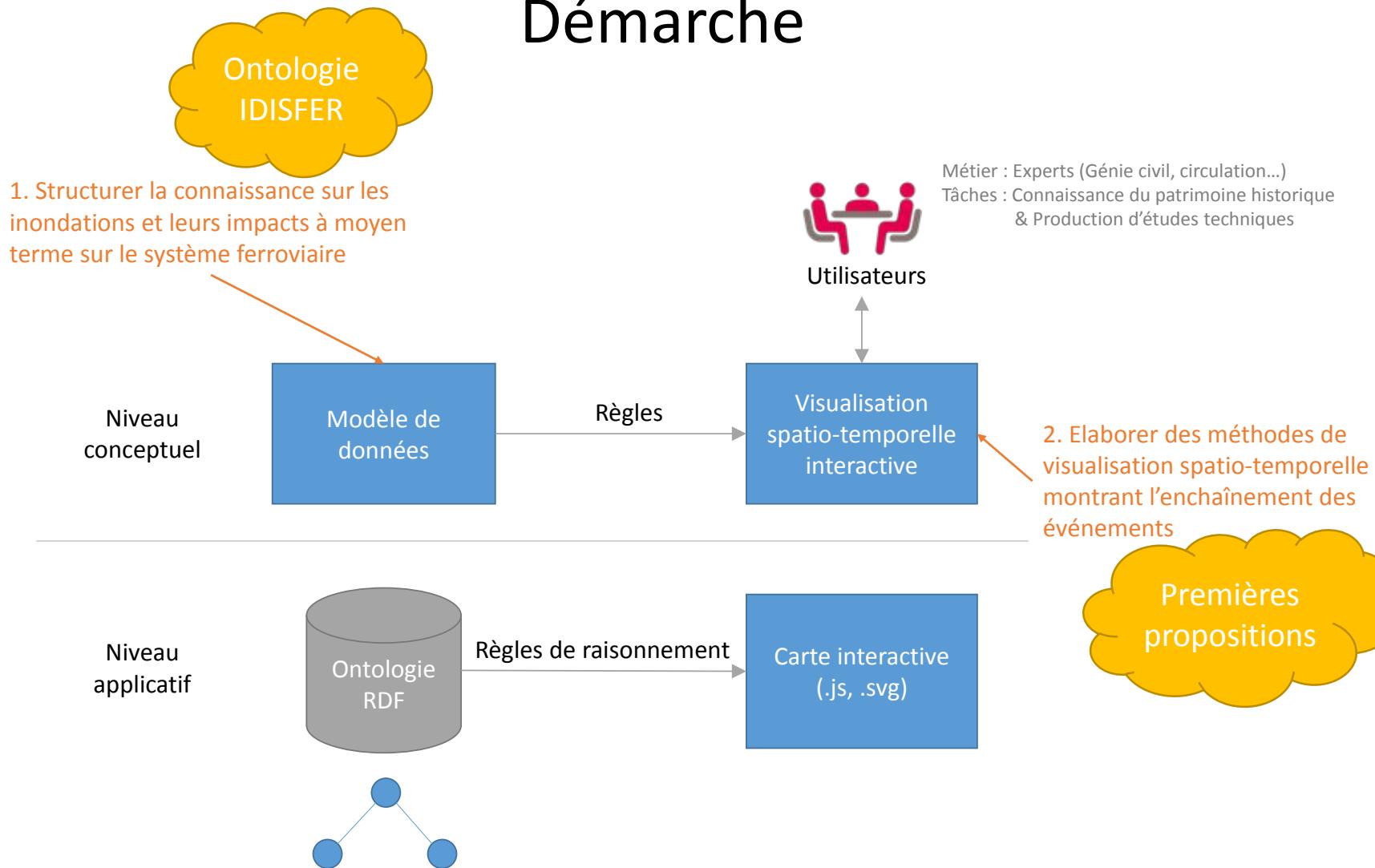
Démarche



Métier : Experts (Génie civil, circulation...)
Tâches : Connaissance du patrimoine historique & Production d'études techniques

