

# Özgür L. Özçep

## **Database Repairs**

Lecture 12 2 July, 2020

Informationssysteme CS4130 (Summer 2020)

#### References

The paper which started it all

Lit: M. Arenas, L. Bertossi, and J. Chomicki. Consistent query answers in inconsistent databases. In Proceedings of the eighteenth ACM SIGMOD-SIGACT-SIGART symposium on Principles of database systems, PODS '99, pages 68–79, New York, NY, USA, 1999. ACM.

- Recent overview given by one of the founders at the gems of PODS session 2019
   Lit: L. Bertossi. Database repairs and consistent query answering: Origins and further developments. In Proceedings of the 38th ACM SIGMOD-SIGACT-SIGAI Symposium on Principles of Database Systems, PODS '19, pages 48–58, New York, NY, USA, 2019. Association for Computing Machinery.
- The core ideas in a small textbook
   Lit: L. Bertossi. Database Repairing and Consistent Query Answering. Morgan & Claypool Publishers, 2011.

## References (continued)

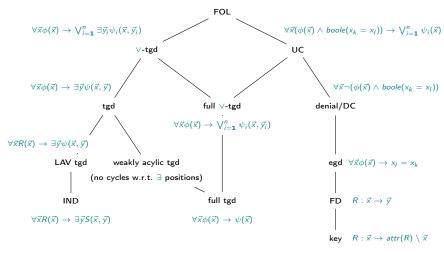
- Slides that are adapted and extended in the following
   Lit: Phokion Kolaitis: Coping with Inconsistent Databases Semantics.
   Algorithms, and Complexity. Invited talk given at the International Conference on Theory and Applications of Satisfiability Testing (SAT 2016), 2016.
- Another nice overview on recent developments (also used here)
   Lit: J. Wijsen. Foundations of query answering on inconsistent databases.
   SIGMOD Rec., 48(3):6–16, Dec. 2019.
- Will occasionally draw connections to belief revision
- Own more extensive additions prefixed by "Ö.:"

## **Basic Notions**

## Integrity constraints (ICs)

- For many (DB) purposes sufficient
  - tuple-generating dependencies (TGDs)
  - equality generating dependencies (EGDs) over the same schema (Compare lectures on data integration and data exchange)
- Special cases of EGDs
  - ► Functional dependency (FD) R : X → Y If two tuples in R agree on X then they agree on Y.
  - Key Constraint  $R: X \longrightarrow Y$  and  $Y = Attributes(R) \setminus X$ .
- There are further ICs ...

## Common Classes of ICs (Wijsen 19))



Convention:  $\phi(\vec{x})$  contains exactly variables in  $\vec{x}$ ;  $\psi(\vec{x}), \psi_i(\vec{x})$  may contain subvector of  $\vec{x}$ 

#### Inconsistent Databases

- Σ: Set of integrity constraints (ICs)
- Inconsistent DB  $\mathfrak{A}$ : Does not satisfy  $\Sigma$ , for short:  $\mathfrak{A} \not\models \Sigma$
- Context of and reasons for inconsistent DBs
  - Lacking support of (some) ICs
  - Heterogeneous sources with different ICs in data integration
  - ► Data warehouse/ETL: data to be cleaned up beforehand

## Coping with Inconsistent DBs

- Data cleaning: Make DB consistent
  - By say adding, deleting, updating rows
  - Chase-procedure can be understood as systematic cleaning (compare notion of Null-based repairs)
  - In industrial-strength practice: ad-hoc, based on heuristics, for specific domains only
  - Main approach in industry (e.g., IBM InfoSphere Quality Stage, Microsoft DQS)
- ► Database repairs: repair (only) virtually & provide DB services
  - In particular: Enable consistent query answering (cqa) over inconsistent DBs
  - Parameters
    - Kinds of allowed repair operations (not discussed in detail)
    - Minimality notion for repairs (see next slides)
  - Compare also: Paraconsistent logics

