berlin ਸੀਪੁਸ MANNHEIM

Scalable Frequent Sequence Mining With Flexible Subsequence Constraints

Goal: Useful, Useable, and Scalable FSM

- **Useful:** Develop general FSM algorithms that can be tailored to diverse applications, such as natural language processing, information extraction, web usage mining, market-basket analysis, and computational biology
- **Useable:** Flexible subsequence constraints allow applications to specify patterns of interest intuitively
- Scalable: Ability to deal with large input, search space, and output

DESQ [1, 2]: Usefulness and Usability

A unified FSM framework to specify flexible subsequence constraints in an intuitive, declarative way (1) Hierarchies

- Allow for discovering non-trivial patterns
- Example: $DSLR \rightarrow Tripod \rightarrow Flash$
- (2) Pattern Expressions
- Specify subsequence constraints
- Example: $(Book)[.\{0,2\}(Book)]\{1,4\}$
- (3) FSTs
- Computational framework: translate input sequence to candidate subsequences
- Example:



Example

- Sequence database: $T_1: a_1cdcb$ T_2 : eea_1ea_1eb T_3 : cdcb T_4 : a_2db $T_5: a_1 a_1 b$
- Item hierarchy: c d e
- Pattern expression: $(A)[(.^{\uparrow}).^{*}]^{*}(b).^{*}$ Minimum support: $\sigma = 2$ Output: a_1b (3), a_1a_1b (2), a_1Ab (2)

2a.

Alexander Renz-Wieland¹, Matthias Bertsch², Rainer Gemulla²

¹Technische Universität Berlin, ²Universität Mannheim

Our Contribution: Scalability

1. A general framework for **distributed FSM** with flexible subsequence constraints

2. Two algorithms within this framework:

(a) D-CAND: Communicate compressed candidate subsequences

(b) D-SEQ: Communicate rewritten input sequences

3. Large-scale experimental study

1. General Framework

• We generalize existing item-based partitioning approaches (MG-FSM, LASH) to a general framework that supports flexible subsequence constraints

• Key questions: how to distribute computation and what to communicate



Communicate Candidates

• Communicate candidate subsequences as compressed nondeterministic finite automata (NFA)

• Beneficial for selective constraints

• Example:



2b. Communicate Inputs

- Example:

3. Experimental Study naïve approaches by up to 50x Naïve SemiNaïve D-SEQ $N_1(10)$ $N_2(100) = N_3(10) = N_4(1k)$ $N_5(1k)$ Subsequence constraint (a) New York Times data

• **Traditional Patterns:** Both algorithms exhibited acceptable generalization overhead over existing, specialized methods

References and Resources

- [1] K. Beedkar and R. Gemulla.
- [2] K. Beedkar, R. Gemulla, and W. Martens.

Code is **open source** and available at https://github.com/rgemulla/desq/tree/distributed.

• Communicate task-relevant items of input sequences • Robust across a wide range of mining tasks

 $ee \ a_1 ea_1 eb \ ee ee$

• Flexible Patterns: Both algorithms outperformed





DESQ: Frequent sequence mining with subsequence constraints. IEEE International Conference on Data Mining '16.

A unified framework for frequent sequence mining with subsequence constraints. To appear in ACM Transactions on Database Systems '19.