



Universiteit Utrecht

[Faculty of Science  
Information and Computing Sciences]

# Characteristic Relational Patterns

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# Characterising the Database

- Relational database models
  - Local models: frequent pattern mining
  - Global models: probabilistic relational model
- Characterising the database
  - Combine patterns to form a global model
- Experiments

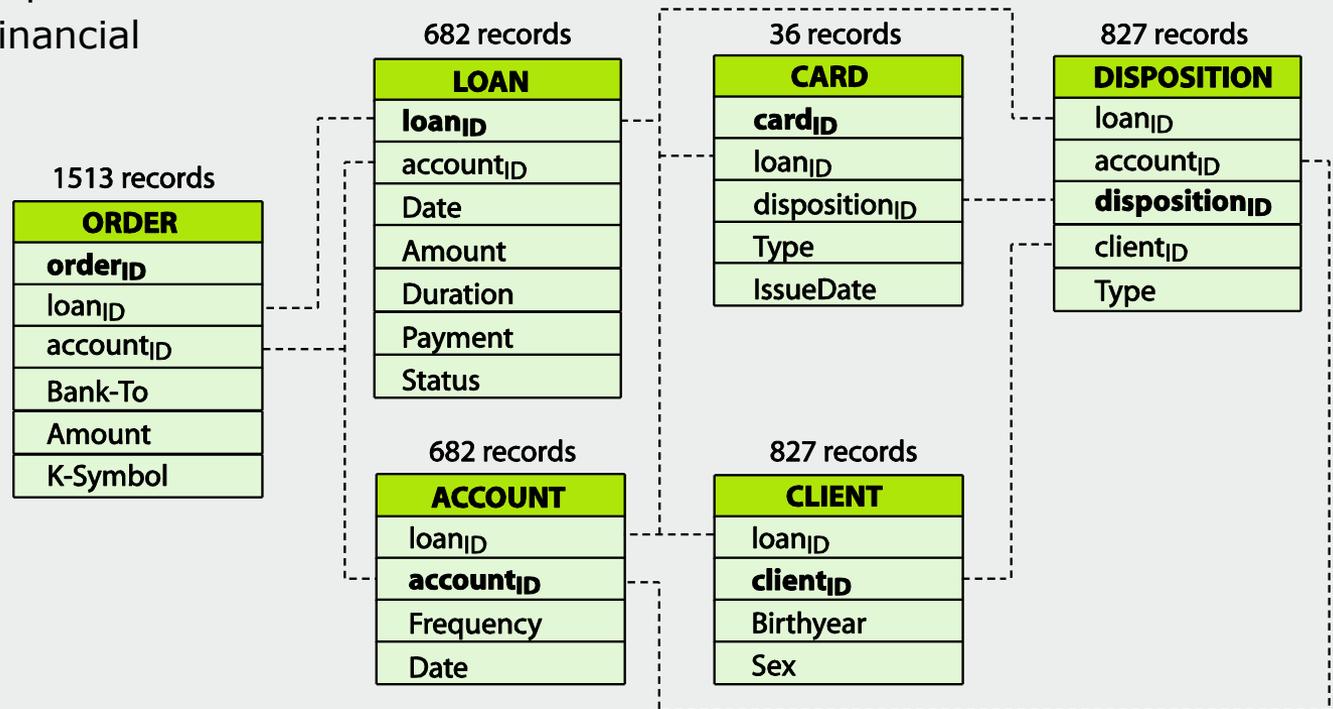


# Relational Databases

## ■ KDD cup relational databases

- Genes
- Hepatitis
- Financial

## Financial database



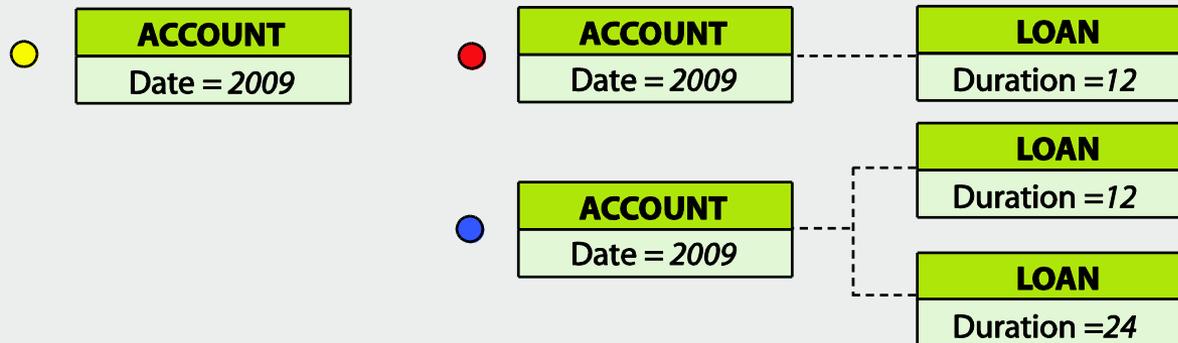
# Local Models

ACCOUNT	
accountID	Date
10	2009
11	2009

LOAN		
loanID	accountID	Duration
100	10	12
101	11	12
102	11	24

## Patterns



■ Frequent pattern mining: too many patterns!