## Ö.: Relation to Belief Revision

- See discussion on database update in previous lectures
- Can consider DB as a model (corresponding to a complete theory)
- Model describable as belief base B
- Idea: Use Σ as trigger info (multiple revision):
- Data cleaning becomes  $B * \Sigma$
- Hence there is actually a theoretically well investigated theory (belief revision) that could be used for data cleaning
- Why not used?
  - Mainly due to worst-case complexity
  - The same holds, imho, also for data repair (only prototypes available, not industrial-strength software)

#### Definition (Database Repair, (Arenas/Bertossi/Chomicki 99))

DB  $\mathfrak{B}$  is a repair of inconsistent DB  $\mathfrak{A}$  w.r.t. ICs  $\Sigma$  iff

- 1.  $\mathfrak{B} \models \Sigma$
- 2.  $\mathfrak{B}$  is minimally different from  $\mathfrak{A}$
- No unique definition for being minimally different
- In DB community the following instances were investigated
  - Classical set-based repair
  - Cardinality-based repairs
  - Attribute-based repairs (in particular: null-based repairs)
  - Preferred repairs

## Ö.: The problem of minimality in revision

- The idea of minimal repair does not work for belief sets
- Symmetric difference  $X \oplus Y = X \setminus Y \cup Y \setminus X$
- Assume a propositional logic with AGM-constraints

#### Proposition (Rott 2000)

Assume that

- X is a consistent belief set with  $\neg \alpha \in X$
- $Y_1 \neq Y_2$  are belief sets with  $\alpha \in Y_1 \cap Y_2$ .

Then  $X \oplus Y_1$  und  $X \oplus Y_2$  are not comparable w.r.t.  $\subseteq$ , i.e.  $X \oplus Y_1 \not\subseteq X \oplus Y_2$  and  $X \oplus Y_2 \not\subseteq X \oplus Y_1$ 

► Hence all repairs used for revision  $X * \alpha$  would be  $\oplus$ -minimal!

Lit: H. Rott. Two dogmas of belief revision. The Journal of Philosophy, 97(9):503–522, 2000.

## Ö.: The problem of minimality in revision (continued)

- DBs are rather belief bases (not deductively closed)
- Hence Rott's observation not applicable
- Under what kinds of closures non-trivial minimality notion ensured?
- Disjunctive closure allows minimality considerations
   Lit: Özgür L. Özcep. Semantische Integration durch Reinterpretation ein formales Modell, 2009, PhD thesis, http://www.sub.uni-hamburg.de/opus/ volltexte/2010/4428/pdf/oezcepDiss2009.pdf (in German)
   Lit: Özgür L. Özcep. Representation Theorems in Computer Science - A Treatment in Logic Engineering. Springer, 2019.

#### Definition (Classical Subset Repair)

For a set of integrity constraints  $\Sigma$  and an inconsistent database  $\mathfrak{A}$  we say that  $\mathfrak{B}$  is a classical subset-repair of  $\mathfrak{A}$  w.r.t.  $\Sigma$  iff

- 1.  $\mathfrak{B} \subseteq \mathfrak{A}$
- 2.  $\mathfrak{B} \models \Sigma$
- 3. and there is no  $\mathfrak{B}'$  with properties 1. and 2. and  $\mathfrak{B} \subsetneq \mathfrak{B}'$

ICs and database

$$\Sigma = \{ \forall x \forall y \forall z ( (R(x, y) \land R(x, z)) \rightarrow y = z) \}$$
  
$$\mathfrak{A} = \{ R(a_1, b_1), R(a_1, b_2), R(a_2, b_1), R(a_2, b_2) \}$$

Classical subset repairs

$$\mathfrak{B}_{1} = \{R(a_{1}, b_{1}), R(a_{2}, b_{1})\}$$
  

$$\mathfrak{B}_{2} = \{R(a_{1}, b_{1}), R(a_{2}, b_{2})\}$$
  

$$\mathfrak{B}_{3} = \{R(a_{1}, b_{2}), R(a_{2}, b_{1})\}$$
  

$$\mathfrak{B}_{4} = \{R(a_{1}, b_{2}), R(a_{2}, b_{2})\}$$

Exponentially many repairs in general

## Minimal repairs (Wijsen 19)

• Capture notions of minimal repairs with relation  $\leq_{\mathfrak{A}}^{\Sigma}$ 

