

On-demand Relational Concept Analysis

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Context

Large multi-relational datasets



<http://www.edeation.fr/wp-content/uploads/2015/01/Graphe-Global-Géant.png> - <https://gisellezeno.com/category/academic-work.html> -

<https://www.frontiersin.org/articles/10.3389/fdigh.2017.00011/full> - <http://social-dynamics.org/tag/clustering-algorithm/> -

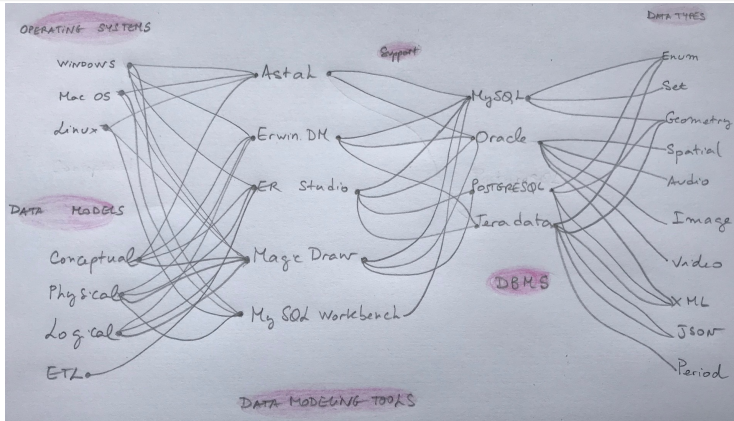
<https://www.bmj.com/content/337/bmj.a2338>

Data exploitation tasks

- Querying
- Extraction of knowledge patterns
- Classification
- Browsing, Exploring

Illustrative exploration example

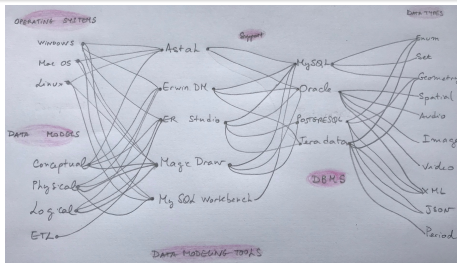
Select a Data Modeling Tool for conceptual model, running on Windows



⇒ Astah, Erwin DM, Magic Draw, ER/Studio

Illustrative exploration example

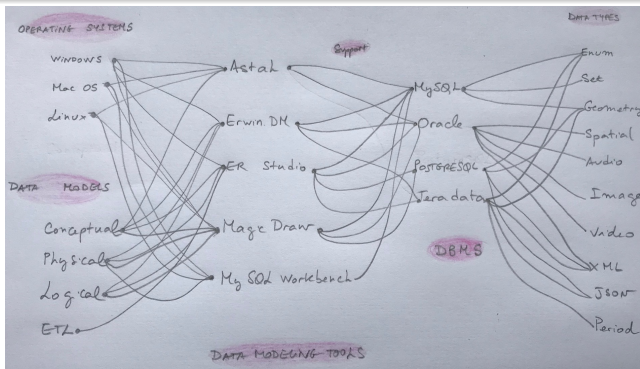
Are Astah, Erwin DM, Magic Draw, ER/Studio equivalent? What are their commonalities and variability in terms of properties and properties of their supported DBMS?



⇒ Astah, Magic Draw are available on Mac OS and Linux. Erwin DM, Magic Draw, ER/Studio allow physical and logical modeling (...) Magic Draw is the more complete (except it has no ETL modeling)

Illustrative exploration example

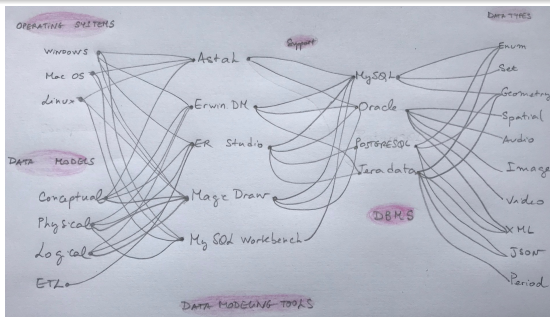
What if I relax the constraint "conceptual modeling"? This is not so important, having conceptual or physical modeling would be great.



