

Querying Temporal Anomalies in Healthcare Information Systems and Beyond

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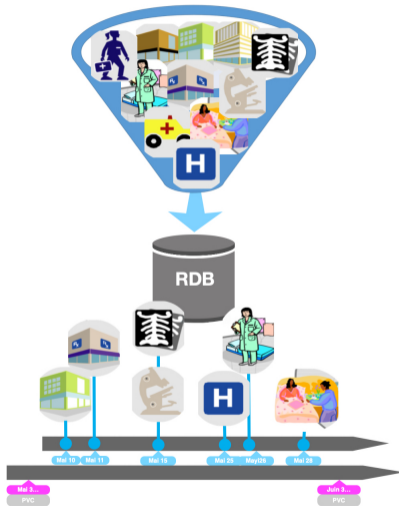
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Outline

- 1 Introduction
- 2 Temporal Anomalies
- 3 SQL Implementations
- 4 Experimental evaluation
- 5 Summary and Future work

Motivation & goal



- Temporal data integration is challenging task
- Anomalies could occur when feeding healthcare data system
- Detection of anomalies is crucial for consistency preserving and repairing

Contributions

- Defining the temporal anomalies:
 - temporal redundancy,
 - temporal contradiction,
 - temporal incompleteness.

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- Implementing the solutions using temporal aggregation.

Contributions

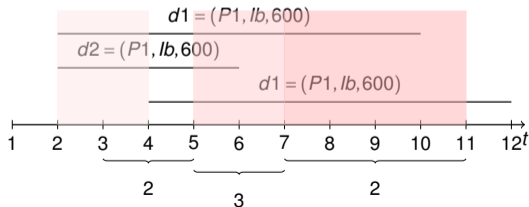
- Defining the temporal anomalies:
 - temporal redundancy,
 - temporal contradiction,
 - temporal incompleteness.
- Presenting two solutions for anomaly detection: **anomaly labelling** and **anomaly retrieval**.
- Implementing the solutions using temporal aggregation.
- Performing experiments on real-world medical data.

Temporal Redundancy

Same information stored at the same time point.

Prescription

	Pat	Name	Dose	T
d_1	P1	Ibuprofen	600mg	[2,10)
d_2	P1	Ibuprofen	600mg	[2,6)
d_3	P1	Ibuprofen	600mg	[4,12)
d_4	P2	Clopidogrel	75mg	[1,12)
d_5	P2	Clopidogrel	85mg	[10,14)
d_6	P2	Clopidogrel	95mg	[8,11)

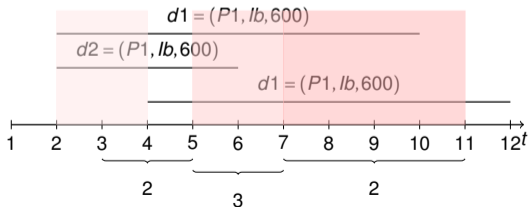


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d_5	P2	Clopidogrel	85mg	[10,14)
d_6	P2	Clopidogrel	95mg	[8,11)



Two solutions for temporal redundancy detection:

Pat	Name	Dose	T	Count
P1	Ibuprofen	600mg	[2,4)	2
P1	Ibuprofen	600mg	[4,6)	3
P1	Ibuprofen	600mg	[6,10)	2
P1	Ibuprofen	600mg	[10,12)	1
P2	Clopidogrel	75mg	[1,12)	1
P2	Clopidogrel	85mg	[10,14)	1
P2	Clopidogrel	95mg	[8,11)	1

(a) Redundancy labelling

Pat	Name	Dose	T	Count
P1	Ibuprofen	600mg	[2,4)	2
P1	Ibuprofen	600mg	[4,6)	3
P1	Ibuprofen	600mg	[6,10)	2

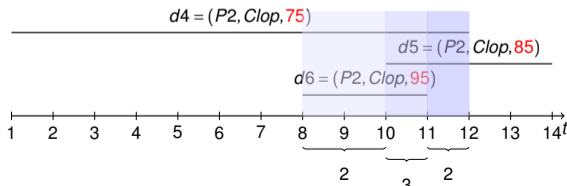
(b) Redundancy retrieval

Temporal Contradiction

Contradicting information stored at the same time point.

Prescription

	Pat	Name	Dose	T
d_1	P1	Ibuprofen	600mg	[2,10)
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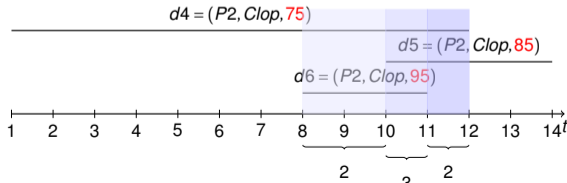


Temporal Contradiction

Contradicting information stored at the same time point.