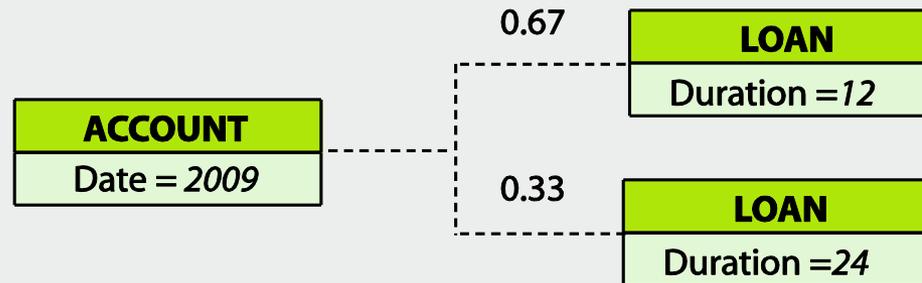


# Global Model

ACCOUNT	
accountID	Date
10	2009
11	2009

LOAN		
loanID	accountID	Duration
100	10	12
101	11	12
102	11	24

## Probabilistic Relational Model



- Probabilistic Relational Models: local co-occurrence information lost

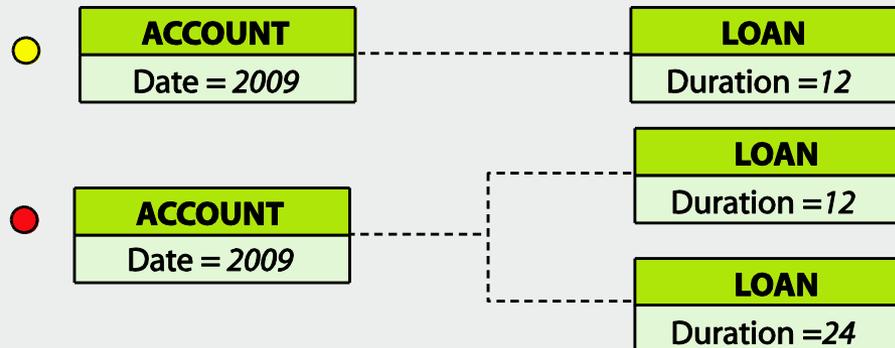


# Combined Model

ACCOUNT	
accountID	Date
10	2009
11	2009

LOAN		
loanID	accountID	Duration
100	10	12