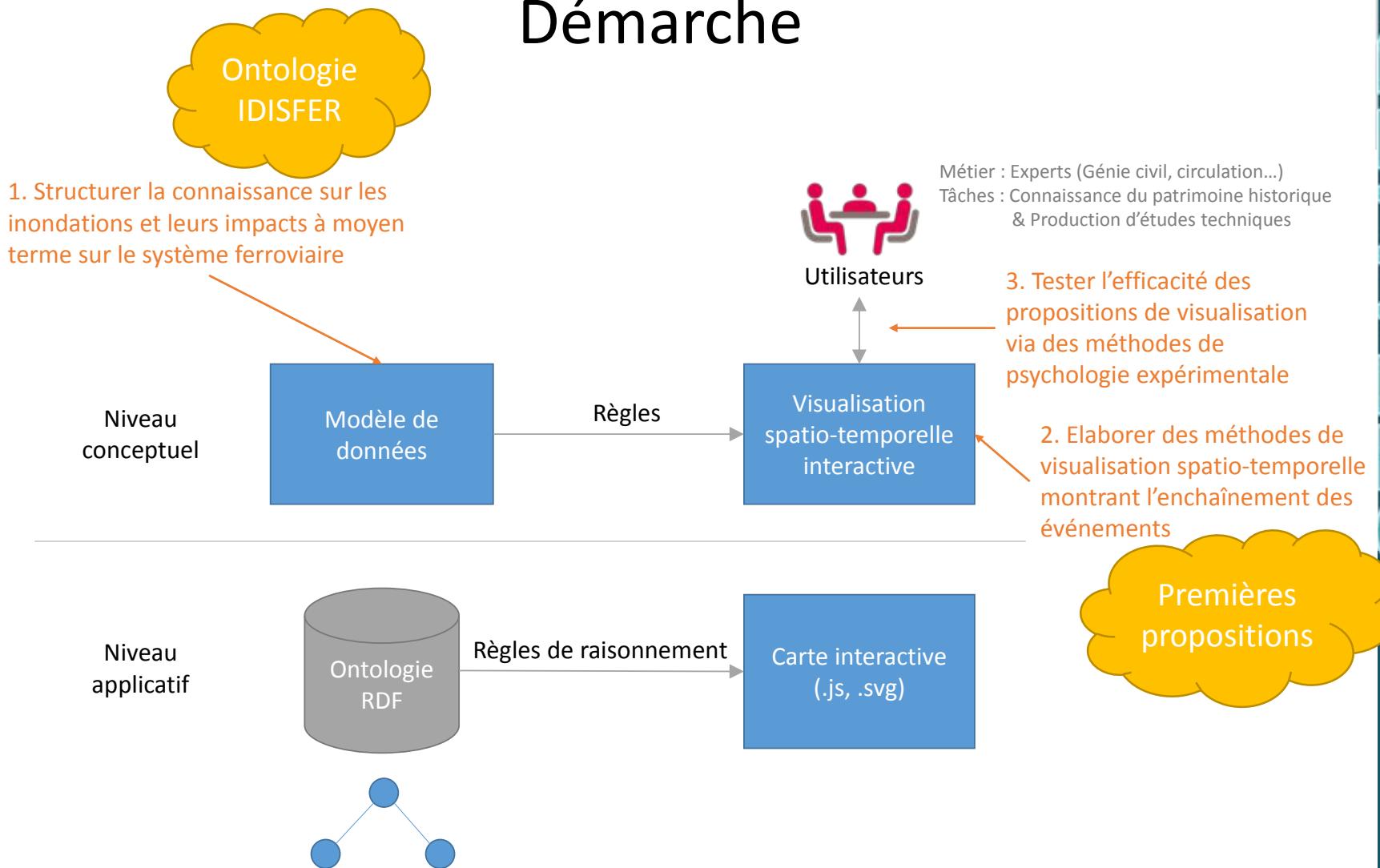


Démarche



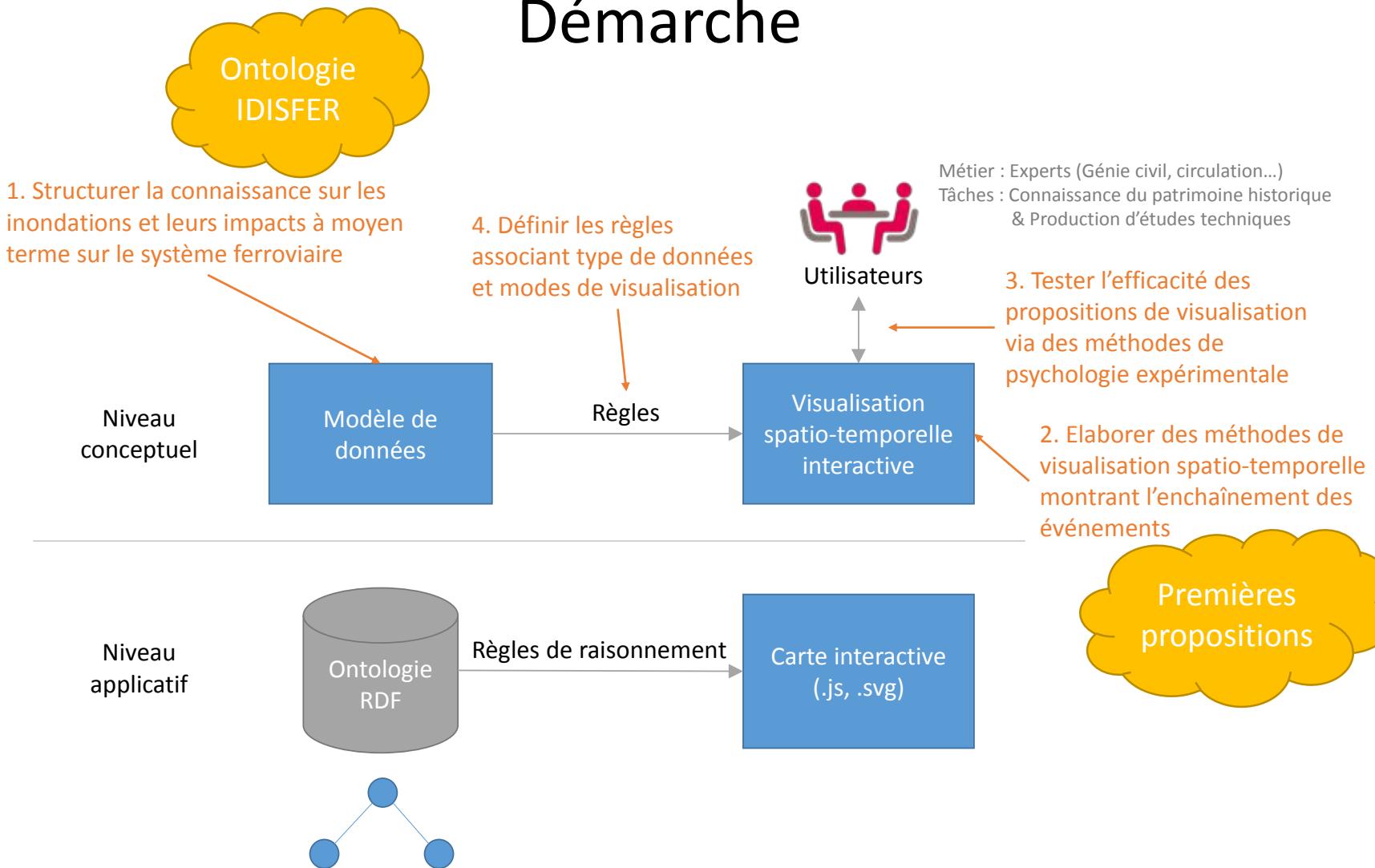


Démarche



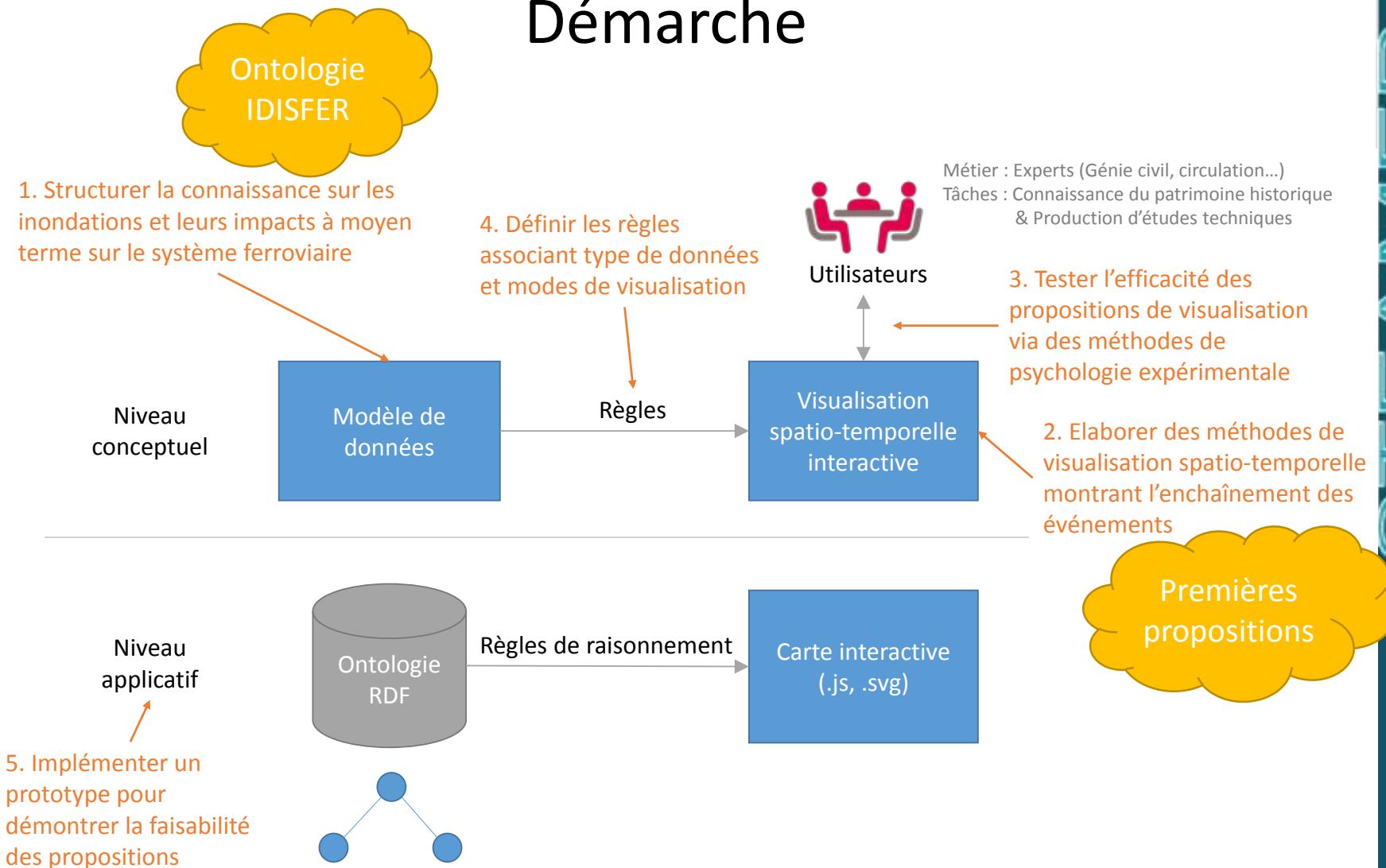


Démarche





Démarche



Towards Matching Improvement between Tasks and Workers in Spatial Crowdsourcing Systems

