

Our application: **situation recognition for context-aware systems**

- observe complex hard- & software system
Goal: save energy
- recognize situations, where adaptation is beneficial
- **Description Logic-based** approach:
TBox for background knowledge, ABox for observations
- employ Description Logic reasoning: **conjunctive query answering**
- collecting data from several sources! — **sensors**

Application setting

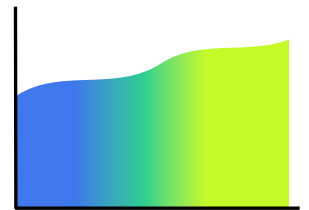
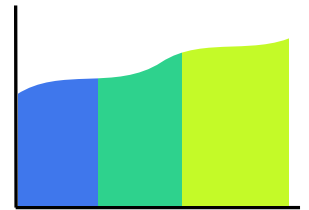
- data (& ontology) stored in a **data base**
- **sliding windows** approach to monitored data
 - observations are recorded at fixed intervals
 - sequence of system snapshots is obtained
- raw data is **preprocessed** by data base
 - raw data is 'cleaned' and aggregated
- situation descriptions are stored as **DB queries**



Why fuzzy information?

Represent **sensor values** 'faithfully'

- Logic: symbolic representation of values
- values: low, medium, high
- small change in value \longrightarrow similar classification
- can level out noise



Semantics:

- **Classical logic:** $\{0, 1\}$ binary
- **Fuzzy logic:** $\{0, \dots, 1\}$ infinitely many **membership degrees**

Results on DL & fuzzy information:

- easily **undecidable!**
- decidable, if **finitely many** membership degrees are used.
But: **costly reduction!**

Our goal: employ **reasoner for DL-Lite_R** (a.k.a. OWL 2 QL)

- DL-Lite_R-reasoners: **optimized implementations!**
- only the ABox and the queries are **fuzzy!**
(TBox stays crisp)

The description logic DL-Lite_R and its fuzzy variant

DL-Lite_R **concepts**: $B \rightarrow A \mid \exists Q \quad C \rightarrow \top \mid B \mid \neg B$
 $Q \rightarrow P \mid P^- \quad R \rightarrow Q \mid \neg Q$

DL-Lite_R **axioms**: $B \sqsubseteq C \quad Q \sqsubseteq R \quad \text{funct}(Q)$

Fuzzy DL-Lite_R **assertions**: $\langle B(a), d \rangle \quad \langle P(a, b), d \rangle$

Interpretation for fuzzy DL-Lite_R: $\mathcal{I} = (\Delta^{\mathcal{I}}, \cdot^{\mathcal{I}})$

- **individuals**: $a^{\mathcal{I}} \in \Delta^{\mathcal{I}}$
- **named concepts**: $A^{\mathcal{I}} : \Delta^{\mathcal{I}} \rightarrow [0, 1]$
- **atomic roles**: $P^{\mathcal{I}} : \Delta^{\mathcal{I}} \times \Delta^{\mathcal{I}} \rightarrow [0, 1]$

Fuzzy operators and semantics of complex concepts

	t-norm $a \otimes b$	negation $\ominus a$
Gödel	$\min(a, b)$	$\begin{cases} 1, & a = 0 \\ 0, & a > 0 \end{cases}$
Lukasiewicz	$\max(a + b - 1, 0)$	$1 - a$
Product	$a \times b$	$\begin{cases} 1, & a = 0 \\ 0, & a > 0 \end{cases}$

Let: $\delta, \delta' \in \Delta^{\mathcal{I}}$

complex concepts:

- $(\exists Q)^{\mathcal{I}}(\delta) = \sup_{\delta' \in \Delta^{\mathcal{I}}} Q^{\mathcal{I}}(\delta, \delta')$
- $(\neg B)^{\mathcal{I}}(\delta) = \ominus B^{\mathcal{I}}(\delta)$
- $\top^{\mathcal{I}}(\delta) = 1$

fuzzy conjunctive query:

$$q(\vec{x}) = \exists \vec{y}. \varphi(\vec{x}, \vec{y}) \geq d$$

Example

$$\mathcal{T} := \{\text{Server} \sqsubseteq \exists \text{hasCPU}, \exists \text{hasCPU}^- \sqsubseteq \text{CPU}, \text{func}(\text{hasCPU}^-)\}$$

