



# Building and representing the patient pathway from multidimensional and large datasets

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# Introduction

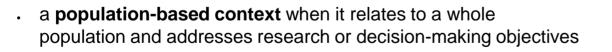
## **Clinical pathways**

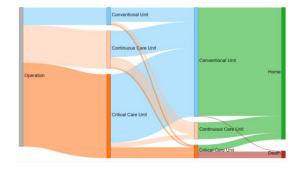
Clinical pathway : sequence of interventions from which the patients benefit during their encounters with health care structures.

Two contexts:

• an **individual-based context** when the visualization targets a single patient to provide appropriate care

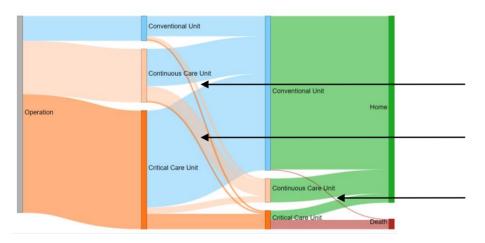
Unit A Unit B Discharge
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## Clinical pathways – Some examples

Clinical pathway : sequence of interventions from which the patients benefit during their encounters with health care structures.



#### Post-operation period

Atypical pathways :

- patients transfered too late to acute care units
- patients died in unmonitored units

## Barriers

#### Different entities with different dimensions

patient			
patient_i d	age	sex	bmi
1	25	М	22.3
2	78	F	25.1
3	54	М	
4	65	М	
5	19	F	

	hos	pital stay	
stay_id	patient_id	admission	discharge
1	1	23/09/2019	30/09/2019
2	2	20/09/2019	10/10/2019
3	1	12/11/2019	20/11/2019
4	3	22/10/2019	25/10/2019
5	4	23/10/2019	27/10/2019

	proce	dure	
procedure_id	visit_id	procedure_date	procedure
1	1	24/09/2019	ADDA200
2	2	24/09/2019	DSZA300
3	3	12/11/2019	BJAD320
4	3	13/11/2019	NJMA421

unit stay			
unit_id	stay_id	admission	discharge
1	1	23/09/2019	25/09/2019
2	1	25/09/2019	30/09/2019
3	2	23/09/2019	25/09/2019
4	2	23/09/2019	25/09/2019
5	2	23/09/2019	25/09/2019
6	3	12/11/2019	13/11/2019
7	3	13/15/2019	15/11/2019
8	3	15/11/2019	20/11/2019

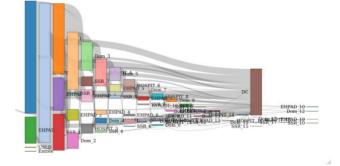
diagnosis
drug
measurement
•••

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High numbers of patients High number of events and of modalities Heterogeneity in duration

increase the complexity of datasets and representations **pattern explosion** 

Event	Number of modalies
Diagnoses	CIM10 : 14 400 codes
Procedures	CCAM : 10675 codes
Drugs	ATC : 6,331 codes
Hospital units	> 500

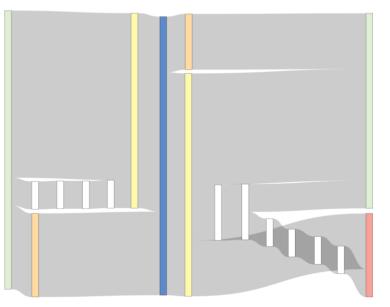


Perer A, Wang F, Hu J. Mining and exploring care pathways from electronic medical records with visual analytics. J Biomed Inform. 2015 Aug;56:369-78. doi: 10.1016/j.jbi.2015.06.020. Epub 2015 Jul 2. PMID: 26146159.