André Sales Fonteles

Advisors: Jérôme Gensel and Sylvain Bouveret

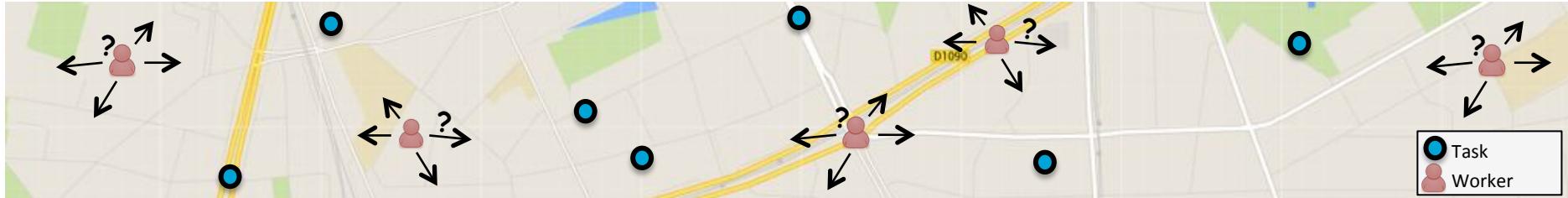


MINISTÈRE
DE L'ENSEIGNEMENT SUPÉRIEUR
ET DE LA RECHERCHE

Context



Context



Matching Points of View

System

Maximize overall number of tasks accomplished.