⇒ All DM tools become relevant - Increases the choice number

Illustrative exploration example

All DM tools share MySQL. My boss is against. Which other similar DBMS could I find? With which additional/lost properties?



↪ Compared with MySQL: PostgreSQL has in common Enum and Geometry data types; PostgreSQL also has JSON; Teradata is even more complete, proposing Period data type; Oracle does not share anything.

Illustrative exploration example

Lessons learned

- Using tables, or following links between individual objects is tricky
- Receiving a flat set of answers does not help so much

~> Need for data structuring to foster exploration

- object aggregation by categories (groups)
 - *DBMS with Enum and Geometry*
- link aggregation (between-group links)
 - *DM tools that support one DBMS with Enum and Geometry*
- informed ranking, observed patterns
 - quantitative similarity may help (how much similar)
 - qualitative similarity is necessary (why/how similar, specialization)
 - *Teradata is more complete than PostgreSQL, proposing Period data type*
 - *Json → XML*

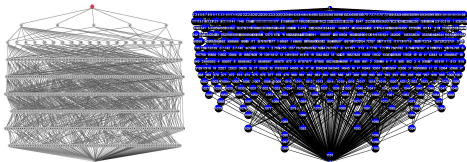
Solution tracks

Concepts in multi-relational datasets

- Power contexts families [Wille, 2002]
- Triadic [Lehmann and Wille, 1995], Polyadic Concept Analysis [Voutsadakis, 2002]
- Cubes of Concepts [Ferré et al., 2012]
- Relational, windowed structures [Kötters, 2013]
- Graph-FCA [Ferré, 2015]

→ Relational Concept Analysis [Hacene et al., 2013]

Concept lattices may be large (Villerd et al., Alam et al. (Latviz))



→ Adopt exploratory approaches, cf. Abstract Conceptual Navigation [Ferré, 2014]

Relational Concept Analysis [Hacene et al., 2013]

- Extends the purpose of FCA for taking into account **object categories** and **links** between objects
- Main principles:
 - a relational model based on an **entity-relationship model**
 - integrate relations between objects as **relational attributes**
 - a variety of **quantifiers** for creating the *relational* attributes: $\exists, \exists \forall, \exists \supseteq, \dots$
 - **iterative** process
- RCA provides a set of **interconnected lattices**
- Produced structures can be represented as **ontology concepts** within a knowledge representation formalism such as description logics (DLs)

Relational Context Family

<i>DM_tools</i>	OS:Windows	OS:Mac OS	OS:Linux	DM:Conceptual	DM:Physical	DM:Logical	DM:ETL
Astah	x	x	x	x			
Erwin DM	x			x	x	x	
ER/Studio	x			x	x	x	x
Magic Draw	x	x	x	x	x	x	
MySQL Workb.	x	x	x		x		

Formal context *DM_tools*

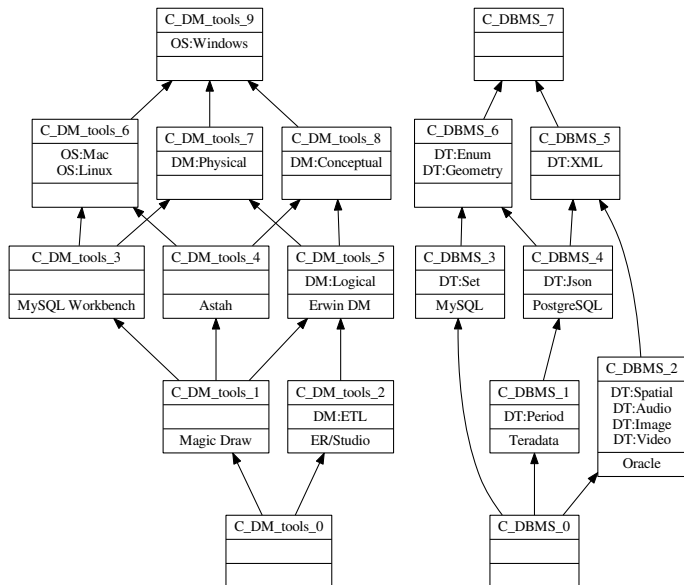
<i>DBMS</i>	DT:Enum	DT:Set	DT:Geometry	DT:Spatial	DT:Audio	DT:Image	DT:Video	DT:XML	DT:JSON	DT:Period
MySQL	x	x	x							
Oracle				x	x	x	x	x		
PostgreSQL	x		x					x	x	
Teradata	x		x					x	x	x