$$\mathcal{A} := \{\langle \text{Server}(\text{server}_1), 1 \rangle, \langle \text{hasCPU}(\text{server}_1, \text{cpu}_1), 1 \rangle, \\ \langle \text{OverUtilized}(\text{cpu}_1), 0.6 \rangle, \langle \text{hasCPU}(\text{server}_1, \text{cpu}_2), 1 \rangle, \\ \langle \text{OverUtilized}(\text{cpu}_2), 0.8 \rangle\}$$

$$q_1(x) = \text{CPU}(x)$$

$$q_2(x, y) = \text{hasCPU}(x, y) \wedge \text{OverUtilized}(y)$$

$$q_3(x) = \exists y \text{ hasCPU}(x, y) \wedge \text{OverUtilized}(y)$$

$$\text{ans}(q_1(x), \mathcal{O}) = \{(\text{cpu}_1, 1), (\text{cpu}_2, 1)\}$$

$$\text{ans}(q_2(x, y), \mathcal{O}) = \{(\text{server}_1, \text{cpu}_1, 0.6), (\text{server}_2, \text{cpu}_2, 0.8)\}$$

$$\text{ans}(q_3(x), \mathcal{O}) = \{(\text{server}_1, 0.8)\}.$$

Approach for query answering in DL-Lite

- 1.) Use TBox to reformulate query q into a FOL-query $q_{\mathcal{T}}$,
discard the TBox
"Compile TBox information into the query."
- 2.) View ABox \mathcal{A} as a relational database $\mathcal{I}_{\mathcal{A}}$, where
 - $A^{\mathcal{I}_{\mathcal{A}}} = \{a \mid A(a) \in \mathcal{A}\}$
 - $r^{\mathcal{I}_{\mathcal{A}}} = \{(a, b) \mid r(a, b) \in \mathcal{A}\}$
- 3.) Evaluate $q_{\mathcal{T}}$ in the DB $\mathcal{I}_{\mathcal{A}}$ using a relational query engine.

The black box rewriting algorithm for fuzzy queries

1. Compute **crisp query rewriting** $(q(\vec{x}) \rightsquigarrow q_{\mathcal{T}}(\vec{x}))$

2. **rewrite** query $q_{\mathcal{T}}(\vec{x})$ by

- **degree variables:** placeholders for degrees, e.g.: x_d
- replacing

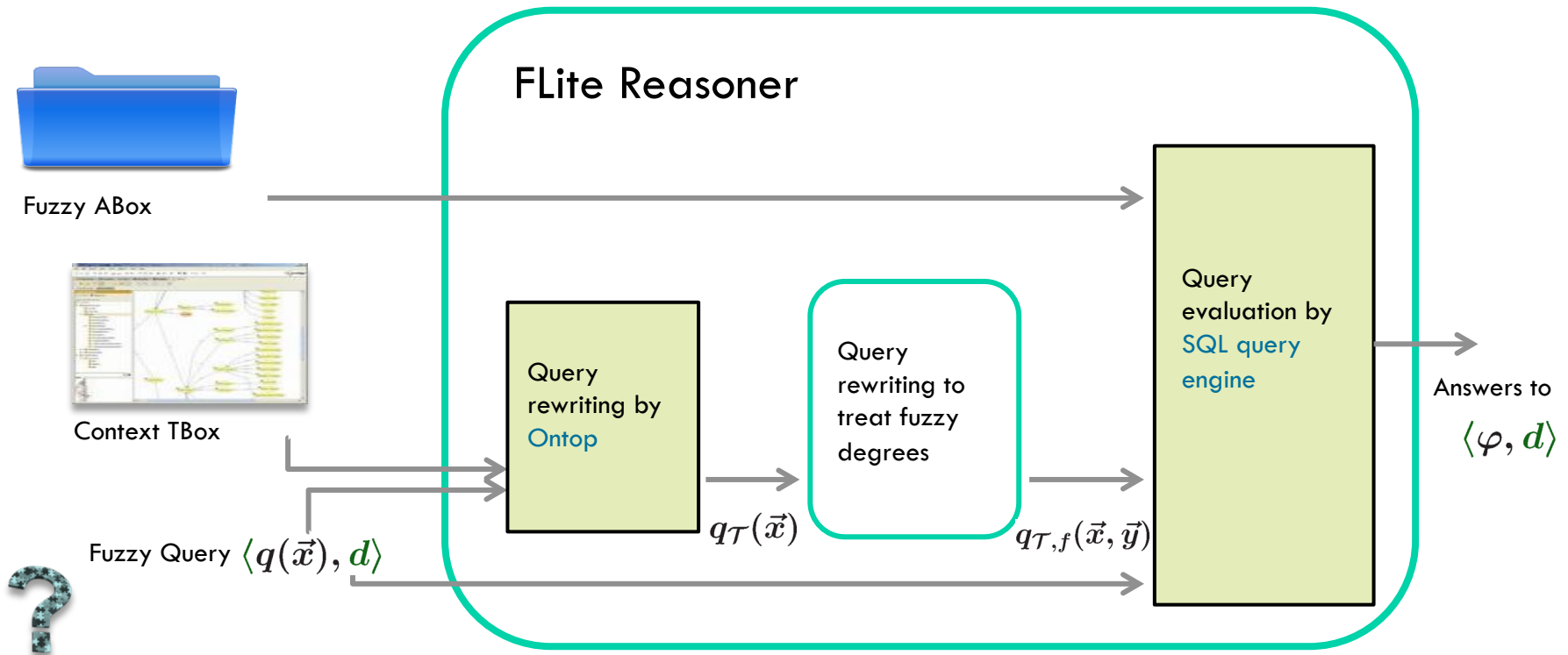
- concepts in concept atoms by binary predicates
- roles in role atoms by ternary predicates