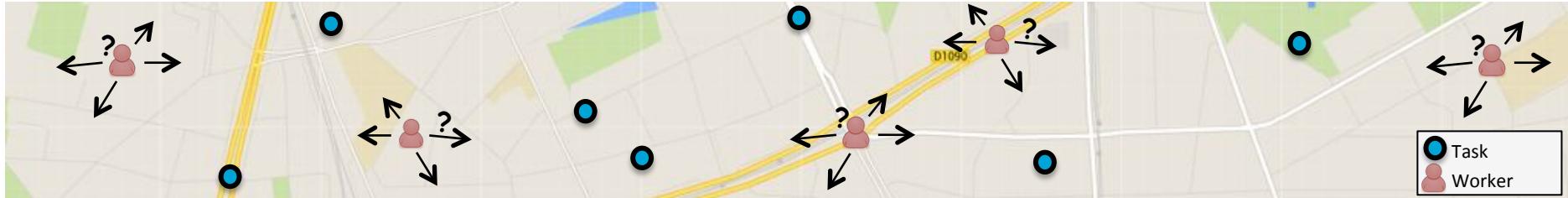
Worker

Tasks that best match worker's preferences.

Task

Workers that best match task required skills.

Context



Matching Points of View

System

Maximize overall number of tasks accomplished.

Worker

Tasks that best match worker's preferences.

Task

Workers that best match task required skills.

Current Objective

Help workers to find spatiotemporal tasks, and/or a sequence of them

- To improve the **overall contribution** of a worker.
- To increase the **quality of the service** provided by workers.

Proposal

Single Task Recommendation (Fonteles et al., 2014, MobiGIS) (Fonteles et al., 2014, SAGEO)

Worker w

The interface shows a worker profile icon (purple person) and a smartphone icon. Below is a section titled "History" containing a list of completed tasks:

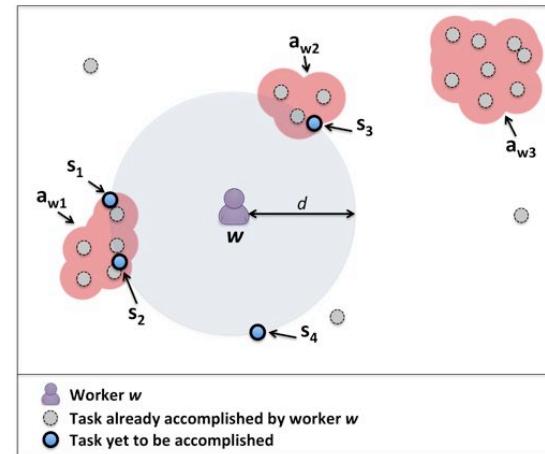
- picture..... photo
- take a picture.
- landmarks ...
- take a photo..... camera

Below the history is a "Task" card with a red dashed border:

Task \$1.40
Take a picture of the intersection of streets ...

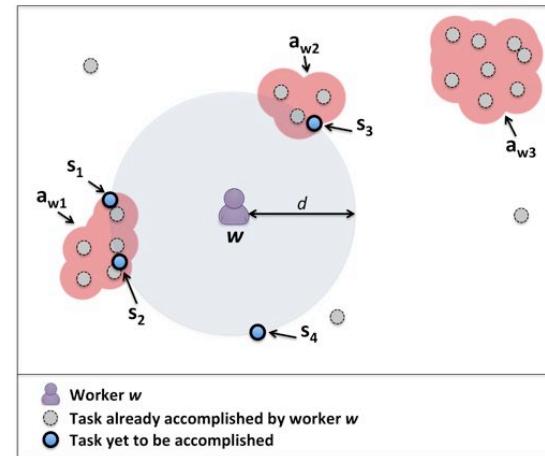
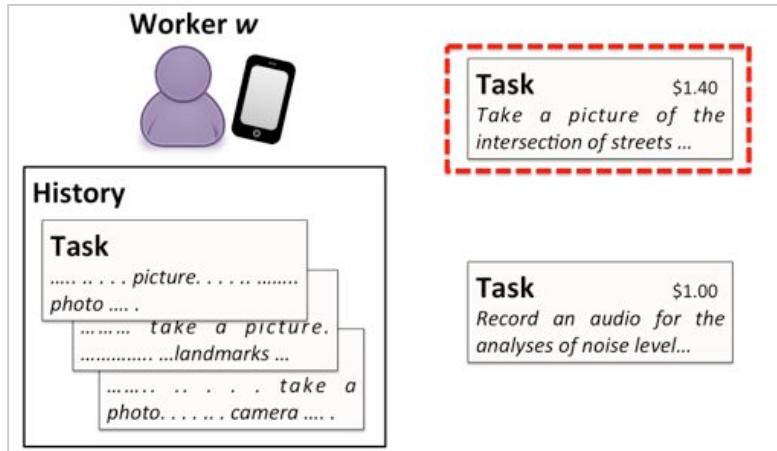
Below this is another task card:

Task \$1.00
Record an audio for the analyses of noise level...

