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Existential Rules – Lecture 3

Adapted from slides by Andreas Pieris and Michaël Thomazo

Syntax of Existential Rules

An **existential rule** is an expression

$$\forall \mathbf{X} \forall \mathbf{Y} (\underbrace{\varphi(\mathbf{X}, \mathbf{Y})}_{\text{body}} \rightarrow \exists \mathbf{Z} \underbrace{\psi(\mathbf{X}, \mathbf{Z})}_{\text{head}})$$

- \mathbf{X}, \mathbf{Y} and \mathbf{Z} are tuples of variables of \mathbf{V}
- $\varphi(\mathbf{X}, \mathbf{Y})$ and $\psi(\mathbf{X}, \mathbf{Z})$ are (constant-free) conjunctions of atoms

...a.k.a. tuple-generating dependencies, and Datalog[±] rules



Semantics of Existential Rules

- An instance J is a **model** of the rule

$$\sigma = \forall \mathbf{X} \forall \mathbf{Y} (\varphi(\mathbf{X}, \mathbf{Y}) \rightarrow \exists \mathbf{Z} \psi(\mathbf{X}, \mathbf{Z}))$$

written as $J \models \sigma$, if the following holds:

whenever there exists a homomorphism h such that $h(\varphi(\mathbf{X}, \mathbf{Y})) \subseteq J$,

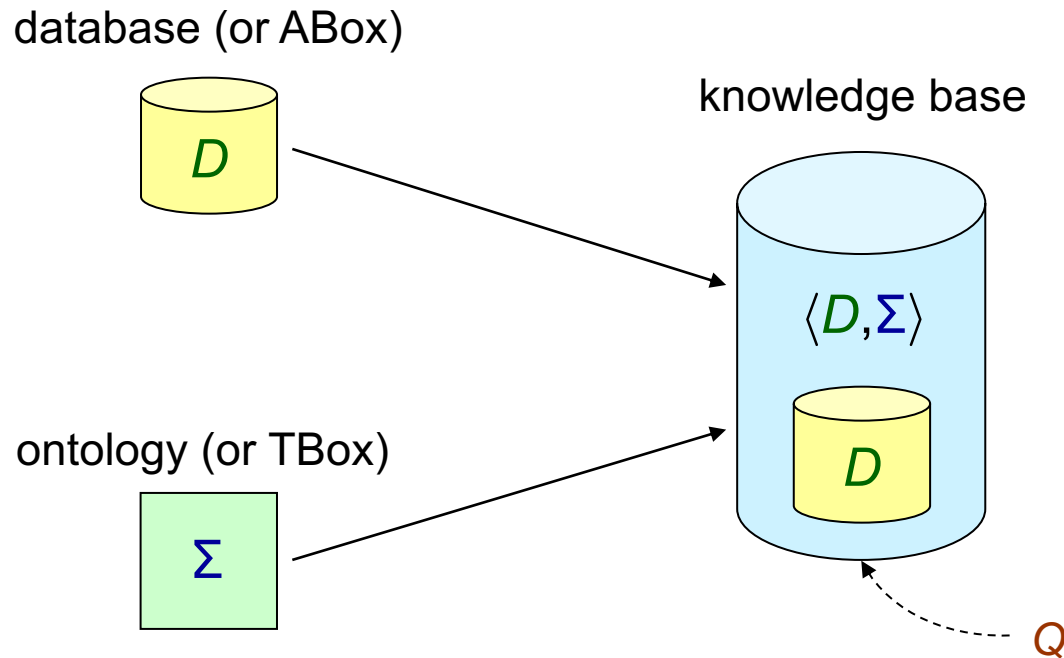
then there exists $g \supseteq h|_{\mathbf{X}}$ such that $g(\psi(\mathbf{X}, \mathbf{Z})) \subseteq J$

$\{t \mapsto h(t) \mid t \in \mathbf{X}\}$ – the **restriction** of h to \mathbf{X}

- Given a set Σ of existential rules, J is a **model** of Σ , written as $J \models \Sigma$, if the following holds: for each $\sigma \in \Sigma$, $J \models \sigma$
- It can be shown that $J \models \Sigma$ iff J is a model of the first-order theory $\bigwedge_{\sigma \in \Sigma} \sigma$