Definition (Binary repair relation  $\leq_{\mathfrak{A}}^{\Sigma}$  (informal))

 $\mathfrak{B}_1 \leq_{\mathfrak{A}}^{\Sigma} \mathfrak{B}_2$  iff repair of  $\mathfrak{A}$  into  $\mathfrak{B}_1$  requires no more effort than repair of  $\mathfrak{A}$  into  $\mathfrak{B}_2$ 

• Minimal repairs  $Min_{\leq \frac{\Sigma}{2}}(\mathfrak{A})$  of  $\mathfrak{A}$ :

 $\{\mathfrak{B} \mid \mathfrak{B} \models \Sigma \text{ and there is no } \mathfrak{B}' \text{ s.t.: } \mathfrak{B}' \models \Sigma \text{ and } \mathfrak{B}' <^{\Sigma}_{\mathfrak{A}} \mathfrak{B}\}$ 

- What properties to require of  $<_{\mathfrak{A}}^{\Sigma}$ ?
  - Acyclicity

Minimal repairs w.r.t. some order

Symmetric-difference order

 $\mathfrak{B}_1 \leq_{\mathfrak{A},\oplus} \mathfrak{B}_2 \text{ iff } \mathfrak{B}_1 \oplus \mathfrak{A} \subseteq \mathfrak{B}_2 \oplus \mathfrak{A}$ 

► Transitive, reflexive, antisymmetric (partial order)

• Note:  $\mathfrak{B}_1, \mathfrak{B}_2$  not necessarily subsets of  $\mathfrak{A}$ 

Cardinality order

 $\mathfrak{B}_1 \leq_{\mathfrak{A},c} \mathfrak{B}_2$  iff  $|\mathfrak{B}_1 \oplus \mathfrak{A}| \leq |\mathfrak{B}_2 \oplus \mathfrak{A}|$ 

Transitive and reflexiv (pre-order)

Definition (General Subset [Superset] Repair)

 $\mathfrak{B}$  is a general subset-repair [superset-repair] of  $\mathfrak{A}$  w.r.t.  $\Sigma$  iff it is a  $\leq_{\mathfrak{A}}^{\Sigma}$ -minimal repair (for some relation  $\leq_{\mathfrak{A}}^{\Sigma}$ ) and  $\mathfrak{B} \subseteq \mathfrak{A}$  [ $\mathfrak{B} \supseteq \mathfrak{A}$ ].

### Preferred/Prioritized repairs

- Assume  $\Sigma$  consists of FDs
- Inconsistent prioritizing db  $(\mathfrak{A},\succ)$ 
  - Intuition:  $f \succ g$  iff f is fact prioritized over fact g
  - $\succ$  acyclic with:  $f \succ g$  entails  $\{f, g\} \not\models \Sigma$ .

#### Definition (Prioritization-order and g-repair)

► For  $\mathfrak{B}_1, \mathfrak{B}_2 \subseteq \mathfrak{A}, \mathfrak{B}_1 \models \Sigma$ , and  $\mathfrak{B}_2 \models \Sigma$ :  $\mathfrak{B}_1 \leq_{\mathfrak{A},\succ} \mathfrak{B}_2$  iff for every  $g \in \mathfrak{B}_2 \setminus \mathfrak{B}_1$  there is  $f \in \mathfrak{B}_1 \setminus \mathfrak{B}_2$  s.t.  $f \succ g$ .

▶  $\mathfrak{B}$  is a globally optimal repair/g-repair iff  $\mathfrak{B} \in Min_{\leq_{\mathfrak{A},\succ}}(\mathfrak{A})$ .

