

EPISODESUPPORT: A GLOBAL CONSTRAINT FOR MINING FREQUENT PATTERNS IN A LONG SEQUENCE OF EVENTS

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CONTEXT

- ▶ This talk is about finding **frequent episode patterns** in a **long (time-stamped) sequence**
 - Very efficient dedicated algorithms exist (Minepi, Winepi, Emma,...)
 - They are not **flexible** and suffer for **memory problem**
- ▶ Motivation for CP:
 - finding **frequent (constrained) sub-sequences** is a related problem to **frequent (constrained) episode patterns** in a long (time-stamped) sequence
 - constraint example: satisfying a regular constraint
 - CP-based method is the state-of-the-art for finding **frequent (constrained) sub-sequences** in a sequence database [Aoga et al., ECMLPKDD'16; CPAIOR'17]

SPM PROBLEM

	Sub sequence		Sequence		
Client1	Milk	Coffee	Sugar	Coffee	Sugar
Client2	Coffee	Milk	Coffee	Sugar	
Client3	Milk	Coffee			
Client4	Coffee	Sugar	Egg		

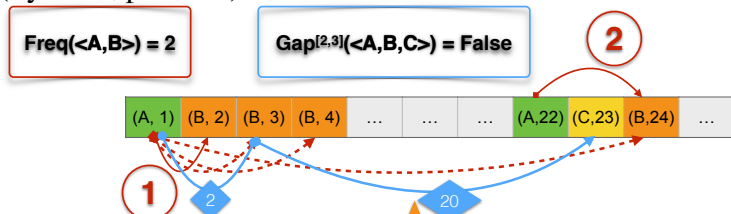
Sequence Database (SDB)

- Sequence : < Milk Coffee Sugar Coffee Sugar >
- Subsequence : <Coffee Sugar >
- Freq (<Coffee Sugar >) = 3

Problem : Find all subsequences with support ≥ Given Threshold

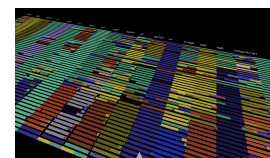
(CONSTRAINT-BASED) EPISODE MINING PROBLEM

(Symbol, position)



Problem : Find All Episodes wrt. user-defined constraints (e.g. support ≥ Given Threshold)