Ontology-Based Query Answering (OBQA)



existential rules

$$\forall X \forall Y (\varphi(X, Y) \rightarrow \exists Z \psi(X, Z))$$

Syntax of Conjunctive Queries

A **conjunctive query (CQ)** is an expression

$$\exists \mathbf{Y} (\varphi(\mathbf{X}, \mathbf{Y}))$$

- \mathbf{X} and \mathbf{Y} are tuples of variables of \mathbf{V}
- $\varphi(\mathbf{X}, \mathbf{Y})$ is a conjunction of atoms (possibly with constants)

The most important query language used in practice

Forms the **SELECT-FROM-WHERE** fragment of SQL

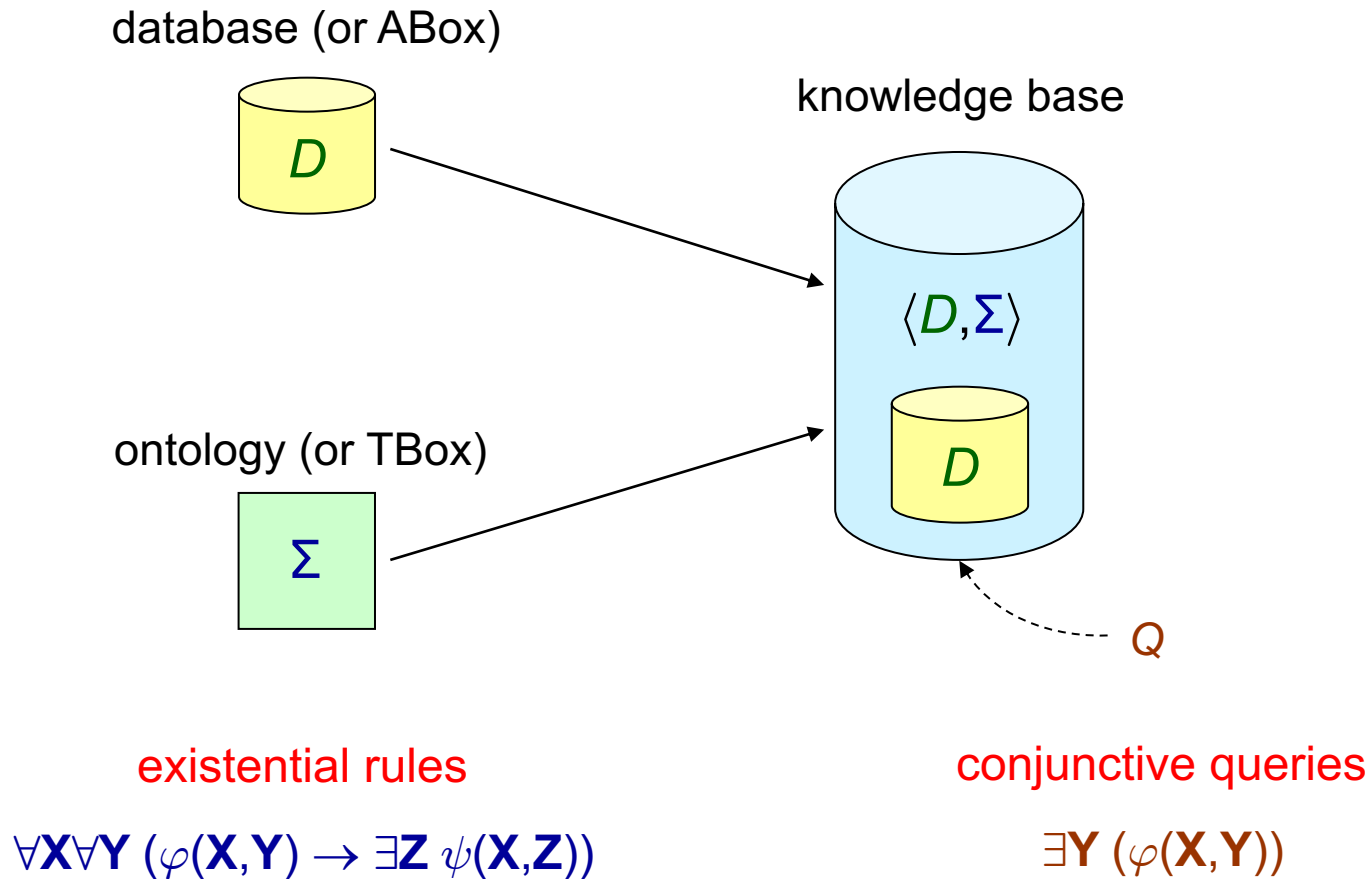


Semantics of Conjunctive Queries

- A **match** of a CQ $\exists \mathbf{Y} (\varphi(\mathbf{X}, \mathbf{Y}))$ in an instance J is a homomorphism h such that $h(\varphi(\mathbf{X}, \mathbf{Y})) \subseteq J$ i.e., all the atoms of the query are satisfied
- The **answer** to $Q = \exists \mathbf{Y} (\varphi(\mathbf{X}, \mathbf{Y}))$ over J is the set of tuples