Lit: S. Staworko, J. Chomicki, and J. Marcinkowski. Prioritized repairing and consistent query answering in relational databases. Annals of Mathematics and Artificial Intelligence, 64(2):209–246, 2012.

#### Example

- ► DB instance over R(X, Y): {R(a, b), R(c, b), R(c, d)} Constraints  $\Sigma$ : { $R: X \to Y, R: Y \to X$ }
- ► ⊕-repairs:
  - {R(a,b), R(c,d)}
  - $\{R(c,b)\}$
- C-repair
  - $\{R(a, b), R(c, d)\}$
- ► G-repair for  $\succ$  with  $R(c, b) \succ R(a, b)$  and  $R(c, b) \succ R(c, d)$ ►  $\{R(c, b)\}$

Null-based repairs: tuple level

Input DB 31

 $\{ S(c_1, r_1, i_1), S(c_2, r_2, i_2), S(c_3, r_3, i_3), Supply(Company, Receiver, Item) \\ A(i_1, 50), A(i_2, 30) \}$ Article(Item, Cost)

• Constraints  $\Sigma$ : { $\forall x \forall y \forall z [S(x, y, z) \rightarrow \exists v A(z, v)]$ } (an IND)

#### ► Repairs

- $\mathfrak{B}_1$ : delete  $S(c_3, r_3, i_3)$  from  $\mathfrak{A}$
- $\mathfrak{B}_2$ : Insert  $(i_3, NULL)$  into relation A.

Lit: L. E. Bertossi and L. Bravo. Consistency and trust in peer data exchange systems. Theory Pract. Log. Program., 17(2):148–204, 2017.

#### Null-based repairs: attribute level

- Input DB A
  - $\{ R(a_4, a_3), R(a_2, a_1), R(a_3, a_3), \\ S(a_4), S(a_2), S(a_3) \}$
- Constraints  $\Sigma$ : { $\neg \exists x \exists y (S(x) \land R(x, y) \land S(y))$ } (a DC)
- Repairs
  - $\mathfrak{B}_1$ : {  $R(a_4, a_3), R(a_2, a_1), R(a_3, a_3), S(a_4), S(a_2), S(NULL)$ }
  - ▶  $\mathfrak{B}_2$ : {  $R(a_4, NULL), R(a_2, a_1), R(a_3, NULL), S(a_4), S(a_2), S(a_3)$ }
- Note: Null prevents a join

Lit: Bertossi and L. Li. Achieving data privacy through secrecy views and null-based virtual updates. IEEE Transactions on Knowledge and Data Engineering, 25(5):987–1000, May 2013.

## Consistent Query Answering (CQA)

Definition (Arenas, Bertossi, Chomicki 99)

The consistent answers of Q on  $\mathfrak{A}$  w.r.t.  $\Sigma$  is the set

 $cqa_r(Q, \mathfrak{A}, \Sigma) = \bigcap \{Q(\mathfrak{B}) \mid \mathfrak{B} \text{ is an } r\text{-repair of } \mathfrak{A} \text{ w.r.t. } \Sigma \}$ 

- Follows the usual pattern for dealing with incomplete information
- Compare this with certain answers in data exchange and OBDA
- ▶ When clear from context repair type *r* not mentioned.

#### Example (Consistent query answering)

- ► ICs, database, and classical subset-repairs
  - $\Sigma = \{ \forall x \forall y \forall z ( (R(x, y) \land R(x, z)) \rightarrow y = z) \}$

 $\mathfrak{A} = \{R(a_1, b_1), R(a_1, b_2), R(a_2, b_1), R(a_2, b_2)\}$ 

- $\mathfrak{B}_1 = \{R(a_1, b_1), R(a_2, b_1)\}$
- $\mathfrak{B}_2 = \{R(a_1, b_1), R(a_2, b_2)\}$
- $\mathfrak{B}_3 = \{R(a_1, b_2), R(a_2, b_1)\}$
- $\mathfrak{B}_4 = \{R(a_1, b_2), R(a_2, b_2)\}$