where new argument is a degree variable
(to prepare for the tables storing fuzzy information)

- **degree atoms:**
 - to capture the fuzzy operators, e.g.: $\Phi_{\otimes}(x_d, x'_d), \Phi_{\ominus}(x_d)$
 - to ensure conditions on degrees, e.g.: $\Phi_{\geq}(x_d, 0.3)$

The black box rewriting algorithm for fuzzy queries

1. Compute **crisp query rewriting** $(q(\vec{x}) \rightsquigarrow q_{\mathcal{T}}(\vec{x}))$
2. **rewrite query** $(q_{\mathcal{T}}(\vec{x}) \rightsquigarrow q_{\mathcal{T},f}(\vec{x}, \vec{x}_d))$
3. **Evaluate** rewritten query $q_{\mathcal{T},f}(\vec{x}, \vec{x}_d)$ over the 'fuzzy' database
4. Keep **highest degree** for each returned tuple



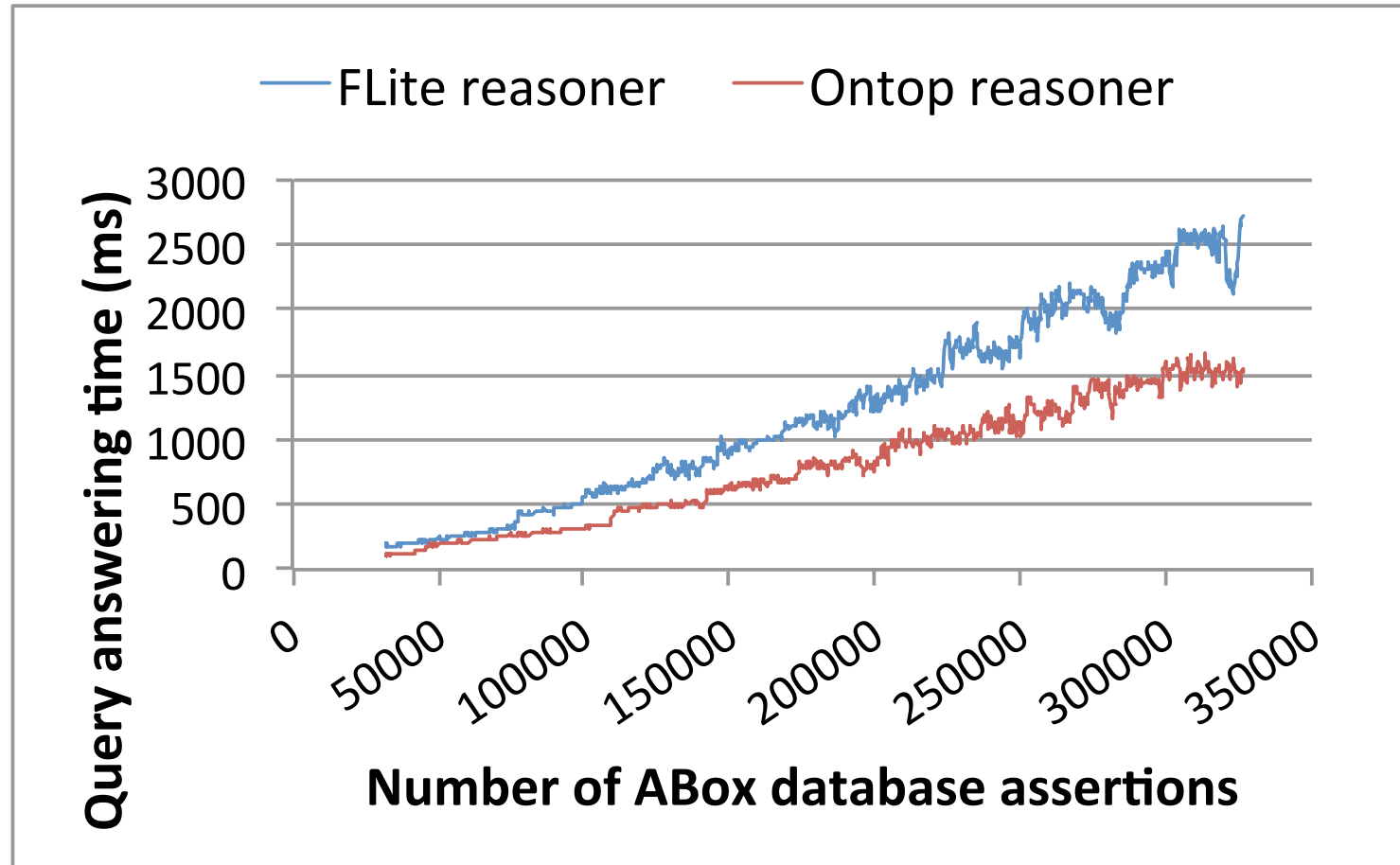
Our implementation: **FLite**

- implements the blackbox rewriting approach for Gödel semantics
- employs **Ontop** to compute the crisp rewriting of the query
- stores fuzzy ABox in database

Empirical test:

- **Test ontology:** TBox with 311 axioms (178 concepts, 39 roles)
ABox (/database) with 10 tables (4 fuzzy ones)
- **Test queries:** fuzzy conjunctive queries
(with 13 atoms, 9 fuzzy ones \rightsquigarrow 9 degree variables)
- **Test set-up:** measure running times of FLite and of **Ontop**

Performance measurement