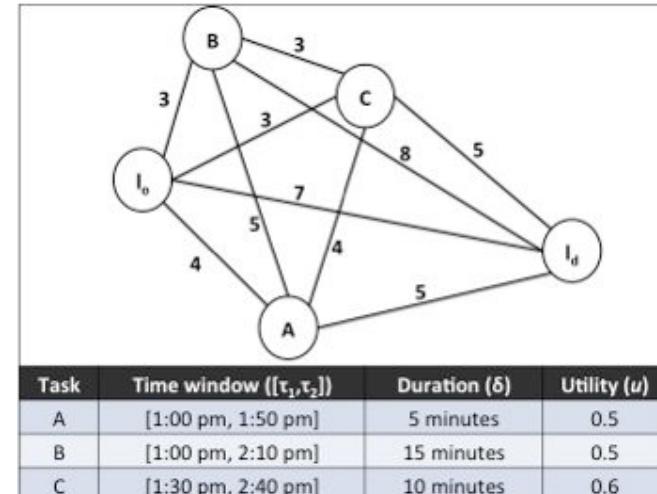
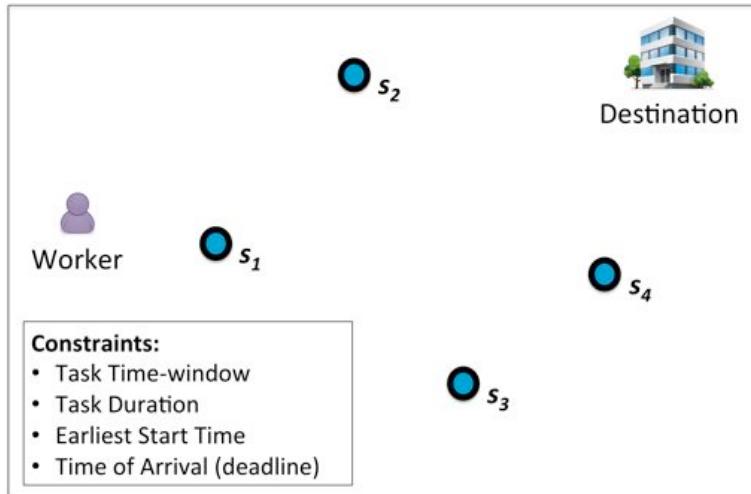


Proposal

Single Task Recommendation (Fonteles et al., 2014, MobiGIS) (Fonteles et al., 2014, SAGEO)



Task Sequence Recommendation (Fonteles et al., 2015, W2GIS)



Towards Matching Improvement between Tasks and Workers in Spatial Crowdsourcing Systems

*For more information:
Come to see the poster*

Merci!





Journée des Doctorants du LIG
26 Mars 2015



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AGENCE NATIONALE DE LA RECHERCHE



Socio-affective intelligence for a robot :