101	11	12
102	11	24

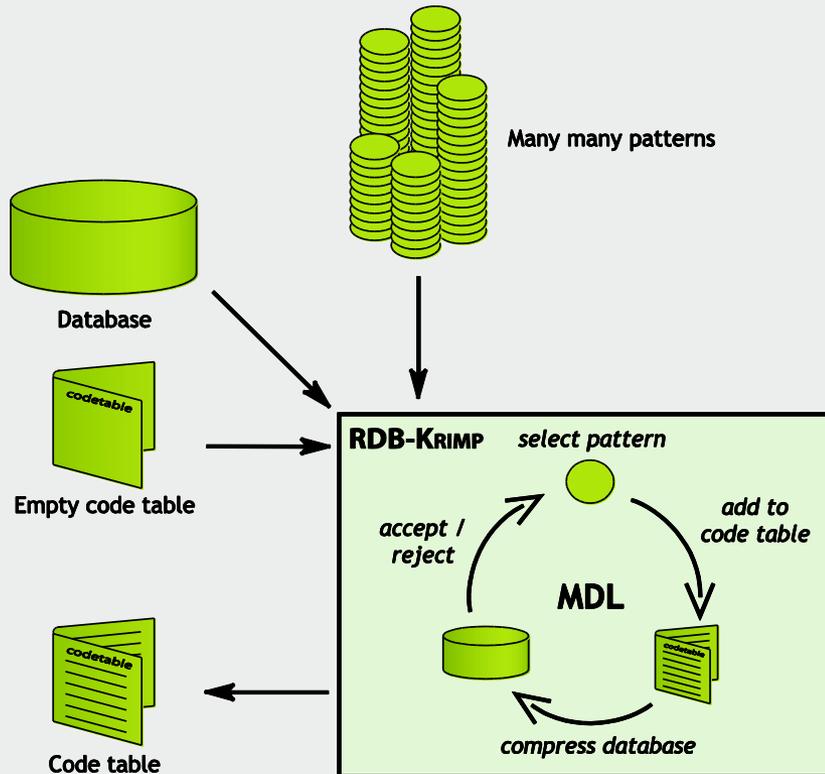
## Code Table



■ Relational Code Table: compact and lossless description of the complete database



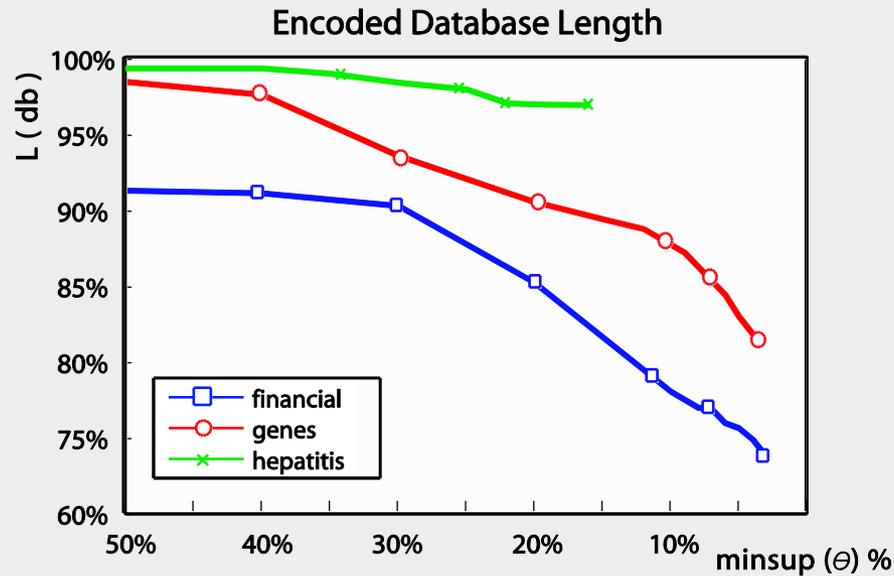
# RDB-KRIMP



- RDB-KRIMP selects patterns that describe the database well
- Candidates are frequent relational patterns
- Describing patterns are placed in a code table
  - shortness code length proportional to usage