✓ Applications



DNA Sequence

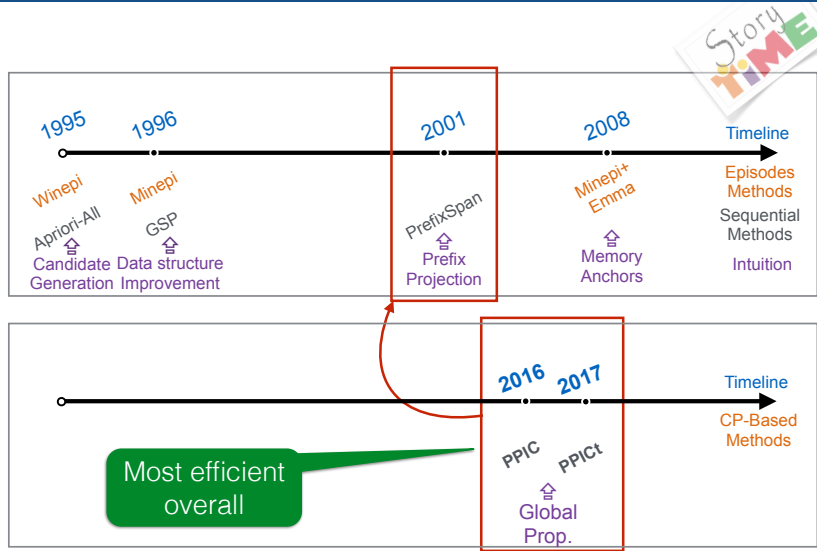


Smartphone lifelogging

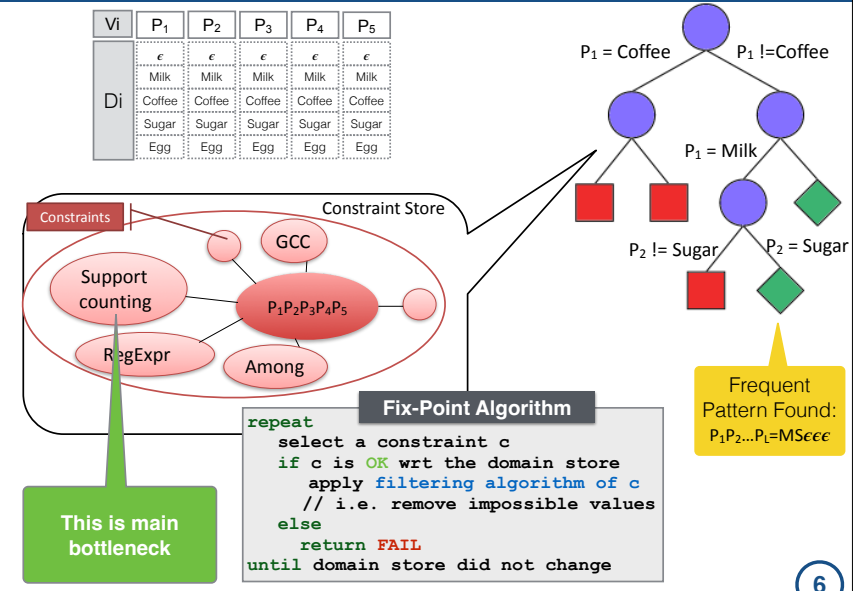


Stock market

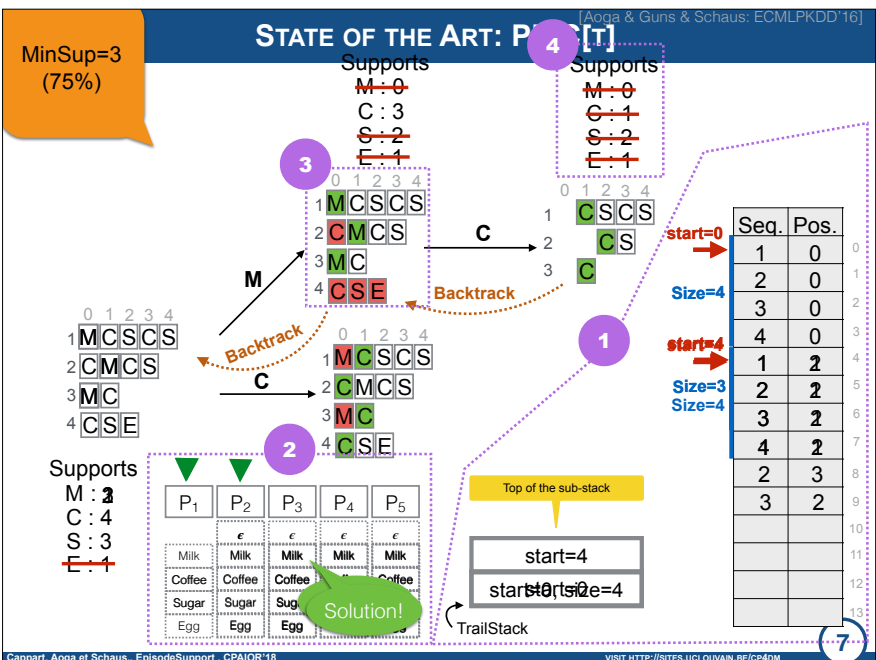
Related Work



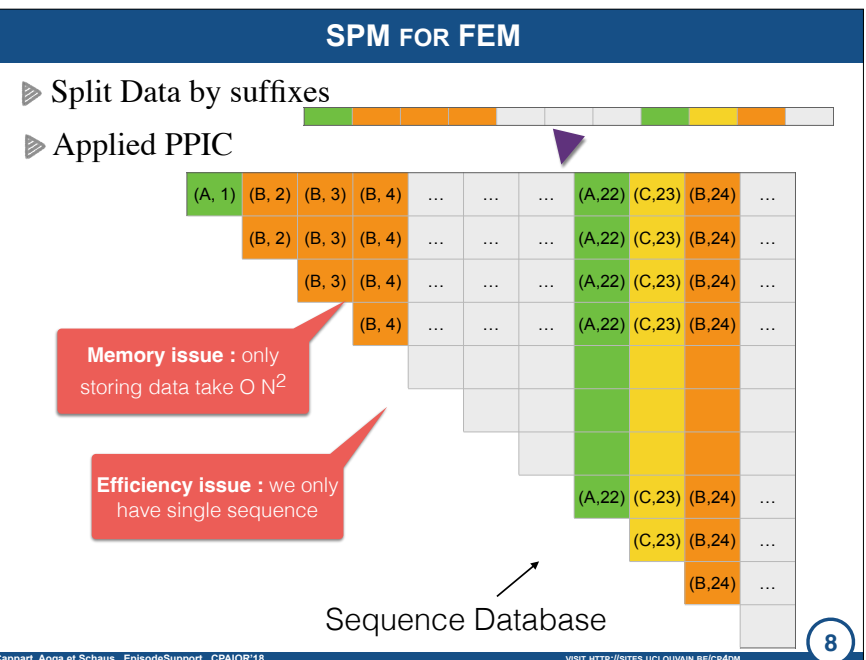
CP : Filtering + DFSearch



STATE OF THE ART: PPIC



SPM FOR FEM



EPISODESUPPORT : CONTRIBUTION

CONTRIBUTION

Goal: Design new Approach for finding Episodes capturing the most common constraints (including syntax and time-related constraints)