$$Q(J) = \{h(\mathbf{X}) \mid h \text{ is a match of } Q \text{ in } J\}$$
- The answer consists of the witnesses for the **free variables** of the query



Ontology-Based Query Answering (OBQA)



OBQA: Formal Definition

active domain – constants occurring in D

CQ-Answering:

Input: database D , existential rules Σ , CQ $Q = \exists \mathbf{Y} (\varphi(\mathbf{X}, \mathbf{Y}))$, tuple $\mathbf{t} \in \text{adom}(D)^{|\mathbf{X}|}$

Question: decide whether $\mathbf{t} \in \text{certain}(Q, \langle D, \Sigma \rangle) = \bigcap_{J \in \text{models}(D \wedge \Sigma)} Q(J)_{\downarrow}$

$$\mathbf{t} \in \text{certain}(Q, \langle D, \Sigma \rangle) \quad \text{iff} \quad \mathbf{t} \in \bigcap_{J \in \text{models}(D \wedge \Sigma)} Q(J)_{\downarrow}$$

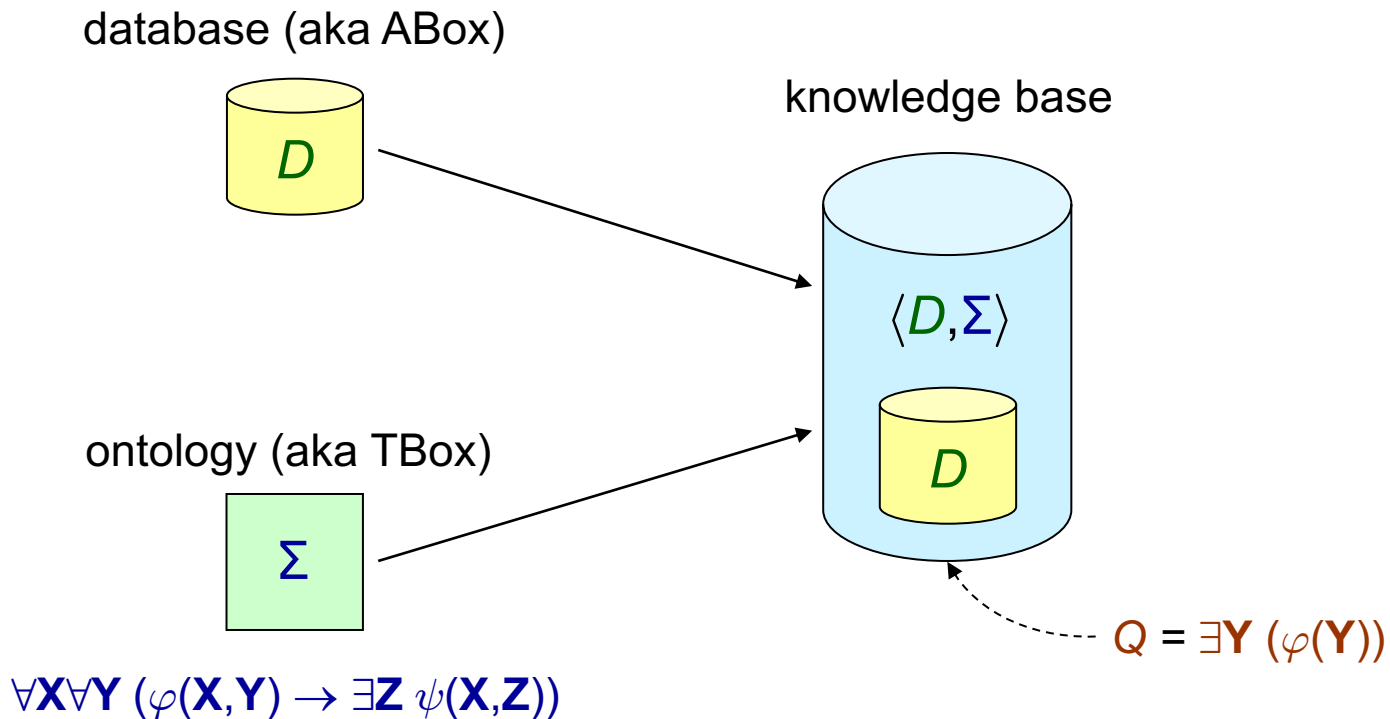
$$\text{iff} \quad \forall J \in \text{models}(D \wedge \Sigma), J \models \exists \mathbf{Y} (\varphi(\mathbf{t}, \mathbf{Y}))$$