**interaction vocal and gestural primitives for a
personalizable robot in a smart home**

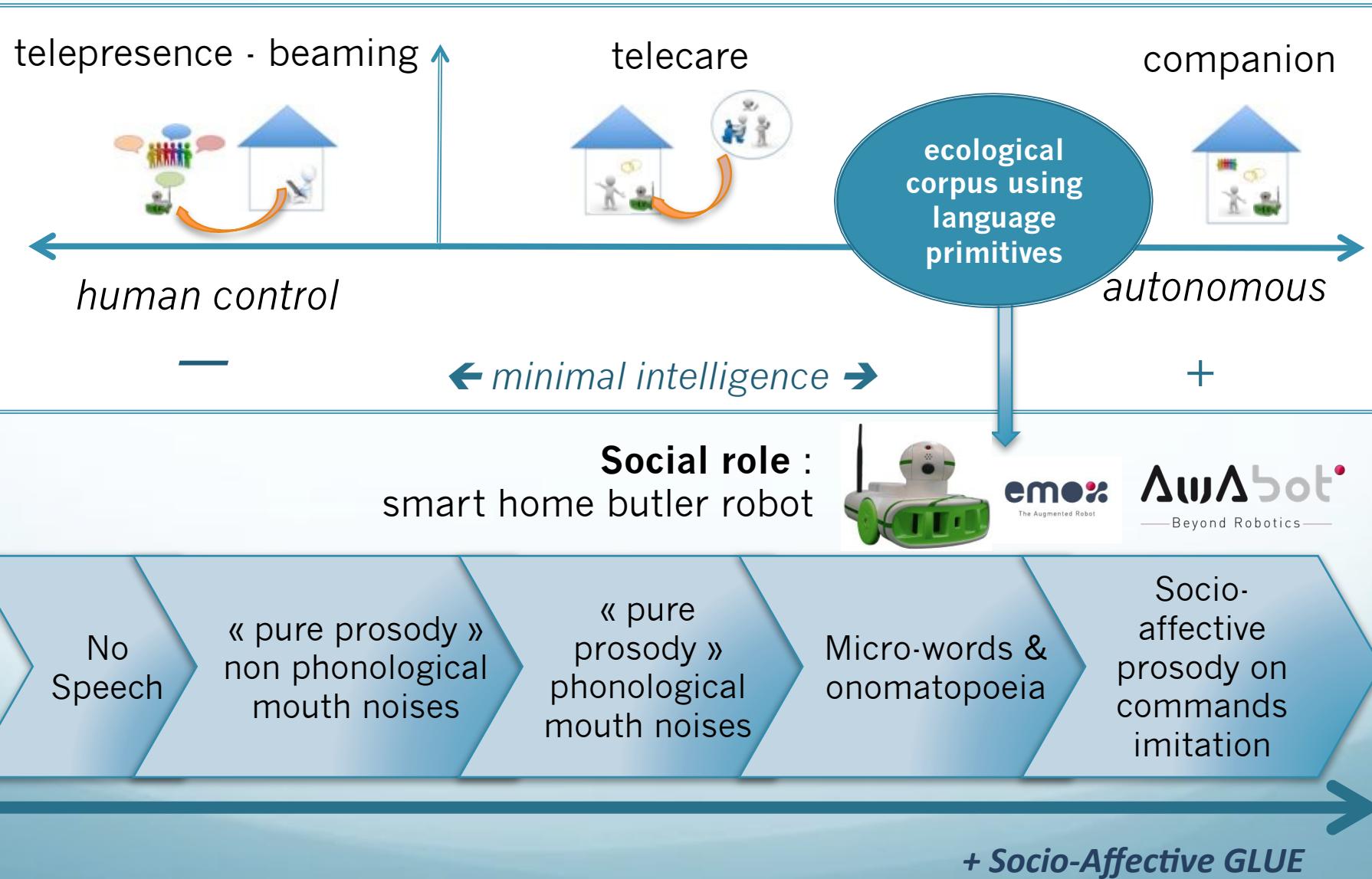
Yuko SASA

Supervisor: Véronique AUBERGÉ (LIG - CNRS)

Co-supervisor: Gang FENG (Gipsa-Lab)

Co-advisor: Yoshinori SAGISAKA (Waseda University)

Context & paradigms



Experiments

EEE: Elderly Emox Expressions corpus

Let socio-isolated elderly talk to Emox with graduated socio-affective prosody, would it become a communication prothesis?

Scenario

- 1) Visit of Domus all together
- 2) Prepare to leave the elderly alone (experimenters acting)
- 3) Introduction to the voice commanded Smart Home and its « butler » Emox
- 4) Emox-Elderly Interactions
- 5) Return of the experimenters and their accomplices
- 6) Debriefing



EmOz : Wizard of Oz interface



Emox Robot control



Domus Smart Home control



Domus: Smart Home / living-lab

Results and perspectives

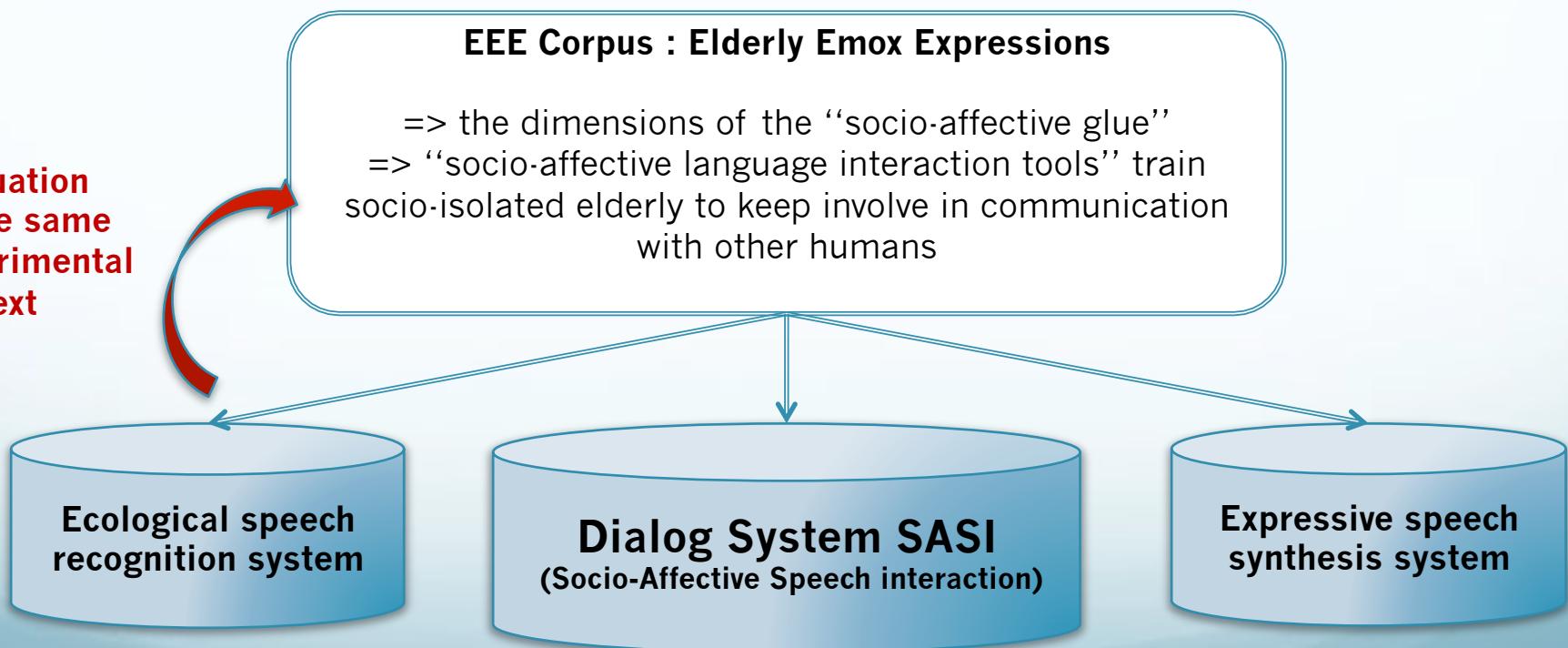
Before “glue”

=> Reading attitudes

After “glue”

- ⇒ Commands paraphasings
- ⇒ Prosodic focuses and characteristics
- ⇒ Caring, politness, guidance cues
- ⇒ Interaction tools change dynamically...

Evaluation
in the same
experimental
context



Thank you for your attention

Hope to see you at the poster
session!