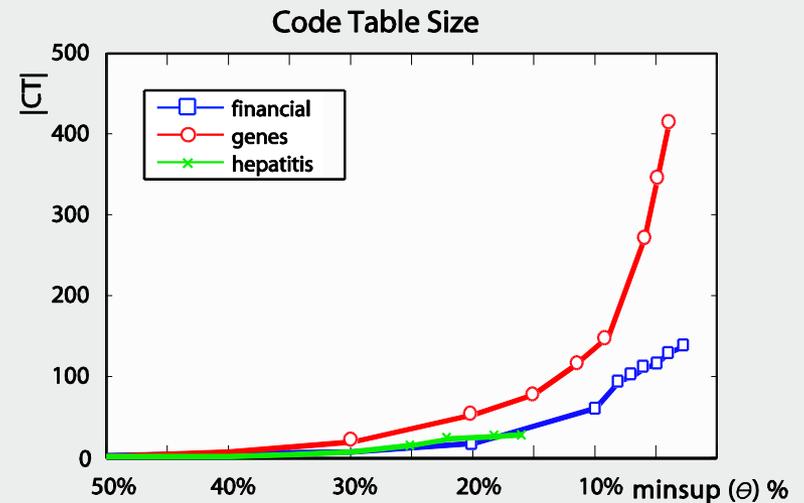
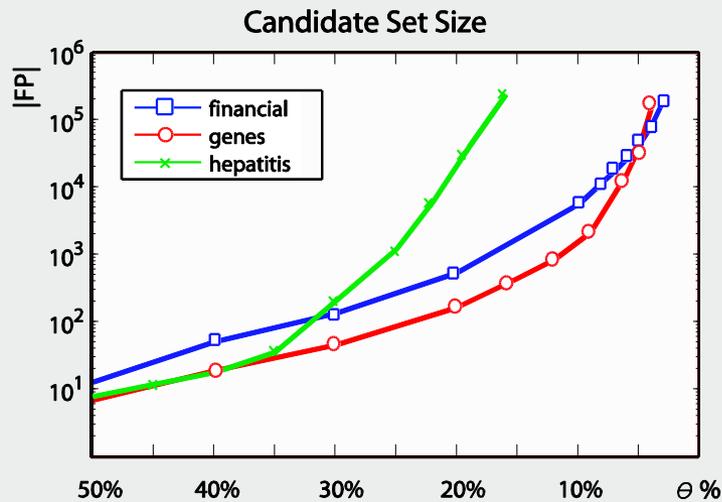
# Compression - 1



- Code tables encode the database
  - Increasingly better encoded sizes for lower minimum supports



# Compression - 2



- Code tables encode the database
  - Increasingly better encoded sizes for lower minimum supports
  - Candidate set grow exponentially
  - Code tables stay compact