Queries and answers

 $Q_1(x) = \exists y R(x, y)$  $cqa(Q_1, \mathfrak{A}, \Sigma) = \{a_1, a_2\}$ 

 $Q_2(x) = \exists z R(z, x)$  $cqa(Q_1, \mathfrak{A}, \Sigma) = \emptyset$ 

## Complexity and Dichotomies

### Two main decision problems

#### Definition (Decision problem $CERTAINTY_r(Q, \Sigma)$ )

Boolean query Q and  $\Sigma$  fixed.

- Input: database 31
- Output:  $cqa_r(Q, \Sigma, \mathfrak{A})$

Complexity ranges from polynomial time computability to undecidability

Definition (Decision problem  $REPAIR_r(\Sigma)$ )

Σ fixed.

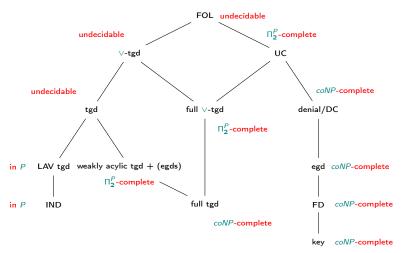
- ► Input: databases 𝔐, 𝔅
- Output: Is  $\mathfrak{B}$  an *r*-repair of  $\mathfrak{A}$  w.r.t.  $\Sigma$ ?

Complexity ranges from polynomial time computability to coNP-completeness

#### Definition

- Consistent query answering for class of queries Q and class of integrity constraints *IC* is in complexity class C iff CERTAINTY(Q, Σ) is in C for all Q ∈ Q and all Σ ⊆ *IC*.
- Consistent query answering for Q and IC is C-complete iff it is in C and CERTAINTY(Q, Σ) is C-complete for some Q ∈ Q and some Σ ⊆ IC.

## Complexities for $\textit{CERTAINTY}_\oplus(Q,\Sigma)$ for $Q \in \textit{CQs}$

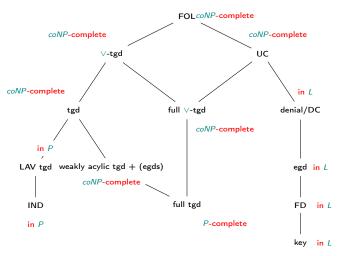


Lit: S. Arming, R. Pichler, and E. Sallinger. Complexity of repair checking and consistent query answering. ICDT, pages 21:1–21:18, 2016.

What do the complexity results tell us?

- Even for very common queries and ICs untractable problems (coNP-complete problems)
- But: By definition this means only that there are some intractable (coNP-problems); does not say anything about tractable-untractable boundary
- Tackle this with dichtomy/trichotomy theorems

## Complexities for $\textit{REPAIR}_{\oplus}(\Sigma)$ for $\textit{Q} \in \textit{CQs}$



Lit: S. Arming, R. Pichler, and E. Sallinger. Complexity of repair checking and consistent query answering. ICDT, pages 21:1–21:18, 2016.

## Complexity of CQA: An Illustration

- ► R, S: binary relations with first argument as keys:  $\Sigma = \{R(u, v) \land R(u, w) \rightarrow v = w, S(u, v) \land S(u, w) \rightarrow v = w\}$
- $\blacktriangleright Q_1 = \exists x, y, z(R(x, y) \land S(y, z))$ 
  - $CERTAINTY(Q_1, \Sigma) \in P$
  - ► Even FOL-rewritable  $\exists x, y, z(R(x, y) \land S(y, z) \land \forall y'[R(x, y') \rightarrow \exists z'S(y', z')])$
- $Q_2 = \exists x, y(R(x, y) \land S(y, x))$ 
  - CERTAINTY $(Q_2, \Sigma) \in P$
  - but not FOL-rewritable
- $Q_3 = \exists x, y, z(R(x, y) \land S(z, y))$ 
  - CERTAINTY( $Q_2, \Sigma$ ) coNP-complete
- Note: All queries are CQs but of different types
   Classification with di-/trichotomies

### Descriptive Complexity and Rewriting

- Instead of computational complexities can also use descriptive complexity
- Remember notion of logic *L* capturing a complexity class and notion of rewriting.
- ► CERTAINTY(Q, Σ) is expressible in L, alias is L-rewritable, iff there is Q<sub>rew</sub> ∈ L s.t.