$$\text{iff} \quad D \wedge \Sigma \models \exists \mathbf{Y} (\varphi(\mathbf{t}, \mathbf{Y}))$$

Boolean CQ (BCQ) – no free variables

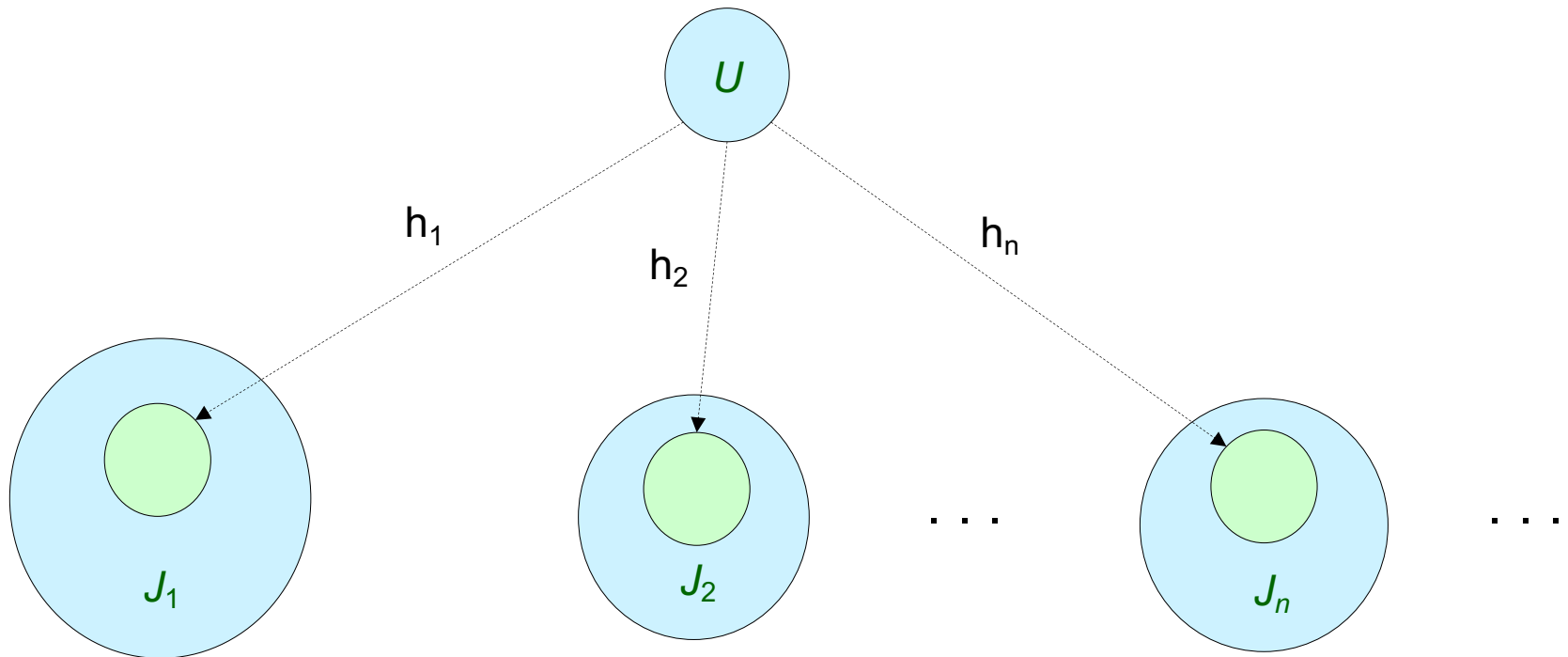


BCQ-Answering: Our Main Decision Problem



decide whether $D \wedge \Sigma \models Q$

Universal Models (a.k.a. Canonical Models)



An instance U is a **universal model** of $D \wedge \Sigma$ if the following holds:

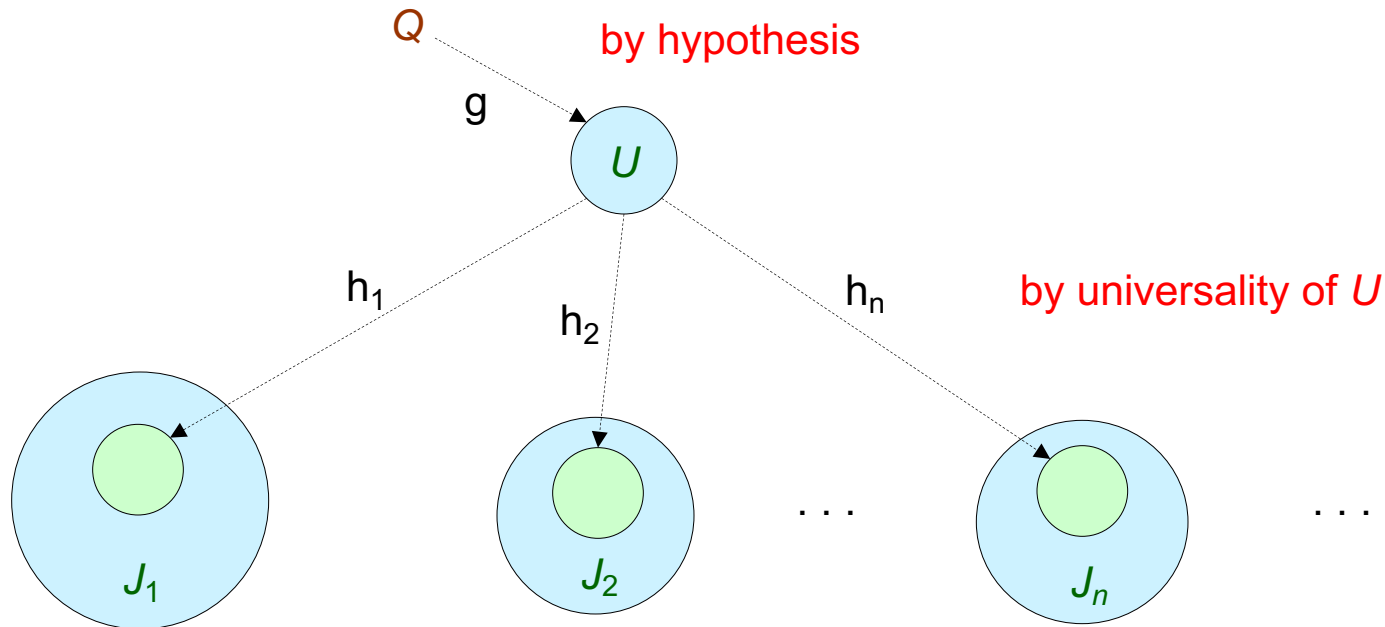
1. U is a model of $D \wedge \Sigma$
2. $\forall J \in \text{models}(D \wedge \Sigma)$, there exists a homomorphism h_J such that $h_J(U) \subseteq J$

Query Answering via Universal Models

Theorem: $D \wedge \Sigma \models Q$ iff $U \models Q$, where U is a universal model of $D \wedge \Sigma$

Proof: (\Rightarrow) Trivial since, for every $J \in \text{models}(D \wedge \Sigma)$, $J \models Q$