Temporal fuzzy queries over fuzzy DL-Lite ontologies

Recall:

- temporal operators in the TBox cause **undecidability**
- temporal information given by **sequence of (fuzzy) ABoxes**
- navigate “backwards” on the sequence

Temporal queries use the **operators from LTL**:

\square^- ‘always in the past’

\diamond^- ‘eventually in the past’

\bigcirc^- ‘previous’

Temporal query: query “wrapped” in LTL operators $\bigcirc^- \square^- \langle q(\vec{x}), d \rangle$

FLite and QuAnTOn combined

Fuzzy ABox



Context TBox



$\square - \langle q(\vec{x}), d \rangle$

Fuzzy Query

Temporal fuzzy Reasoner

Query rewriting by **Ontop**

Query rewriting to treat **fuzzy degrees**

Query rewriting to treat **LTl operators**

Query evaluation by **SQL query engine**

$q_{\mathcal{T}}(\vec{x})$

$q_{\mathcal{T},f}(\vec{x}, \vec{y})$

$q_{\mathcal{T},f}(\vec{x}, \vec{y})$

Answers to

$\square - \langle q(\vec{x}), d \rangle$

Conclusions

- introduced a decidable variant of temporal fuzzy query answering
- proposed an pragmatic approach to temporal fuzzy query answering that enriches crisp query rewritings for DL-Lite_R
- method implemented on top of OnTop reasoner
- first evaluation suggest: acceptable overhead for fuzzy information
Temporal fuzzy QA?