 $cqa(Q, \mathfrak{A}) = true \text{ iff } \mathfrak{A} \models Q_{rew}$ 

Most attractive: L = FOL

#### Famous open question

#### Dichotomy-Conjecture

For every set  $\Sigma$  of primary keys, for every query Q that is a disjunction of Boolean CQs, *CERTAINTY* $(Q, \Sigma)$  is either in *P* or *coNP*-complete.

### Reminder: Dichotomies

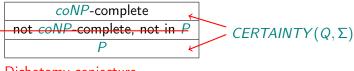
#### Theorem (Ladner 1975)

If  $P \neq NP$  then there is a decision problem Q s.t.

- $\blacktriangleright$  Q is in NP but not in P
- ▶ *Q* is not *NP*-complete

(Similar results for *coNP* obtainable.)

The fine structure of coNP



Dichotomy conjecture

### Progress towards the Conjecture or: 3 is more than 2

#### Theorem (Trichotomy (Koutris/Wijsen 19))

For every set  $\Sigma$  of primary keys and self-join-free Boolean CQs Q, CERTAINTY $(Q, \Sigma)$  is either in FOL or L-complete or coNP-complete.

The proof moreover reveals:

- 1. Membership in *L* shown by rewriting into symmetric stratified Datalog with aggregation.
- 2. Membership in tractable classes  $(FOL \cup L)$  iff joins are foreign-key-primary-key joins

 $\implies$  most SPJ queries are tractable

Lit: P. Koutris and J. Wijsen. Consistent query answering for primary keys in logspace. In: ICDT 2019, pages 23:1–23:19, 2019

#### Connections to CSPs

- Unexpected connection between Consistent Query Answering and Constraint Satisfaction Problems (CSPs).
- Shows that the dichotomy conjecture for CERTAINTY likely not trivial (as proof of CSP dichotomy highly non-trivial)

#### Theorem (Fontaine 2015 (informal))

CERTAINTY( $Q, \Sigma$ ) dichotomy (Conjecture)  $\models$  CSP dichotomy under specific conditions

Lit: G. Fontaine. Why is it hard to obtain a dichotomy for consistent query answering? ACM Trans. Comput. Log., 16(1):7:1–7:24, 2015.

#### Constrain Satisfaction Problems

 Traditionally (as used in AI research) considered as subclass of search problems with states and a goal test

#### Definition

- ► D = domain
- $\Gamma$  = Constraint language = set of relations  $\{R_i\}_{i \in I}$  over D

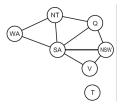
The constraint satisfaction problem  $CSP(\Gamma)$  associated with  $\Gamma$  is the problem defined by instances of the form (V, D, C) where

- V = set of variables
- $C = \text{set of constraints } (\vec{v}, R_i) \text{ (notated also as } R_i(\vec{v}) \text{) with}$ 
  - $R_i \in \Gamma$  an *n*-ary relation
  - $\vec{v} = (v_1, \dots, v_n)$  with  $v_i \in V$  is the scope (state variables)

A solution to the problem (the goal) is a mapping  $\phi : V \longrightarrow D$  fulfilling all constraints, i.e.,  $\phi(\vec{v}) \in R_i$  for all  $(\vec{v}, R_i) \in C$ .