# Pattern Languages

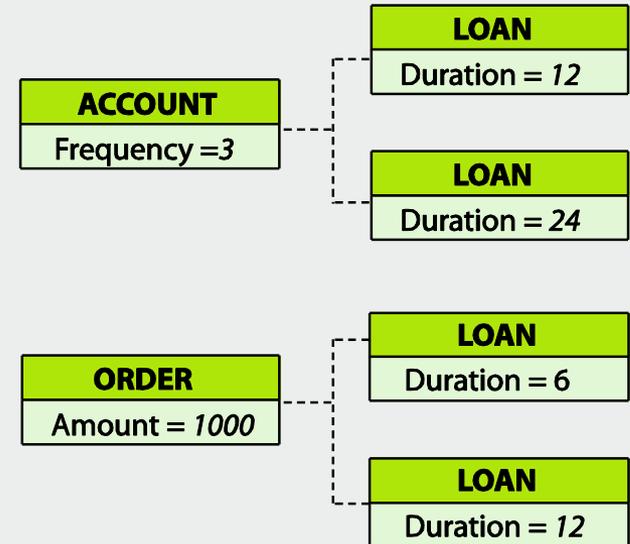
## local table



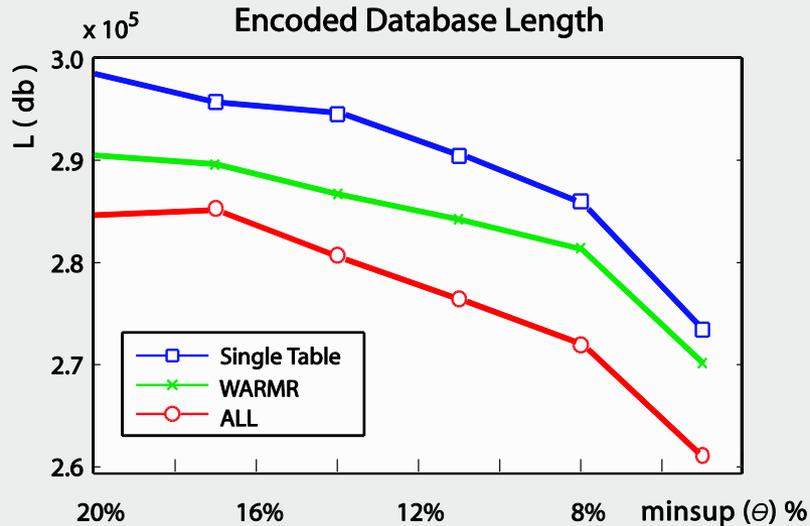
## WARMR



## ALL: FARMER w/o target



# Pattern Complexity



- More complex patterns lead to better descriptions.
- Thus, they encode MDL-relevant structure

Database	single table		WARMR		all	
	L	CT	L	CT	L	CT
Financial	91%	29	76%	130	76%	117
Genes	87%	72	86%	191	83%	342
Hepatitis	99%	5	98%	13	97%	26



# Conclusions

- Code tables describe the database while preserving local information
- Code tables stay compact
  - Stay compact for low minimum support values
  - Reductions up to 4 orders of magnitude
- Richer patterns lead to better models
  - Smaller encodings
  - Better descriptions without target tables



# Questions?



# Database Encoding

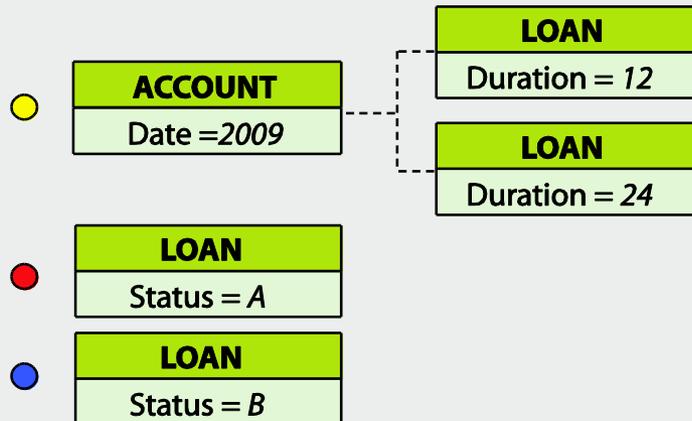
lossy encoded database

ACCOUNT	
accountID	Date
10	2009
11	2009

LOAN			
loanID	accountID	Duration	Status
30	10	12	A
31	10	24	A
35	11	24	A
36	11	12	B