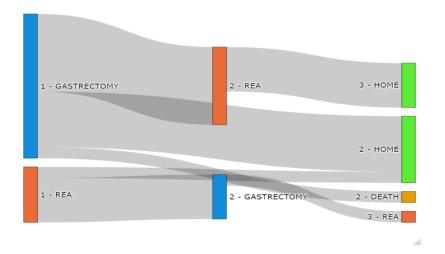
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increase the complexity of datasets and representations

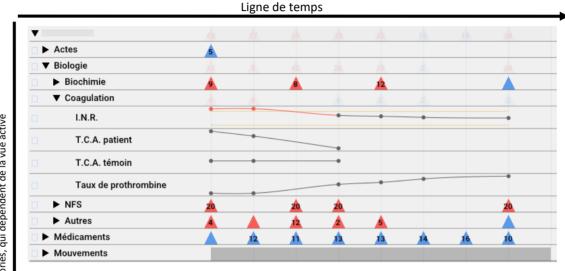
pattern explosion



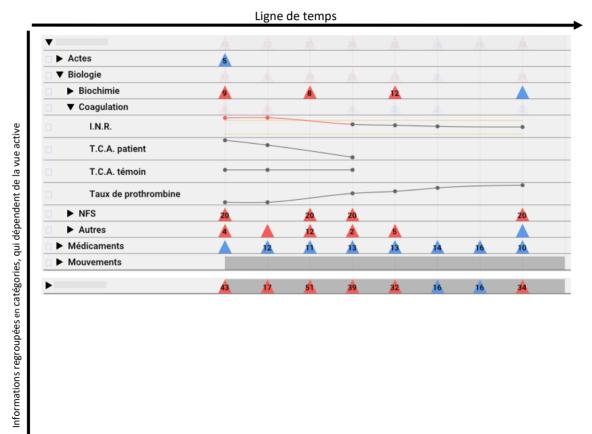
The number of events variable from one individual to another, and each patient encounters the events at its own dateseach patient encounters the events at its own dates.

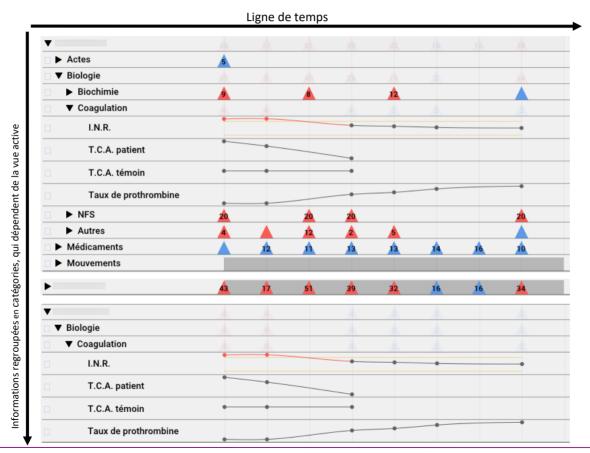


Gastrectomy, post-operative stay in intensive care unit and home stay occur at different ranks of the pathway while they represent the same steps



vue active Informations regroupées en catégories, qui dépendent de la





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# **Population-based context**

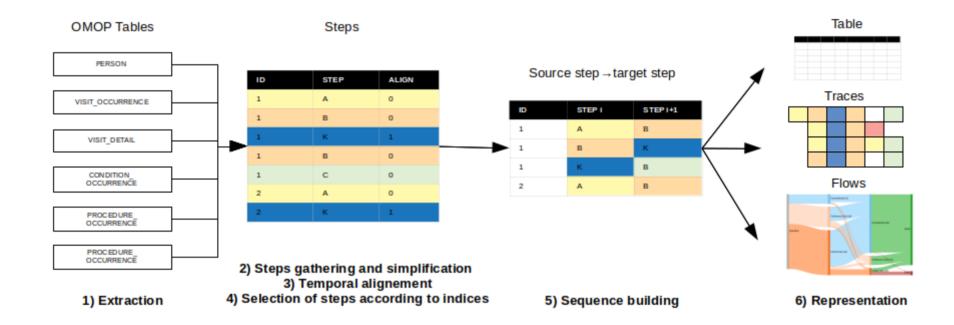
### Objective

Build patient pathway in preserving essential information and in a intelligible way:

- . Extract and gather relevent data
- Address the complexity of pathway (heterogeneity, temporality, ...)
- Evaluate the final information
- . Identify clusters
- . Compare clusters

# Implementation

# Implementation



- 2) Steps gathering and simplification
- 3) Temporal alignement
- 4) Selection of steps according to indices

Three main transformations :

- . Filter
- . Simplification
- Time-oriented transformations

#### Filter Events selection :

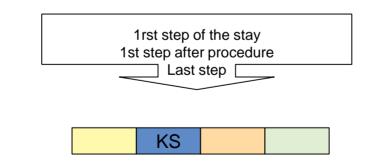
- identification and extraction of useful information in the source database. take into account the hierarchy, and extracting event based on the hierarcal level (e.g. a class of drugs)
  <u>Records filtering :</u>
  - at the individual level or population level
- at the start or at the end of the process

Both events and records may be filtereg based on the presence or absence of an event (e.g. administration of drug A), but also based on the value of an attribute of the event (e.g. administration of drug A with posology over 50 µg, the posology being the attribute)

#### Filter

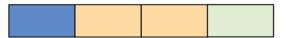
Events selection : identification and extraction of useful information in the source database





#### Simplication

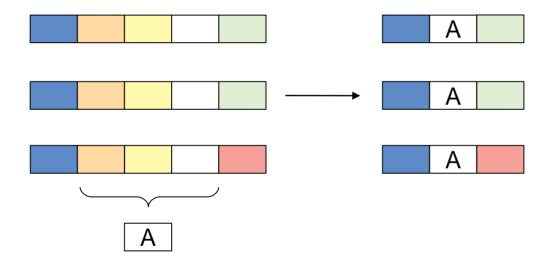
<u>Consecutive event merging</u>: two consecutive events represent the same state (e.g. a diagnosis) and are merged into a single event to reduce information overload, and the complexity and length of the trajectories.





Simplication

Frequency consecutive merging : Merging / grouping in a super-event based on the frequency of consecutive events



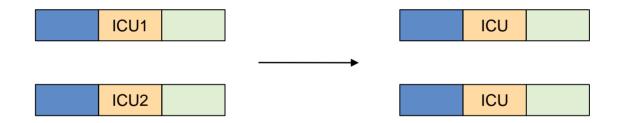
Perer A, Wang F, Hu J. Mining and exploring care pathways from electronic medical records with visual analytics. J Biomed Inform. 2015 Aug;56:369-78.

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#### Simplication

Label events merging : multiple events or categories to be combined into a single meta-event or meta-category

With a business rule



ICU : Intensive Care Unit

#### Simplication

Label events merging : multiple events, categories or clusters to be combined into a single meta-event, meta-category or meta-cluster

With a similarity distance between two nodes (Here, on the outcome). Minimum distance being defined by the user

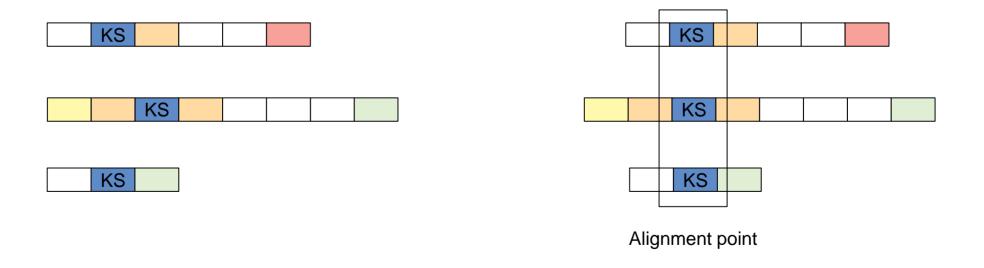
$$d(node_A, node_B) = |node_A.outcome - node_B.outcome$$

$$d(cluster_X, cluster_Y) = \frac{\sum_{m \in cluster_X \& n \in cluster_Y} d(m, n)}{size(cluster_X) * size(cluster_Y)}$$