## Example: Map Colouring





- $D = \{red, green, blue\}$
- ►  $\Gamma = \{R_1\} = \{\{(x, y) \in D \times D \mid x \neq y\}\}$
- $\blacktriangleright V = \{WA, NT, Q, NSW, VI, SA, T\}$
- ►  $C = \{ WA \neq NT, WA \neq SA, NA \neq SA, NA \neq Q, SA \neq Q, SA \neq Q, SA \neq NSW, SA \neq VI, Q \neq NSW, NSW \neq VI \}$
- A solution

 $\begin{aligned} \phi: & \textit{WA} \mapsto \textit{red}, \textit{NT} \mapsto \textit{green}, \textit{SA} \mapsto \textit{blue}, \textit{Q} \mapsto \textit{red}, \\ & \textit{NSW} \mapsto \textit{green}, \textit{VI} \mapsto \textit{red}, \textit{SA} \mapsto \textit{blue}, \textit{T} \mapsto \textit{green} \end{aligned}$ 

## CSP = Finding a homomorphism

#### Definition

Given a  $CSP(\Gamma)$  instance (V, D, C) the associated homomorphism instance  $h : \mathfrak{S} \xrightarrow{hom} \mathfrak{T}$  is defined by

- ► source structure  $\mathfrak{S} = (\text{variables, scopes})$  $\mathfrak{S} = (V, R_i^{\mathfrak{S}} = \{\vec{v} \mid (\vec{c}, R) \in C\})_{i \in I}$
- ► target structure  $\mathfrak{T} = ($ values, constraint relations) $\mathfrak{T} = \{D, R_i^{\mathfrak{T}} = R_i\}_{i \in I}$
- homomorphism h = solution  $\phi$

Each homomorphism problem  $?\exists h : \mathfrak{A} \xrightarrow{hom} \mathfrak{B}$  gives rise to a CSP-instance: Generate constraint  $(\vec{v}, R^{\mathfrak{B}})$  for each  $\vec{v} \in R^{\mathfrak{A}}$ . Lit: T. Feder and M. Y. Vardi. The computational structure of monotone monadic SNP and constraint satisfaction: A study through datalog and group theory. SIAM J. Comput., 28:57–104, 199 Lit: P. G. Kolaitis and M. Y. Vardi. A Logical Approach to Constraint Satisfaction, pages 125–155. Springer, 2008

### Dichotomies for CSP

- ► Complexity of finding solutions depends on Γ
- Dichotomy theorem for subclass of conservative CSPs (c-CSP) which are CSPs with additionally:
  - For each variable  $v \in V$  one has a unary relation  $R_v \subseteq D$
  - Solution  $\phi$  must fullfil  $\phi(v) \in R_v$ .

Theorem (Dichotomy for conservative CSPs, (Bulatov 11))

For each  $\Gamma$  the problem  $c - CSP(\Gamma)$  is either in P or NP-complete.

 Proof relies on algebraic machinery based on polymorphisms (Barto et al., 17)

Lit: A. Bulatov. Complexity of conservative constraint satisfaction problems. ACM Trans. Comput. Log., 12(4):24:1–24:66, 2011.

Lit: L. Barto, A. Krokhin, and R. Willard. Polymorphisms, and How to Use Them. In A. Krokhin and S. Zivny, editors, The Constraint Satisfaction Problem: Complexity and Approximability, volume 7 of Dagstuhl Follow-Ups, pages 1–44. Schloss Dagstuhl,Leibniz-Zentrum für Informatik, Dagstuhl, Germany, 2017.

#### Connections to CSPs

#### Theorem (Fontaine 2015 formal)

- ► CERTAINTY(Q, Σ) dichotomy (Conjecture) ⊨ CSP dichotomy where
  - $\Sigma$  is a finite set of Horn constraints (full tgd with atomic head)
  - Q is a union of Boolean CQs
- ► CERTAINTY(Q,Σ) dichotomy (Conjecture) ⊨ c-CSP dichotomy

Lit: G. Fontaine. Why is it hard to obtain a dichotomy for consistent query answering? ACM Trans. Comput. Log., 16(1):7:1–7:24, 2015.