Formal context *DBMS*

<i>support</i>	MySQL	Oracle	PostgreSQL	Teradata
Astah	x	x		
Erwin DM	x	x		x
ER/Studio	x	x	x	x
Magic Draw	x	x	x	
MySQL Workb.	x			

Relational context *support*

Initial Lattices



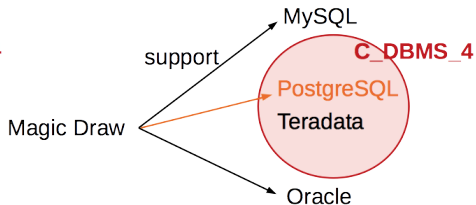
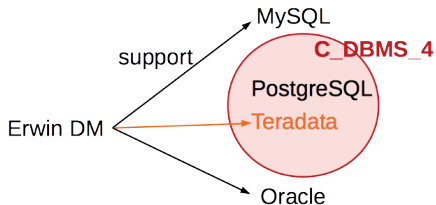
Enriching with *support*

<i>DM_tools</i>	OS:Windows	OS:Mac OS	OS:Linux	DM:Conceptual	DM:Physical	DM:Logical	DM:ETL	support:MySQL	support:Oracle	support:PostgreSQL	support:Teradata
Astah	x	x	x	x				x	x		
Erwin DM	x			x	x	x		x	x		x
ER/Studio	x			x	x	x	x	x	x	x	x
Magic Draw	x	x	x	x	x	x		x	x	x	
MySQL Workb.	x	x	x		x			x			

- Extracted knowledge: ErwinDM and Magic_Draw share MySQL and Oracle
- Knowledge not extracted:
 - Magic_Draw supports PostgreSQL; ErwinDM supports Teradata
 - PostgreSQL and Teradata admit DT:JSON
 - ErwinDM and Magic_Draw both support a DBMS admitting DT:Json

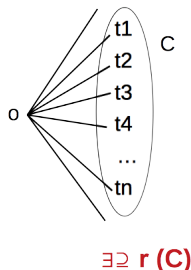
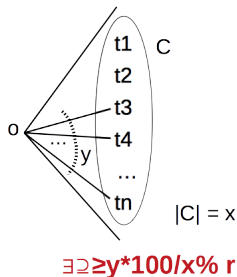
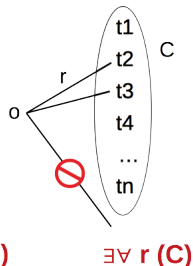
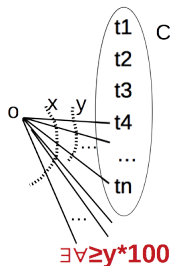
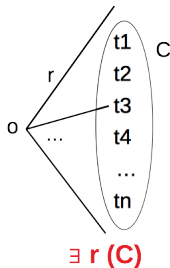
Consider quantifiers