- ✓ Adapt trailed-based data structure to efficiently overcome memory issue
- ✓ Take into account that we have a single sequence with algorithmic improvements
- ✓ Tackle time series data and time-related constraints
- ✓ Show real application handling many other constraints: Regular/Grammar, Gcc, Among, ...



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MinSup=2

PPIC

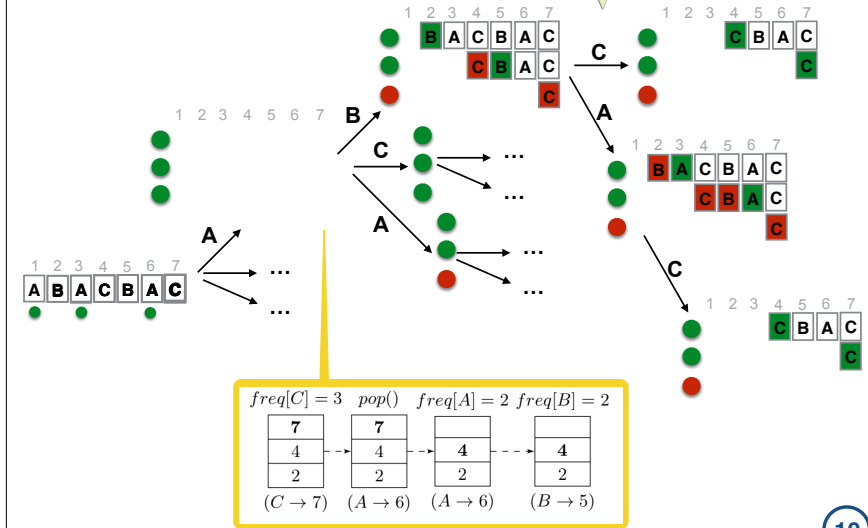
1 data-structure

2 pruning

3 projection

4 support counting

CONTRIBUTION



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EXPERIMENTS

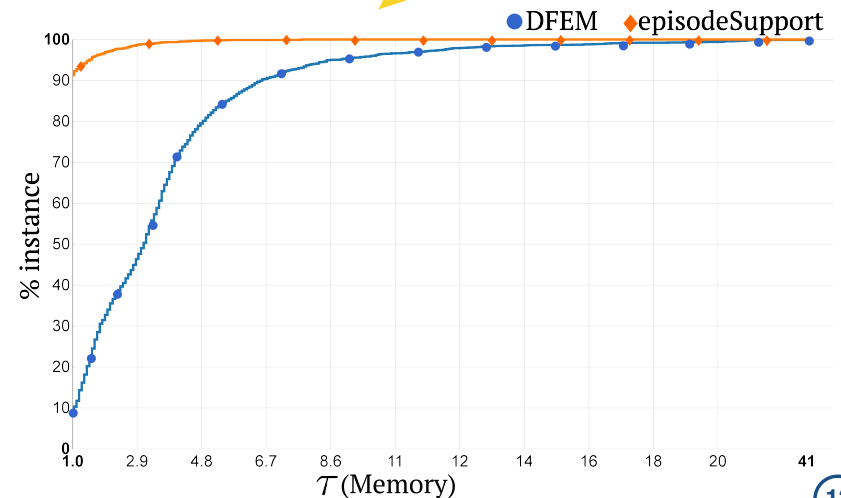


- ▶ EpisodeSupport (2versions — with/without time consideration)
- <https://bitbucket.org/projetsJOHN/episodesupport> (also available in Oscar)

DFEM VS EPISODESUPPORT (MEMORY)

Time limit = 600s (10Minutes)

