SQL, DLs, Datalog, and ASP: comparison

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CWA vs. OWA

- DLs are based on the semantics of classical logic
- Open-world assumption (OWA):
 - The knowledge expressed by the theory is not complete
 - Many possible worlds (models)
- Databases are based on a closed-world assumption (CWA):
 - The knowledge expressed by the database is complete
 - Only one world is possible (the one completely described by the database)
- Datalog and ASP inherit CWA from databases
- But: ASP is able to deal with incomplete knowledge (disjunction)



Complete vs. incomplete knowledge

- DL knowledge bases are incomplete specifications of the domain of interest
- The KB has multiple models (possible worlds in which the KB is satisfied)
- The TBox and the ABox can be used to infer implicit facts about the world

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- E.g., from
ABox = { student(Bob) } and
TBox = { subClassOf(student, person) }
we conclude person(Bob) (which is NOT stated in the ABox)
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Complete vs. incomplete knowledge

- Conversely, the knowledge expressed by a relational database is assumed to be complete
- Only one world is possible (the one completely described by the database)
- In relational DBs, the database instance is a complete specification:
 - Schema constraints in databases are integrity constraints:
 they MUST be satisfied by the database instance (otherwise the database is in an illegal state)
 - For this reason, such constraints cannot derive new facts (not appearing in the database instance)



Recursion

- DLs are able to deal with recursive statements
 - E.g., cyclic definitions of concepts in the TBox
- DLs do not allow for recursive queries
- SQL does not allow for recursive queries
 - only a limited form of recursion in views is allowed
 - Datalog allows for expressing recursive queries over databases
- ASP allows for expressing recursive queries over databases





Negation

- Several DLs allow for expressing negation in the TBox (and in the ABox) and in queries
- But, DLs do not allow for recursive negation in queries
- SQL allows for negation in queries, but does not allow for recursive negation
- Datalog does not allow for expressing negation
- ASP allows for expressing negation, and even recursive negation

Value invention (skolemization)

- DLs allow for expressing so-called value invention (or mandatory participation to roles)
- Example: inclusion axiom Person ⊆ ∃hasMother . T
- The above axiom cannot be represented in Datalog
- E.g., the rule

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hasMother(x,y) :- Person(x).
```

is not range-restricted (so it is not allowed in Datalog)

- Datalog does not allow for talking about elements outside the Herbrand Universe of the program (namely, the constants appearing in the program)
- Same restriction holds for ASP (and SQL too)
- Value invention is a form of incomplete knowledge



Existential rules

- To overcome the above limitation, an extension of Datalog called existential rules, or Datalog +/-, has been proposed
- Existential rules do not have the range restriction on variables: a variable can appear in the head of the rule
- Such variables are interpreted as existentially quantified variables
- E.g.:

hasMother(x,y) :- Person(x).

is an existential rule, whose meaning is: if x is a person, then there exists y such that x has mother y

In this way, value invention can be expressed by rules



Existential rules

- Extending Datalog with existential variables in the head makes the language more expressive
- On the other hand, such an extension makes reasoning harder: in general, reasoning with existential rules is undecidable
- Recent studies have defined restricted classes of existential rules in which reasoning is decidable (and even tractable in data complexity)
- Interesting relationship between some of these classes and Description Logics



Comparison: expressiveness

	Semantics	Recursion	NOT (stratif.)	NOT (recursive)	OR	Value invention
SQL	CWA	no	yes	no	yes	no
DL	OWA	yes/no	yes	no	yes	yes
Positive Datalog	CWA	yes	no	no	no	no
Datalog + stratified negation	CWA	yes	yes	no	no	no
Datalog + negation	CWA	yes	yes	yes	no	no
ASP	CWA	yes	yes	yes	yes	no

Complexity of SQL

- Complexity of evaluating queries in SQL is NP-complete
- A query may have an exponential number of answers
- Example:

Table T:

A

Query:

0

1

select R1.A, R2.A, R3.A,...,Rn.A from TAS R1, TAS R2, TAS R3, ..., TAS Rn

Answers: (2ⁿ tuples)

R1.A	R2.A	R3.A	•••	Rn-1.A	Rn.A
0	0	0	•••	0	0
0	0	0	•••	0	1
	•••	•••	•••	•••	•••
1	1	1		1	0
1	1	1		1	1

Complexity of SQL (contd.)

Previous SQL query:

select R1.A, R2.A, R3.A,...,Rn.A from T AS R1, T AS R2, T AS R3, ..., T AS Rn

Same query expressed in Datalog:

Q(X1, X2, ..., Xn) := T(X1), T(X2), ..., T(Xn).

(which shows that Datalog can build IDB predicates with an exponential number of tuples in the minimal model)



Comparison: complexity

	Reasoning task	Complexity
SQL	Query evaluation	NP-complete
ALC, unfoldable TBoxes	Concept consistency, KB satisfiability, instance checking	PSPACE-complete
ALC, GCIs	Concept consistency, KB satisfiability, instance checking	EXPTIME-complete



Comparison: complexity

	Reasoning task	Complexity
Ground positive Datalog	Building the minimal model	PTIME-complete
Ground Datalog with stratified negation	Building the minimal model	PTIME-complete
Ground ASP (no disjunction)	Consistency, brave reasoning	NP-complete
Ground ASP	Consistency, brave reasoning	NP ^{NP} -complete



Comparison: complexity

	Reasoning task	Complexity
Positive Datalog	Building the minimal model	EXPTIME-complete
Datalog with stratified negation	Building the minimal model	EXPTIME-complete
ASP (no disjunction)	Consistency, brave reasoning	NEXPTIME- complete
ASP	Consistency, brave reasoning	NEXPTIME ^{NP} - complete