Relational attribute $\exists r(C)$



$\exists \text{ support}(C_DBMS_4)$

Consider quantifiers

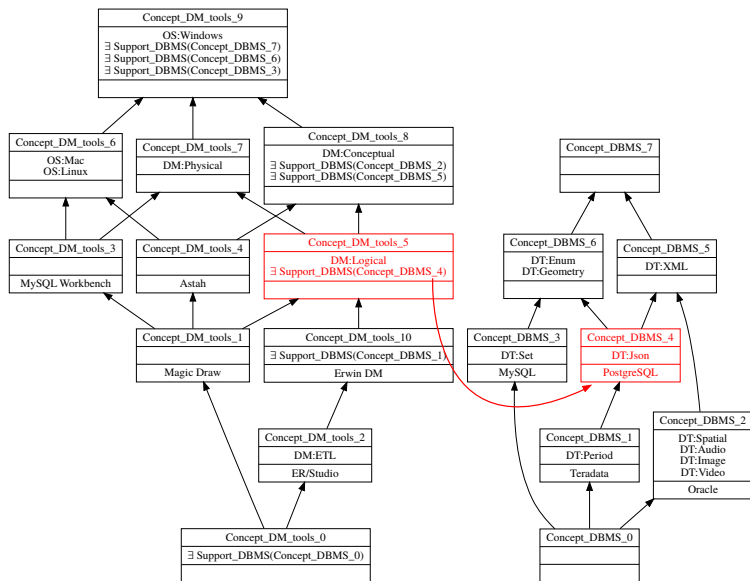


Enriching with \exists support

DM_tools	OS:Windows	OS:Mac	OS:Linux	DM:Conceptual	DM:Physical	DM:Logical	DM:ETL	\exists Support_DBMS(Concept_DBMS_7)	\exists Support_DBMS(Concept_DBMS_6)	\exists Support_DBMS(Concept_DBMS_3)	\exists Support_DBMS(Concept_DBMS_0)	\exists Support_DBMS(Concept_DBMS_2)	\exists Support_DBMS(Concept_DBMS_4)	\exists Support_DBMS(Concept_DBMS_5)	\exists Support_DBMS(Concept_DBMS_1)
Astah	×	×	×	×				×	×	×		×		×	
Erwin DM	×			×	×	×		×	×	×		×	×	×	×
ER/Studio	×			×	×	×	×	×	×	×		×	×	×	×
Magic Draw	×	×	×	×	×	×		×	×	×		×	×	×	
MySQL Workbench	×	×	×		×			×	×	×					

Now ErwinDM and Magic_Draw share "support a DBMS with DT:Json"
(Concept_DBMS_4)

Connected concept lattices

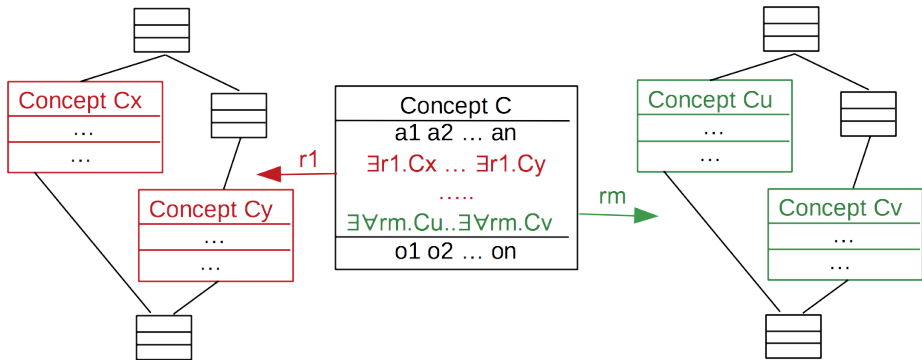


On-Demand: input

Concept C
a1 a2 ... an
o1 o2 ... on

A strategy: pairs of (r_i, ρ_j)
 r_i a relation
 ρ_j a quantifier ($\exists, \exists\forall$)