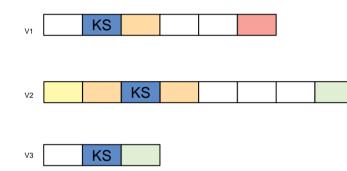
Wongsuphasawat and D. Gotz, "Exploring Flow, Factors, and Outcomes of Temporal Event Sequences with the Outflow Visualization," in IEEE Transactions on Visualization and Computer Graphics, vol. 18, no. 12, pp. 2659-2668, Dec. 2012, doi: 10.1109/TVCG.2012.225.

#### **Time-oriented transformations**

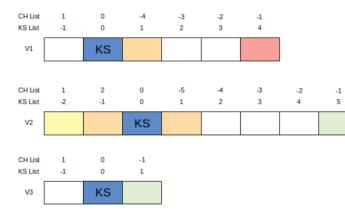
<u>Alignment according to event of interest :</u> the surgery day, a diagnostic, the start of a treatment, each day of the pathway the admission day



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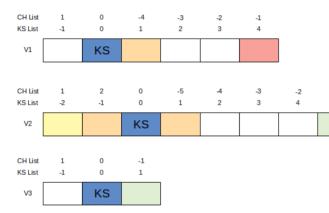


#### Raw data



CH List : chronological order KS List : key step order

#### Raw data

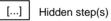


Step
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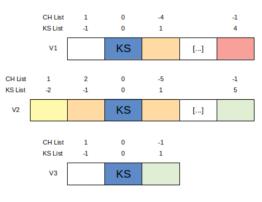
-1

5



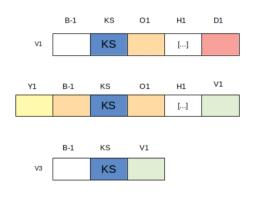


#### Individual traces

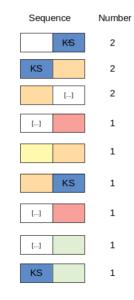


CH List : chronological order KS List : key step order

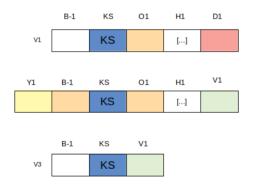
#### Indivicual traces



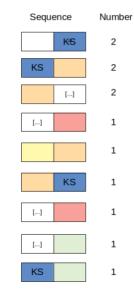
#### Populational flows



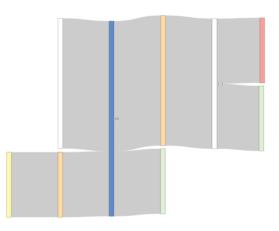
#### Indivicual traces



#### Population flows



#### Population aggregation



# Next steps

Integrating :

- Statistical modeling
- Cluster identification
- Clusters / steps merging (based on outcome?)
- Prediction of the next steps / outcome
- Awareness of temporal dimension in model

Wongsuphasawat K, Plaisant C, Taieb-Maimon M, Shneiderman B. Querying Event Sequences by Exact Match or Similarity Search: Design and Empirical Evaluation. *Interact Comput.* 2012;24(2):55-68. doi:10.1016/j.intcom.2012.01.003 K. Wongsuphasawat and B. Shneiderman, "Finding comparable temporal categorical records: A similarity measure with an interactive visualization," 2009 IEEE Symposium on Visual Analytics Science and Technology, 2009, pp. 27-34, doi: 10.1109/VAST.2009.5332595.

### Discussion and perspective

### **Discussion and perspective**

- Visualize the pathway is an help in generating hypotheses
- Need to select relevant steps and to focus on a segment of the pathway
- Current limit with reccurent and concurrent steps, and in the consideration of time