

Resolving Temporal Conflicts in Inconsistent RDF Knowledge Bases

Maximilian Dylla Mauro Sozio Martin Theobald

Max Planck Institute for Informatics, Saarbrücken, Germany

Motivation

http://en.wikipedia.org/wiki/David_Beckham

David Robert Joseph Beckham, ^[2]OBE (born 2 May 1975)^[3] is an English footballer who plays **midfield** for **Los Angeles Galaxy** in **Major League Soccer**,^[4] having previously played for **Manchester United**, **Preston North End**, **Real Madrid**, and **A.C. Milan**, as well as the **England national team**, for whom he holds the all-time appearance record for an outfield player.^[5]

<http://marriage.about.com/od/sports/a/davidbeckham.htm>

Victoria and David have three sons.

- **Brooklyn Joseph Beckham**: Born in 1999 in London, England. and his godmother is Elizabeth Hurley.
- **Romeo James Beckham**: Born in 2002 in London, England. His godmother is Elizabeth Hurley.
- **Cruz David Beckham**: Born in 2005 in Madrid, Spain.



Extracting Facts

$Facts \subset (Relation \times Entities \times Entities)$

Weight : $Facts \rightarrow \mathbb{R}^+$

Time-Interval : $Facts \rightarrow Intervals$

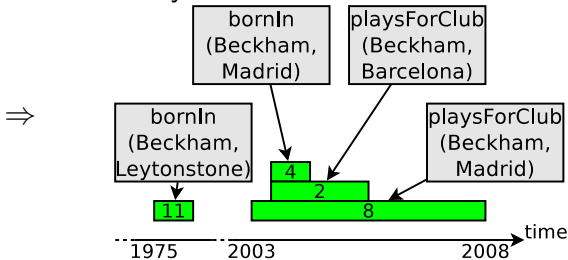
Sources

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Victoria and David have three sons.

- Brooklyn Joseph Beckham: Born in 1999 in London, England, and his godmother is Elizabeth Hurley.
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Noisy Facts



Constraints

Temporal Constraints:

- Precedence (*before*)
- Non-overlapping (*disjoint*)

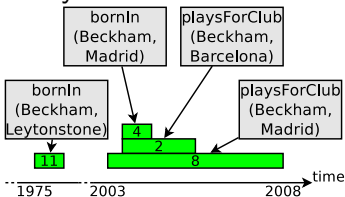
Non-temporal Constraint:

- Functional (*mutEx*)

Each Constraint:

- 2 relations from DB
e.g. *bornIn*
- Both **share a variable**

Noisy Facts



Constraints

$$\left(\begin{array}{l} \text{bornIn}(p, l, t_1) \wedge \\ \text{playsForClub}(p, c, t_2) \end{array} \right) \rightarrow \text{before}(t_1, t_2)$$

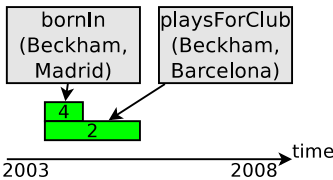
$$\left(\begin{array}{l} \text{playsForClub}(p, c_1, t_1) \wedge \\ \text{playsForClub}(p, c_2, t_2) \wedge \\ c_1 \neq c_2 \end{array} \right) \rightarrow \text{disjoint}(t_1, t_2)$$

Answering Queries

Query

playsForClub(David_Beckham, ?, ?)

Facts



Constraint

$$\left(\begin{array}{l} \text{bornIn}(p, l, t_1) \wedge \\ \text{playsForClub}(p, c, t_2) \end{array} \right) \rightarrow \text{before}(t_1, t_2)$$