On-Demand: (1) relational covers

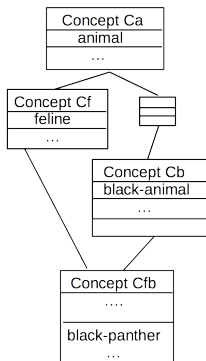


Limiting the number of relational attributes

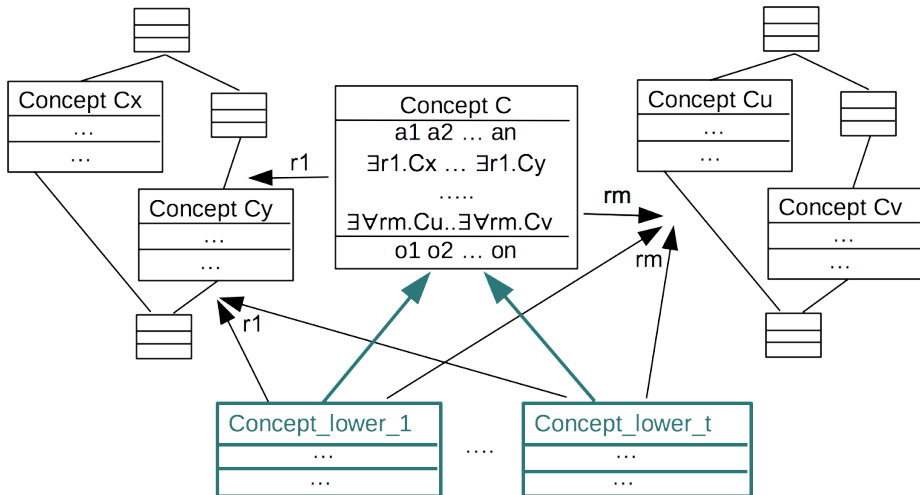
- Using only Object-Concepts (introducing at least an object) to avoid computing whole lattices
- Limited to relations in the strategy
- Specific attribute set intersection

$$A = \{\exists \text{raises.Cf}, \dots\}$$

$$B = \{\exists \text{raises.Cb}, \dots\}$$

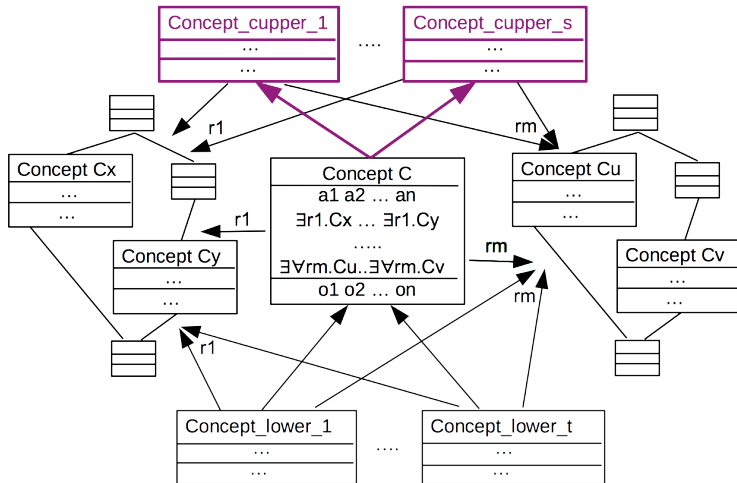
$$A \cap B = \{\exists \text{raises.Ca}, \dots\}$$


On-Demand: (2) lower covers



→ Compute concepts with the maximal extents that are contained in C 's extent and do not contain any of its minimal generators = removing from the extent of C a minimal transversal of the set of minimal generators of C 's extent

On-Demand: (3) upper covers



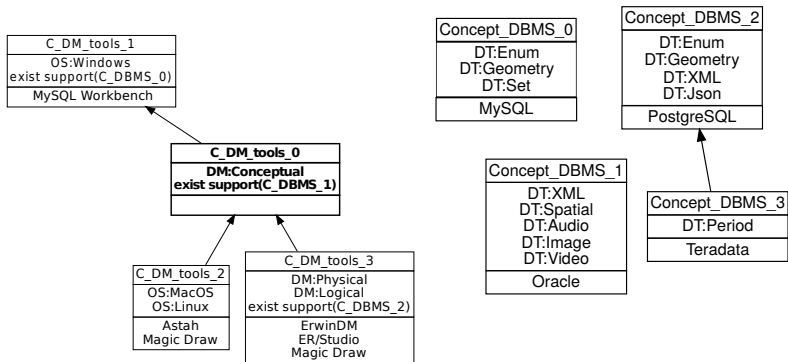
→ Computed by adding an object *o* to the extent of *C*