Alternative route for proving the conjecture

Breakthrough result in 2017 for wider class of CSP

Theorem (Dichotomy for CSPs, (Bulatov 17))

For each  $\Gamma$  the problem  $CSP(\Gamma)$  is either in P or NP-complete.

 Proof strategy for conjecture: Show that it is entailed by Bulatov's dichotomy for CSPs.

Lit: A. A. Bulatov. A dichotomy theorem for nonuniform CSPs. In C. Umans, editor, 58th IEEE Annual Symposium on Foundations of Computer Science, FOCS 2017, Berkeley, CA, USA, October 15-17, 2017, pages 319–330. IEEE Computer Society, 2017.



#### Theory and Practice

- Theory of database repairs a theoretical foundation for coping with inconsistent DBs
- Extensively studied in last 20 years
- Only marginally used in data cleaning (few examples given by (Bertossi 19)
- Industrial-strength cqa-systems have yet to be developed

Lit: L. Bertossi. Database repairs and consistent query answering: Origins and further developments. In Proceedings of the 38th ACM SIGMOD-SIGACT-SIGAI Symposium on Principles of Database Systems, PODS '19, pages 48–58, New York, NY, USA, 2019.

## Prototypes (optional slide)

#### Hippo

Lit: J. Chomicki, J. Marcinkowski, and S. Staworko. Hippo: A system for computing consistent answers to a class of sql queries. In EDBT 2004, pages 841–844, Springer, 2004.

#### ConQuer

Lit: A. Fuxman and R. J. Miller. First-order query rewriting for inconsistent databases. J. Comput. Syst. Sci., 73(4):610–635, 2007.

#### ConsEx

Lit: M. C. Marileo and L. E. Bertossi. The consistency extractor system: Answer set programs for consistent query answering in databases. Data Knowl. Eng., 69(6):545–572, 2010.

#### EQUIP

Lit: P. G. Kolaitis, E. Pema, and W. Tan. Efficient querying of inconsistent databases with binary integer programming. PVLDB, 6(6):397–408, 2013.

## Feature Overview (Optional Slide)

System	Constraints	Queries	Method
Hippo	UC	Projection-free	Direct Alg.
		with $\cup$ and $\setminus$	
ConQuer	Keys	CQs	FO-rewriting
ConsEx	UC + IND	Datalog	ASP
	with acyclicity		
EQUIP	Keys IND	CQs	Reduction to ILP

Kolaitis' vision for a comprehensive system (Optional Slide)

Module-based system depending on complexity of  $cqa(Q, \Sigma)$ 

- ► Preprocessing: Determine evaluation strategy based on complexity classification for cqa(Q, Σ)
- Processing
  - Module A: FOL-rewriting + DB engine if cqa(Q, Σ) FOL rewritable
  - ► Module B: Direct Algorithm or reduction to LP if cqa(Q, Σ) ∈ P \ FOL
  - Module C: Reduction to ILP (oe SAT or QBF) if cqa(Q, Σ) ∈ coNP.

### Lifting to ontologies

Whole idea can of course be lifted also to ontologies

 For a recent contribution for DL-Lite ontologies see (Bienvenu et al 2019)

Lit: M. Bienvenu, C. Bourgaux, and F. Goasdoué. Computing and explaining query answers over inconsistent dl-lite knowledge bases. J. Artif. Int. Res., 64(1):563–644, Jan. 2019.

## Synopsis and Outlook (Kolaitis + Ö.)

- Database repair meeting point for database, logic, and complexity
- Further dichotomies; main conjecture open
- Much to be done for industrial-strength systems for different types of repairs and classes of constraints
  - Promising approach: Combine database engines with SAT solvers and QBF solvers
  - Ö: This fits to the general trend of "SQL-incorporates it all"
    - SQL now supports arrays
    - SQL is going to give support for streams ...
- Ö: Systematic study of connections to belief revision
- Ö: Are there dichotomies also for belief revision?