Goal: Return only consistent Facts

- 1 Obtain consistent $F \subseteq Facts$

$$\max_{F \subseteq Facts} \sum_{f \in F} w(f)$$

where F fullfills constraints

- 2 Answer query within F

- NP-hard

First Approach

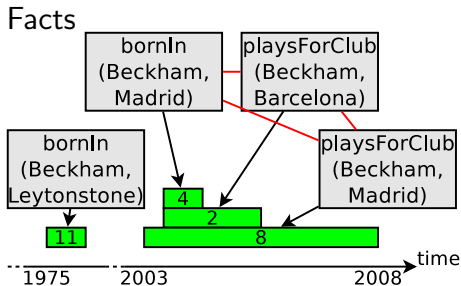
Maximum Weight Independent Set

- Binary constraints
- NP-hard
- Heuristics in $\Omega(|Facts|^2)$

Constraints

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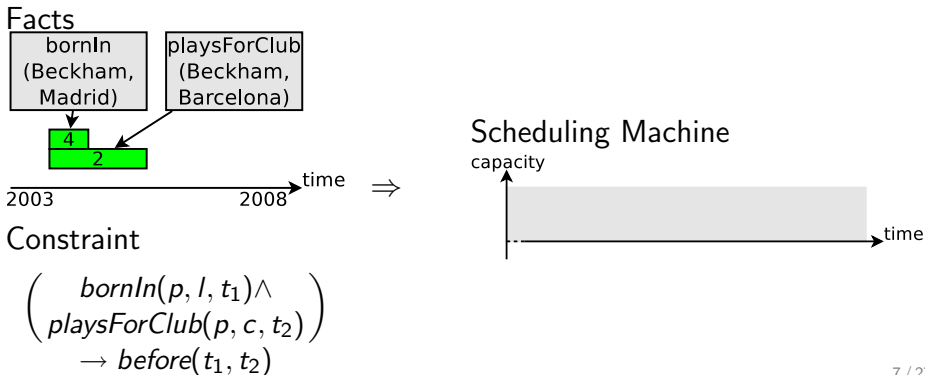
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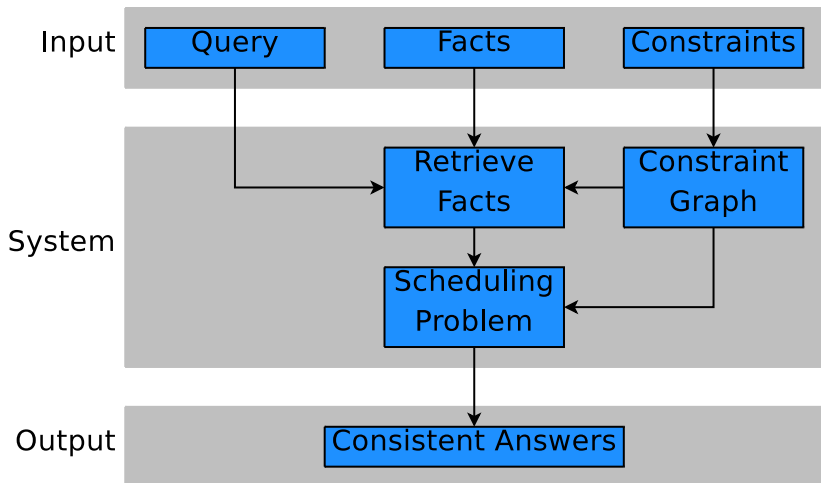
Idea

Scheduling Problem

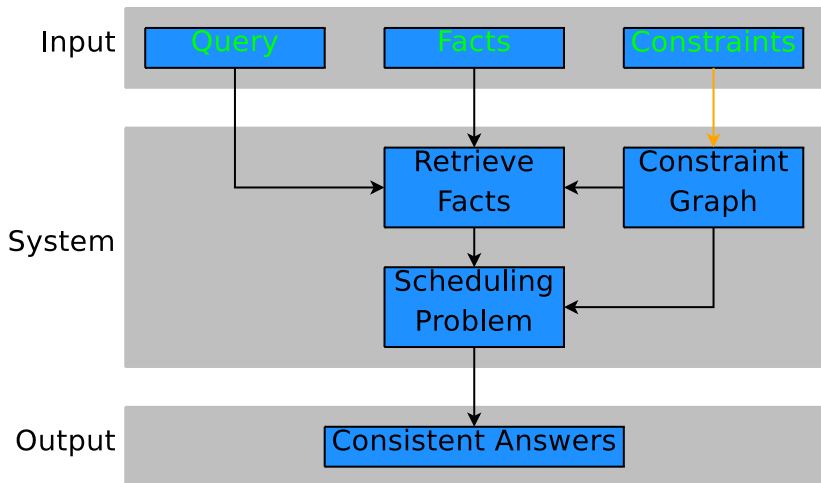
- Temporal constraints
- NP-hard
- Heuristics in $O(|Facts| \log |Facts|)$