Uniprot - Human proteins dataset
(2452 instances) - $\theta=5\%$; MaxSize=5

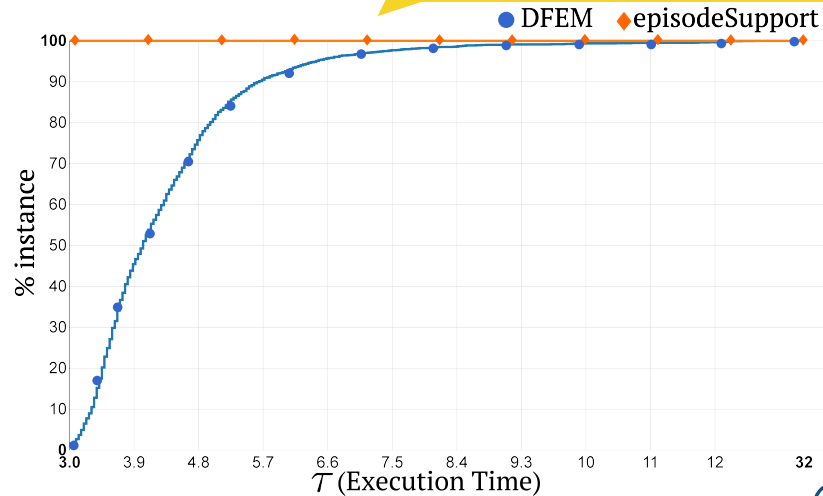


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DFEM VS EPISODESUPPORT (EXECUTION TIME)

Time limit = 600s (10Minutes)

Uniprot - Human proteins dataset
(2452 instances) - $\theta=5\%$; MaxSize=5

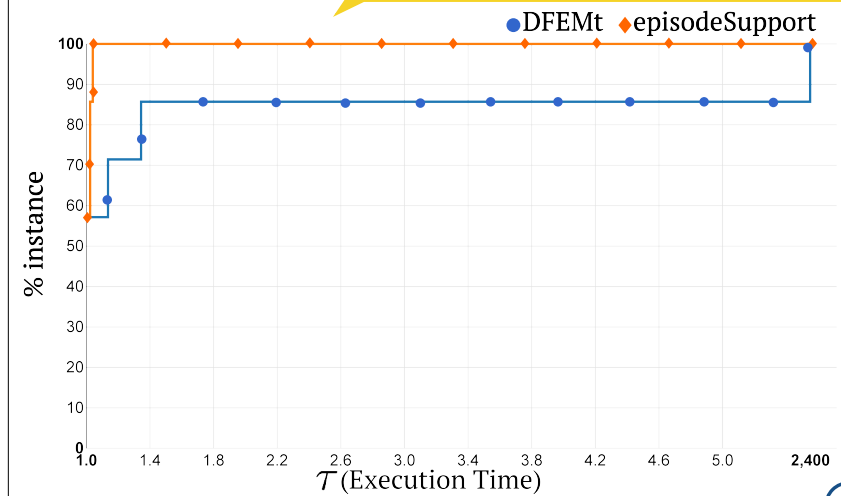


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DFEMT VS EPISODESUPPORT (EXECUTION TIME)

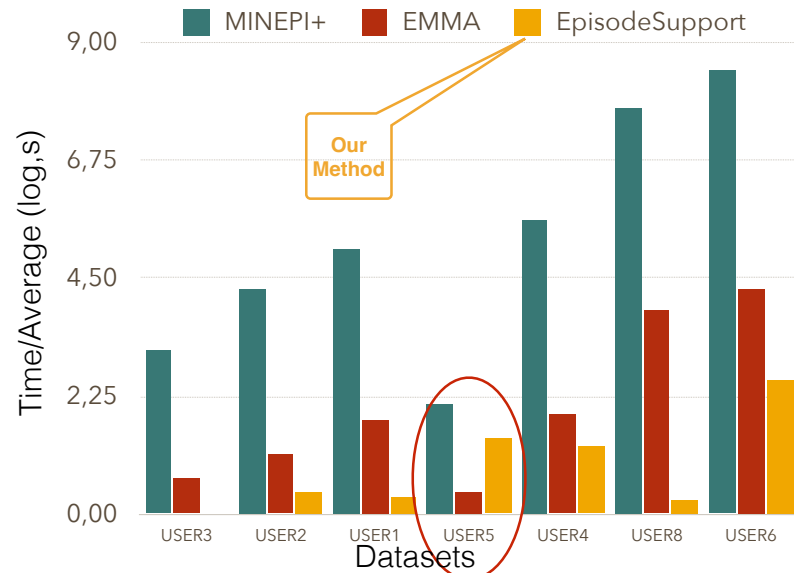
Time limit = 600s (10Minutes)

smartphone lifelogging dataset
(21 instances) - $\theta=5\%$; MaxSize=5 gap[1s, 1hour]



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EPISODESUPPORT VS MINEPI+ AND EMMA



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Take-Away message

- Two versions of Global constraints (with/without time consideration)
- Efficiently split long sequence into small sequences for efficient memory usage
- Many kind of existing modules (in CP-Solvers) are reusable for free
- **Efficient memory using Trail-based backtracking aware data structure adaptation**

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