Integration in RCAExplore

- selection of a formal context, an attribute set
- from concept to concept
- minimal transversal computed with MTMiner [Hébert et al., 2007]

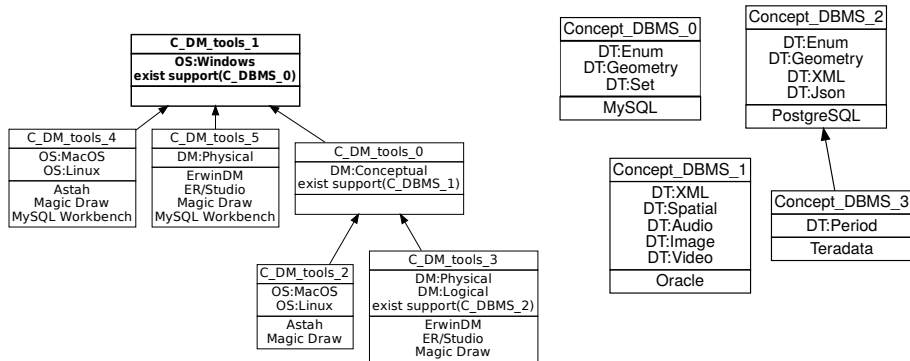
On-demand illustrated (Step 1)

Select a Data Modeling Tool for conceptual model, running on Windows



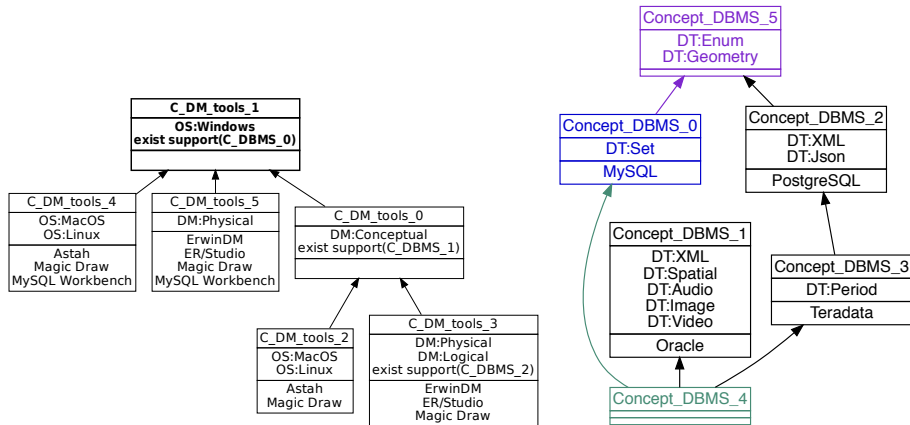
On-demand illustrated (Step 2)

What if I relax constraint "conceptual modeling"? This is not so important, having conceptual or physical modeling would be great.



On-demand illustrated (Step 3)

All DM tools share MySQL. My boss is against. Which other similar DBMS could I find? With which additional/lost properties?



Conclusion






- on-demand computation for RCA
- considers \exists and $\exists\forall$
- relational, upper and lower covers of a concept
- RCAExplore implementation with MTMiner and lattice completion at each step (with-memory approach)

Future work

- graphical user interface, integration within Cogui (conceptual graphs) platform <http://www.lirmm.fr/cogui/>
- reduce computation time for the minimal transversals, e.g. [Murakami and Uno, 2014]
- propose helpers (such as history, breadcrumb trail)
- propose variations (such as no memory progression)
- consider other quantifiers
- user experiment in Fresqueau (hydro-ecology) and Knomana (pesticide plants) projects

Thank you!



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