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Two solutions for temporal contradiction detection:

Pat	Name	T	Contradiction
P1	Ibuprofen	[2,10)	600mg
P1	Ibuprofen	[2,6)	600mg
P1	Ibuprofen	[4,12)	600mg
P2	Clopidogrel	[1,8)	75mg
P2	Clopidogrel	[8,10)	75mg, 95mg
P2	Clopidogrel	[10,11)	75mg, 85mg, 95mg
P2	Clopidogrel	[11,12)	75mg, 85mg
P2	Clopidogrel	[12,14)	85mg

(a) Contradiction labelling

Pat	Name	T	Contradiction
P2	Clopidogrel	[8,10)	75mg, 95mg
P2	Clopidogrel	[10,11)	75mg, 85mg, 95mg
P2	Clopidogrel	[11,12)	75mg, 85mg

(b) Contradiction retrieval

Temporal Incompleteness

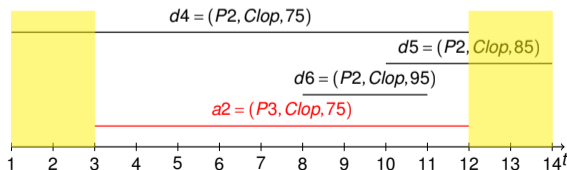
Missing information at the same point of time in the other relation.

Prescription

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Administration

	Pat	Name	Dose	T
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Temporal Incompleteness

Missing information at the same point of time in the other relation.

Two solutions for temporal incompleteness detection:

Prescription

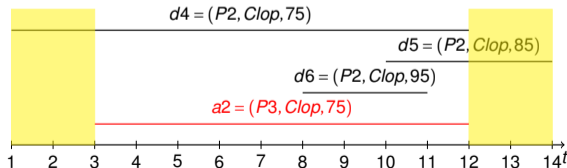
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Administration

	Pat	Name	Dose	T
a_1	P1	Ibuprofen	600mg	[2,12)
a_2	P2	Clopidogrel	75mg	[3,12)

Pat	Name	T	Src
P1	Ibuprofen	[2,12)	presc, admin
P2	Clopidogrel	[1,3)	presc
P2	Clopidogrel	[3,12)	presc, admin
P2	Clopidogrel	[12,14)	presc

(a) Incompleteness labelling



Pat	Name	T	Src
P2	Clopidogrel	[1,3)	presc
P2	Clopidogrel	[12,14)	presc

(b) Incompleteness retrieval

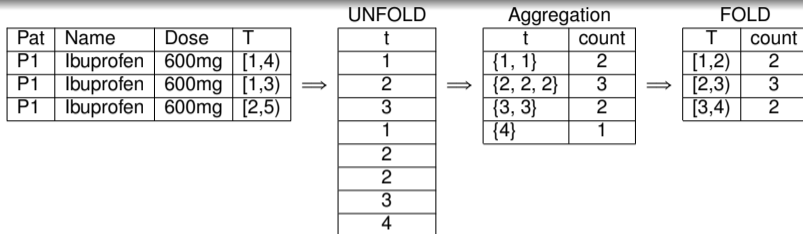
SQL Implementations

- Providing three different implementations for the retrieval and labelling solutions of the three anomalies:
 - Unfold/Fold
 - Unfold/Fold Join Filtered
 - Window Function
- Presenting redundancy retrieval using the three implementations.

SQL Implementations

Unfold/Fold

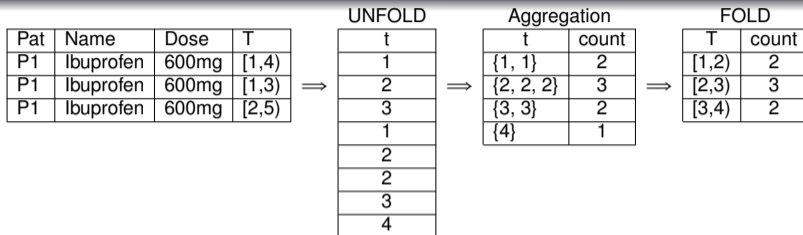
(Lorentzos and Mitsopoulos
1997)



SQL Implementations

Unfold/Fold

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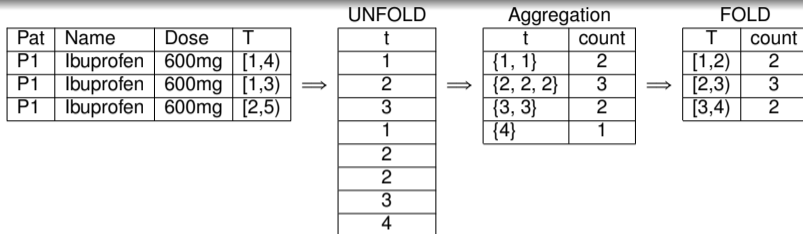
Unfold/Fold Join Filtered

- using a semi join to reduce the input for the unfold function.
- applied only for redundancy retrieval and contradiction retrieval.