REORDERDB

Code Table

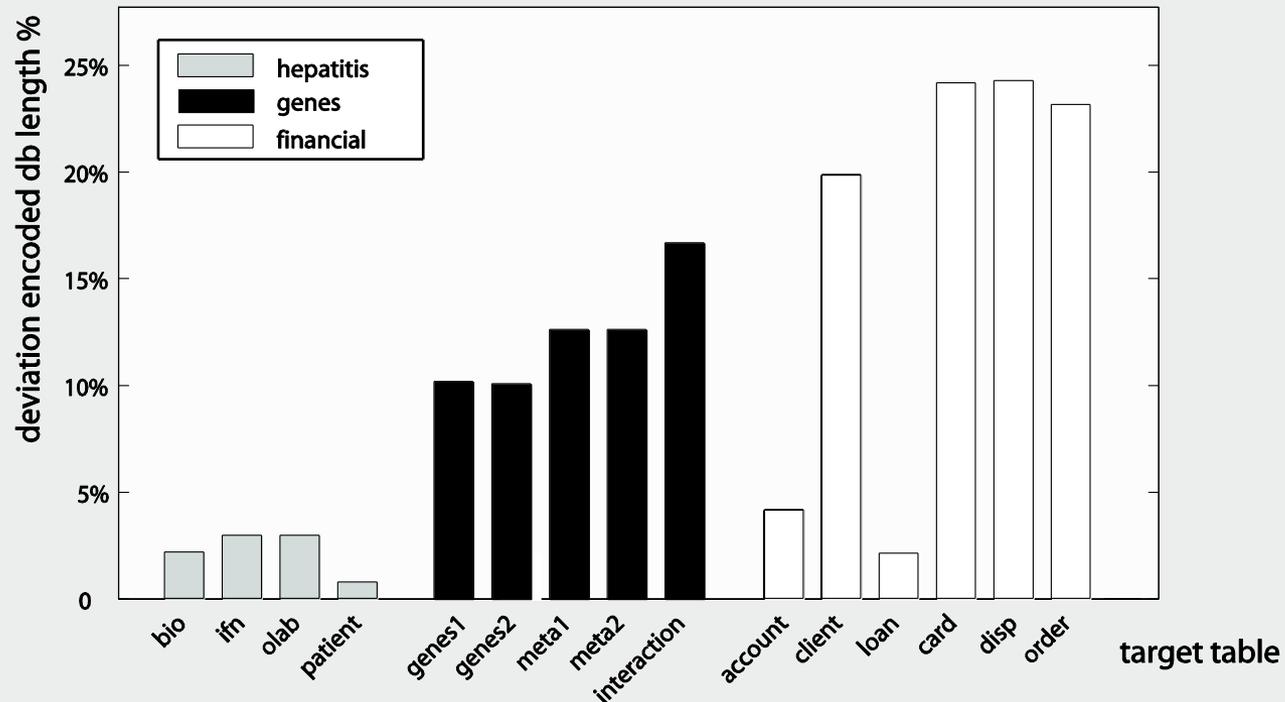


■ Reordering allows for a lossless encoding



# Target Tables

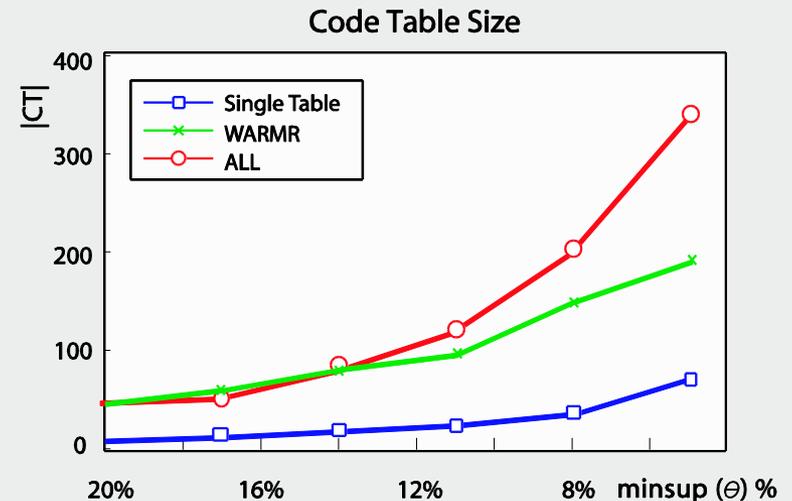
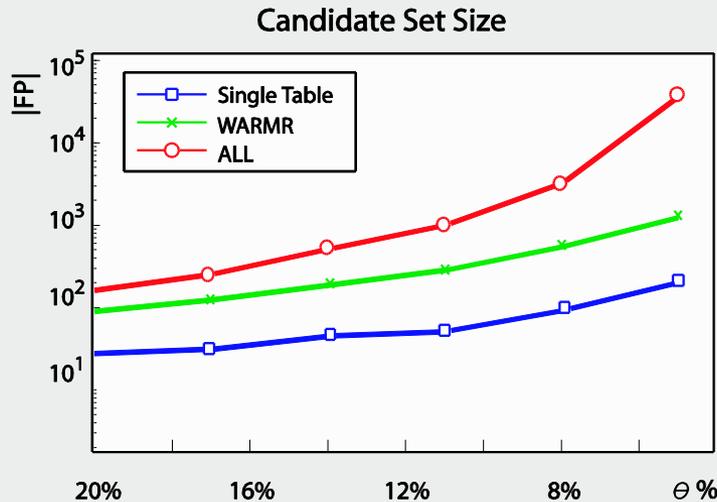
Encoded database length induced by target tables



■ We obtain better encodings without a target table



# Pattern Complexity



- In all cases reductions are obtained
- Additional rich patterns lead better encodings