Overview



Overview



Mapping

- *Relations* \rightarrow *Vertices*

- *Constraints* \rightarrow *Edges*

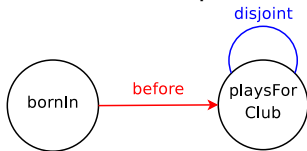
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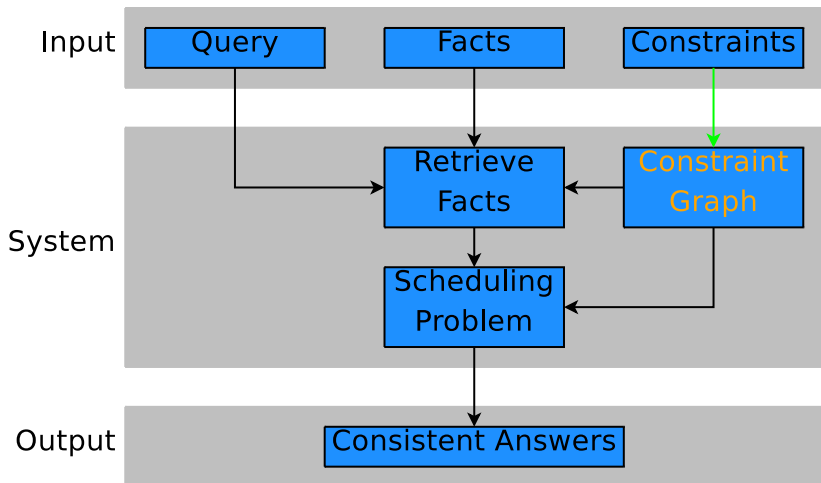
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\Rightarrow

Constraint Graph



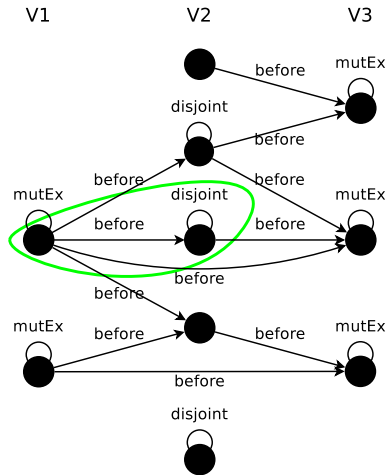
Overview



Constraint Graph

Constraints handled by Scheduling

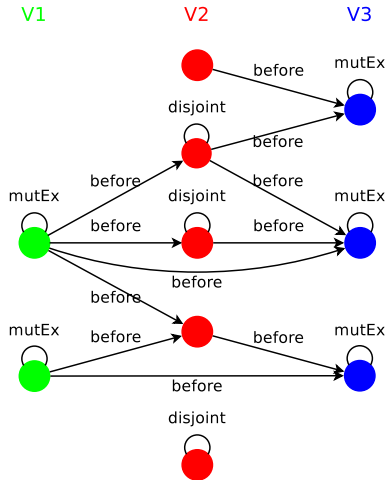
- Tripartite graph $(V_1 \cup V_2 \cup V_3, E)$
- $V_1 \cup V_3$ must have *mutEx* loops
- V_2 can have *disjoint* loops
- *before* can edges point:
 - from V_1 to $V_2 \cup V_3$
 - or from V_2 to V_3



Constraint Graph

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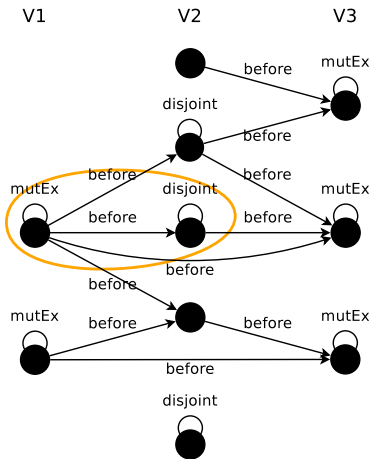
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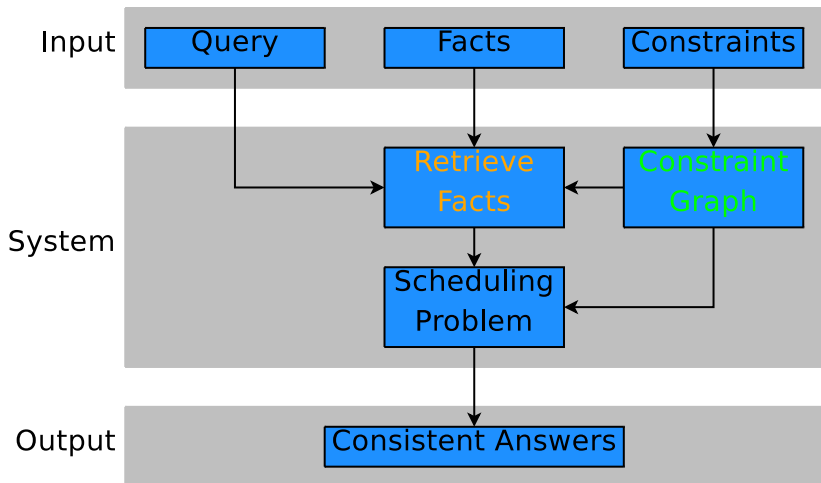
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Overview



Retrieving Facts

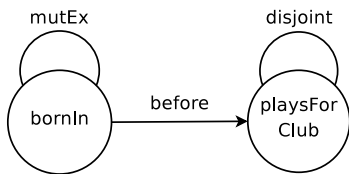
Query

playsForClub(David_Beckham, ?, ?)