SQL Implementations

Unfold/Fold

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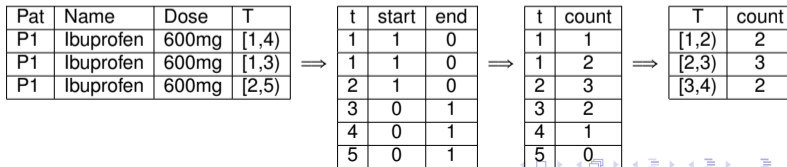


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Window Function

(Dignös et al. 2019)

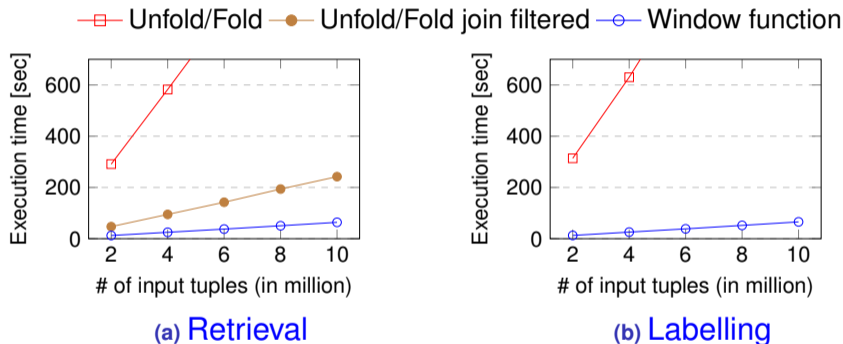


Experimental evaluation

- **Dataset:** Real-world MIMIC-IV (Medical Information Mart for Intensive Care) dataset (Johnson et al. 2020)
 - Prescriptions: 13M
 - Pharmacy: 10M
- **Database Server:** PostgreSQL version 14
- **Measure:** Comparison of runtime for anomaly retrieval and labeling with the three implementations

Experimental evaluation

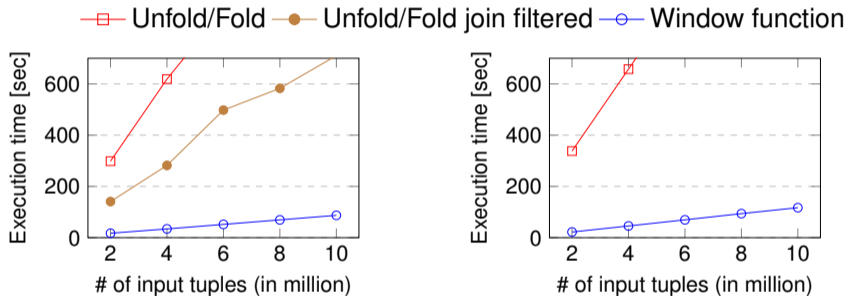
Experiment 1: Runtime for temporal redundancy on the Prescription table.



- Unfold/fold is very slow because unfold operation.
- Join filtering yields a substantial improvement in the performance.
- Window functions is the most efficient for retrieval and labelling.

Experimental evaluation

Experiment 2: Runtime for temporal contradiction on the Prescription table.



(a) Retrieval

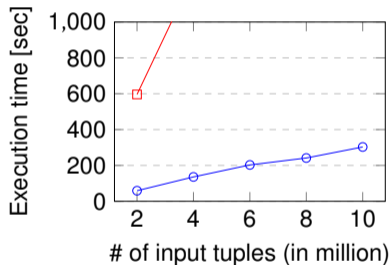
(b) Labeling

- The join filtering implementation is less effective compared to redundancy.
- Window functions is the most efficient.

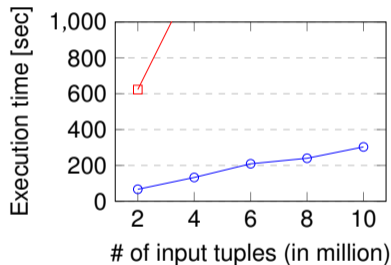
Experimental evaluation

Experiment 3: Runtime for temporal incompleteness retrieval/labeling.

—□— Unfold/Fold —○— Window function



(a) Retrieval



(b) Labelling

- Window functions is the most efficient.

Summary and Future work

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- Defining three temporal anomalies: temporal redundancy, temporal contradiction and temporal incompleteness.
- Presenting temporal queries to detect inconsistencies.
- Providing efficient implementations.

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- Defining three temporal anomalies: temporal redundancy, temporal contradiction and temporal incompleteness.
- Presenting temporal queries to detect inconsistencies.
- Providing efficient implementations.

Future work

- Extending the temporal anomalies operations to cover bitemporal models, cross-relational temporal contradictions, and semantic temporal conflicts.
- Developing a federated version for privacy-preserving anomaly mining.
- Defining a temporal repair system using healthcare expert preferences.

Thank you!