Time Series Centroid Estimation under weighted and kernel dynamic time warping

Saeid SOHEILY KHAH

Supervisors : Ahlame DOUZAL , Eric GAUSSIER

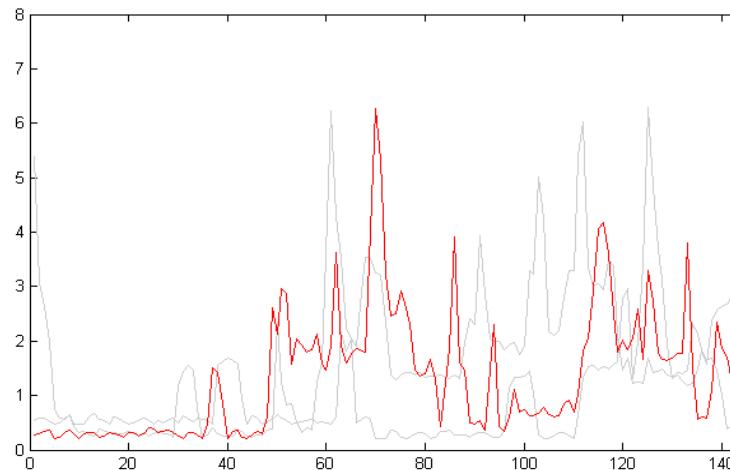


Time Series

A kind of **sequence** data:

- an ordered sort of elements
- order criteria : ***time***

(exp. power consumption)



Averaging a set of time series is involved in many data mining and machine learning processes as:

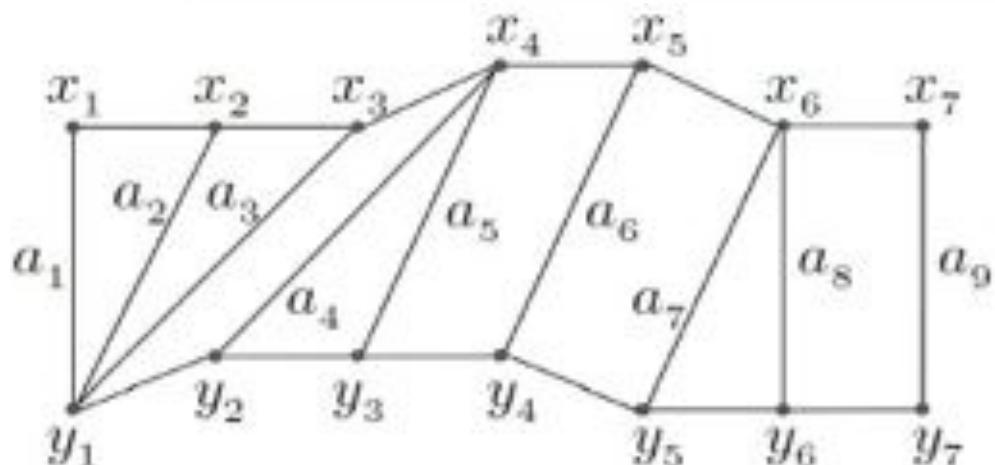
- **Summarizing** a set of time series
- Extracting temporal prototype
- **Clustering** time series

Challenging Question

centering more than two times series under temporal warping

$n = 2$

- pairwise alignment [one standard way]



	1	2	...	9
x	x_1	x_2	...	x_7
y	y_1	y_2	...	y_7
c	$\text{avg}(x_1, y_1)$	$\text{avg}(x_2, y_1)$...	$\text{avg}(x_7, y_7)$

$n > 2$

- need to handle the problem of multiple temporal alignment

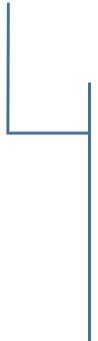
Objective

Formalize the multiple time series averaging problem as an optimization problem

Solution

Propose an **optimal solution** for the barycenter estimation under:

- weighted DTW metric
- kernel DTW metric

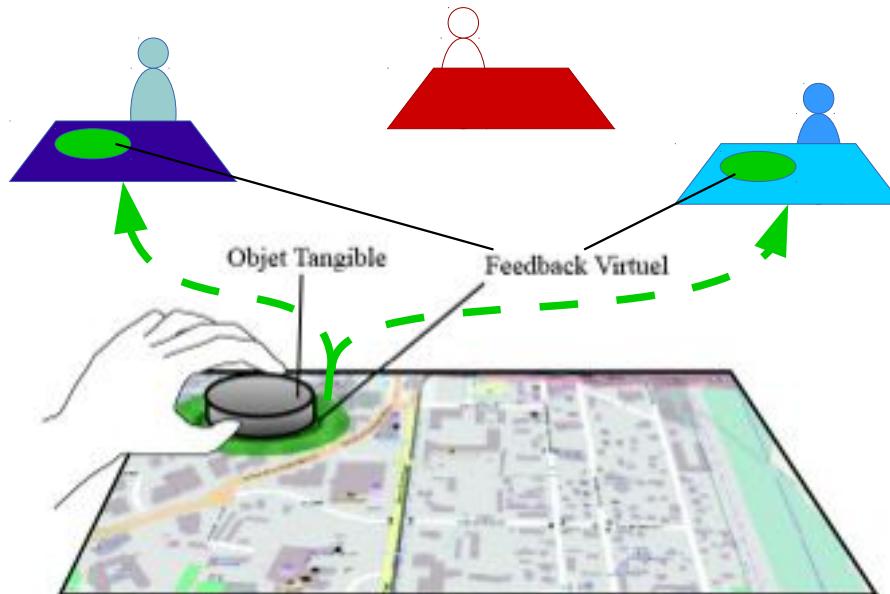


That allow us to estimate:

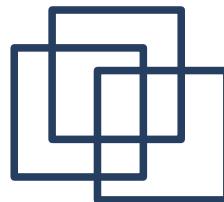
- the time series centroid
- its weight vector



Décisions collectives en environnements interactifs et collaboratifs complexes : Application à la gestion de crise



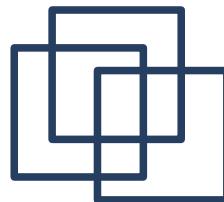
Lauren Thévin
LIG-AMA
LIG-Magma
EMSE-Institut Henri Fayol/ISCOD,



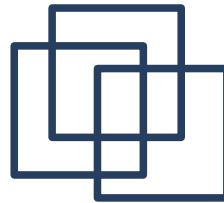
Contexte

- Gestion de crise :
 - Risques majeurs(naturels, industriels, transport)
 - Acteurs secours et sauvegarde (pro et non pro)
→ Plan communal de sauvegarde
- Exercice avec Table TangiSense:
 - test et appropriation



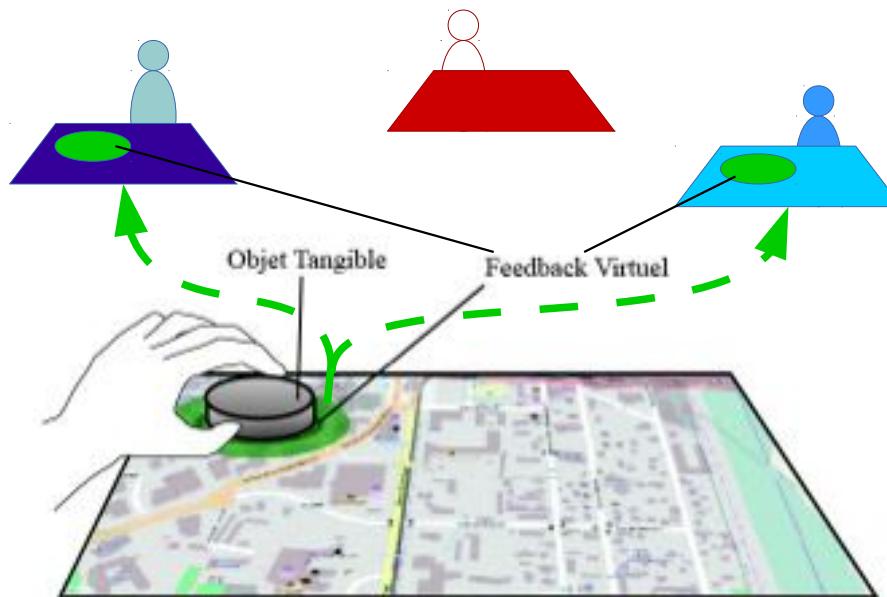


- Points de vue organisationnel multiples
 - Partage : Environnement et Organisation
 - Aide : résolution de conflit
- Coordination à distance
 - Interaction tangible
 - Retour virtuel



Merci de votre attention

- Pour en savoir plus, rendez-vous à la session poster

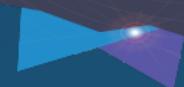




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PERSYVAL -Lab

Software Transactional Memory with Autonomic Management Techniques



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Gwenael Delaval – *Grenoble University, France*

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March 27th, 2015

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PERSYVAL -Lab

1. Introduction to Parallel Program

Multi-core Processor

- Multi-core processors are everywhere, more parallelisms/concurrency levels give higher performance?
- Many threads execute concurrently. Threads share data. More threads maybe more conflict!

Synchronization VS Computation

A high concurrency level may decline computing time, but increase synchronization time. How to handle the trade-off between synchronization and computation?

1. Introduction to Parallel Program

Locks

A traditional way for synchronization. But:

- Deadlocks, vulnerability to failures, faults...
- Difficult to detect deadlocks
- Hard to figure out the interaction among concurrent operations

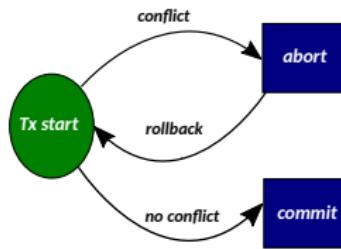
Transactional Memory

Lock-free, therefore no deadlocks! But really? Any problems? Why transactional memory does not become the dominating memory system?

2. Transactional Memory

Concepts

- Shared variables are wrapped by **transactions** (atomic blocks)
- concurrent accesses are performed inside transactions
- Transactions are executed speculatively and can either commit or abort.



2. Transactional Memory

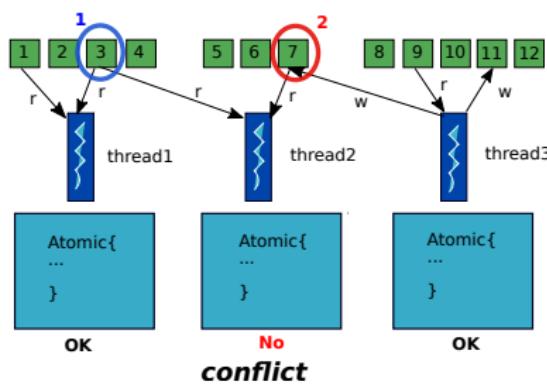
Example

consider three threads read/write data from/to the objects of different memory locations. Access occur inside transactions

how to solve the conflict



I detect the conflict, I wait
then I continue



2. Autonomic Management

Design the system wisely making it:

- **Self-configuration:** a new component learns the system configurations
- **Self-optimization:** seek to improve performance & efficiency
- **Self-healing:** recover from failures
- **Self-protection:** defend against attacks

Questions?