Breadth-First Search

- 1 Start: Nodes matching Query
- 2 At each node:
 - 1 Retrieve Facts from DB
 - 2 If result $\neq \emptyset$:
continue at not visited neighbours, pass argument

Constraint Graph



In General:

- For each shared Argument
 $\Rightarrow O(|Facts| \cdot |Constraints|)$

Retrieving Facts

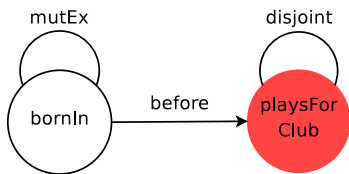
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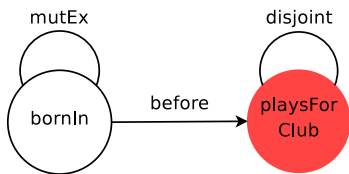
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Query

playsForClub(David_Beckham, ?, ?)

Constraint Graph



Select * from facts where
rel=playsForClub and
arg1=David_Beckham;

Retrieving Facts

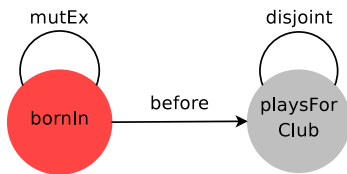
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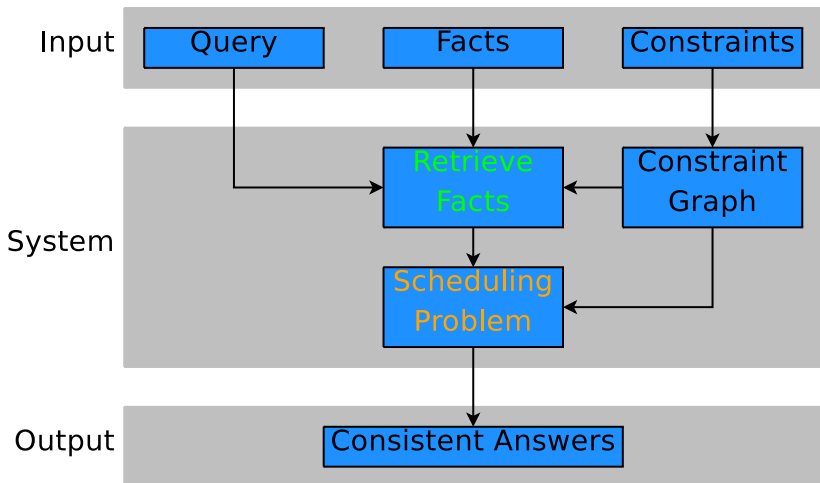
Constraint Graph



In General:

- For each shared Argument
 $\Rightarrow O(|Facts| \cdot |Constraints|)$

Overview



Facts \rightarrow Scheduling Jobs

- Weight: Identical
- Capacity $\in [0, 1]$
- Begin: Identical or 0
- End: Identical or ∞

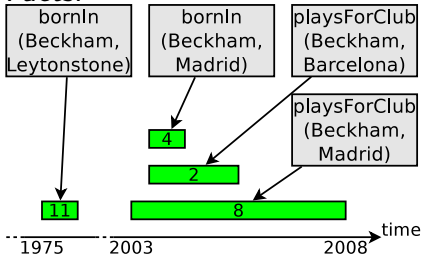
Depending on constraint graph

In General:

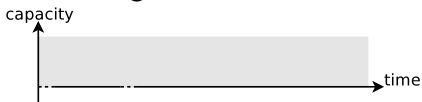
- Several Scheduling Machines
- $O(|F| \log |F| + |F| |machines|)$
[Bar-Noy et al., 2001]

Scheduling

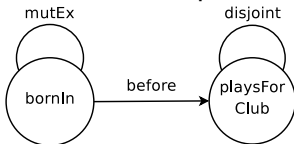
Facts:



Scheduling Machine:



Constraint Graph:



Scheduling

Facts \rightarrow Scheduling Jobs

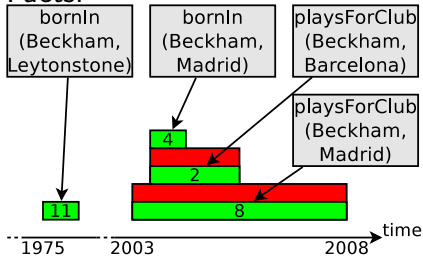
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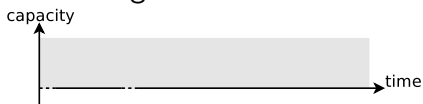
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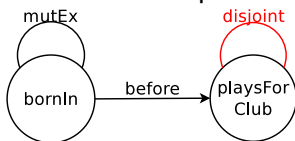
Facts:



Scheduling Machine:



Constraint Graph:



Scheduling

Facts \rightarrow Scheduling Jobs

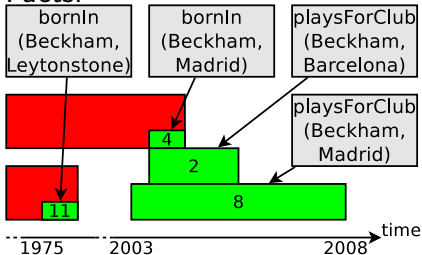
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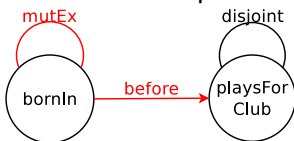
Facts:



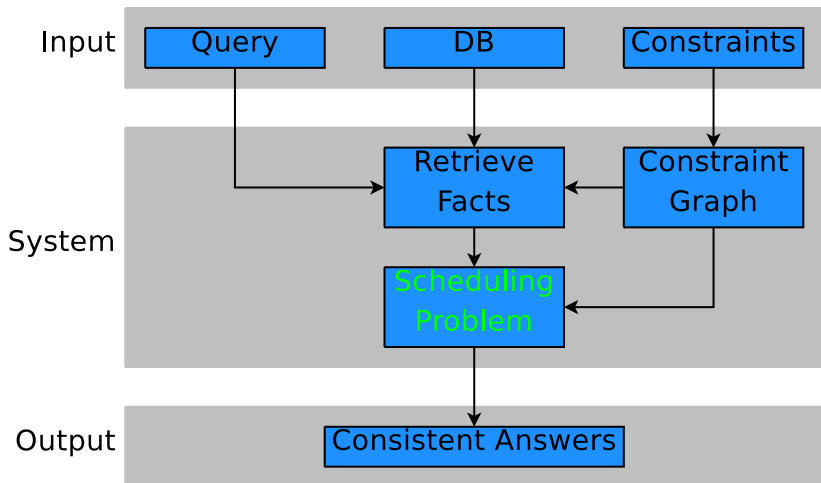
Scheduling Machine:



Constraint Graph:

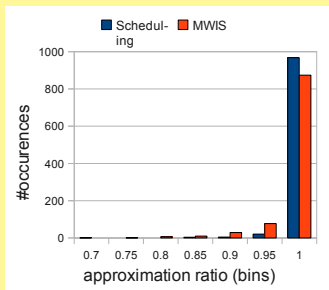


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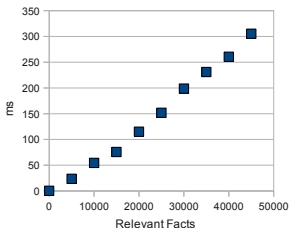


Experiments

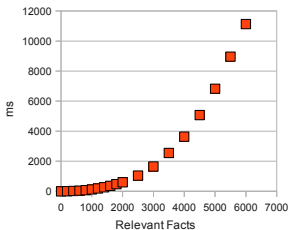
- 4 correct facts, 20 noisy facts
- Approximation ratio : $\frac{W_{heuristic}}{W_{optimal}}$



Runtime Scheduling



Runtime MWIS



The End

Conclusions

- Resolve conflicts by Scheduling
- Runtime in $O(n \log n)$

Future Work

- Generalize Constraints
- Histograms instead of Intervals
- Probabilistic Reasoning



Bar-Noy, A., Bar-Yehuda, R., Freund, A., (Seffi) Naor, J., and Schieber, B. (2001).

A unified approach to approximating resource allocation and scheduling.

J. ACM, 48(5):1069–1090.