(\Leftarrow) By exploiting the universality of U



$$\begin{aligned} \forall J \in \text{models}(D \wedge \Sigma), \exists h_J \text{ such that } h_J(g(Q)) \subseteq J &\Rightarrow \forall J \in \text{models}(D \wedge \Sigma), J \models Q \\ &\Rightarrow D \wedge \Sigma \models Q \end{aligned}$$

The Chase Procedure

- **Fundamental algorithmic tool** used in databases
- It has been applied to a **wide range of problems**:
 - Checking containment of queries under constraints
 - Computing data exchange solutions
 - Computing certain answers in data integration settings
 - ...

... what's the reason for the ubiquity of the chase in databases?

it constructs universal models



The Chase Procedure

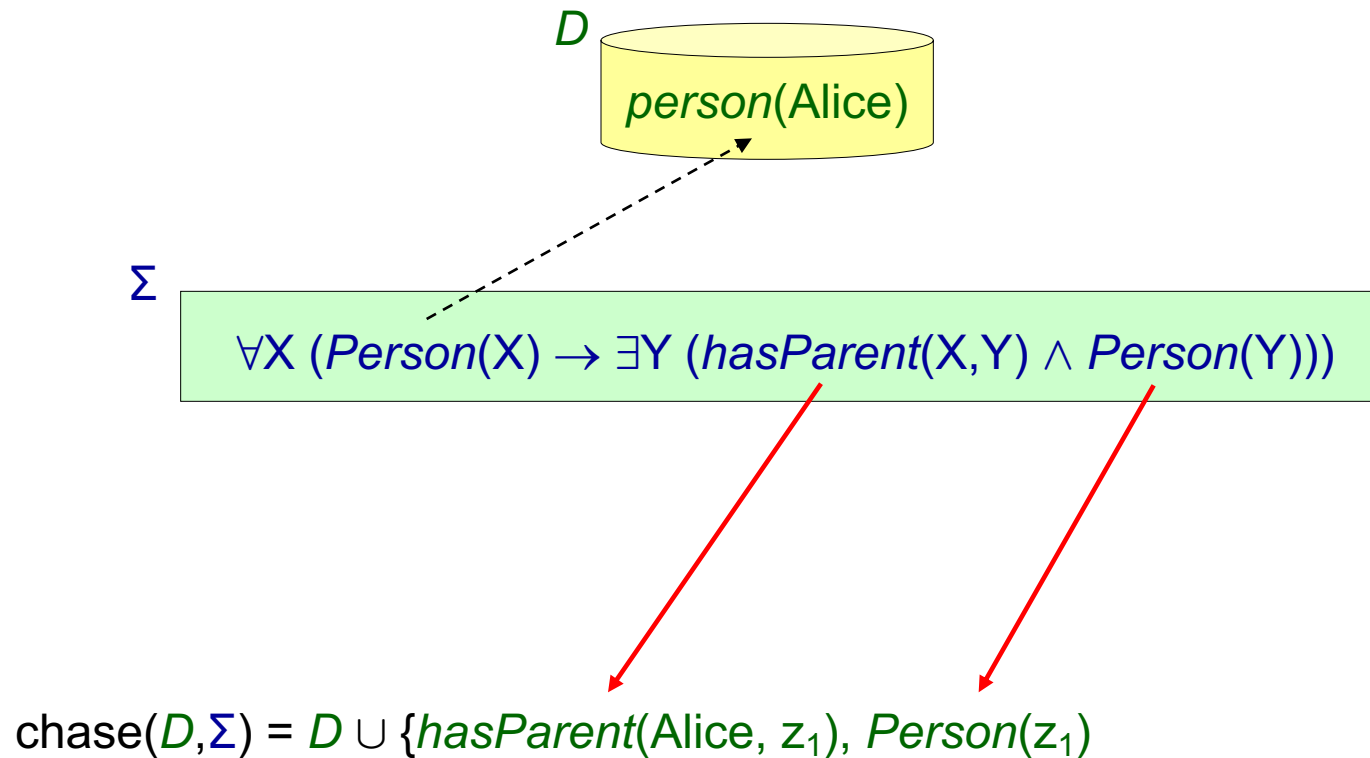


Σ

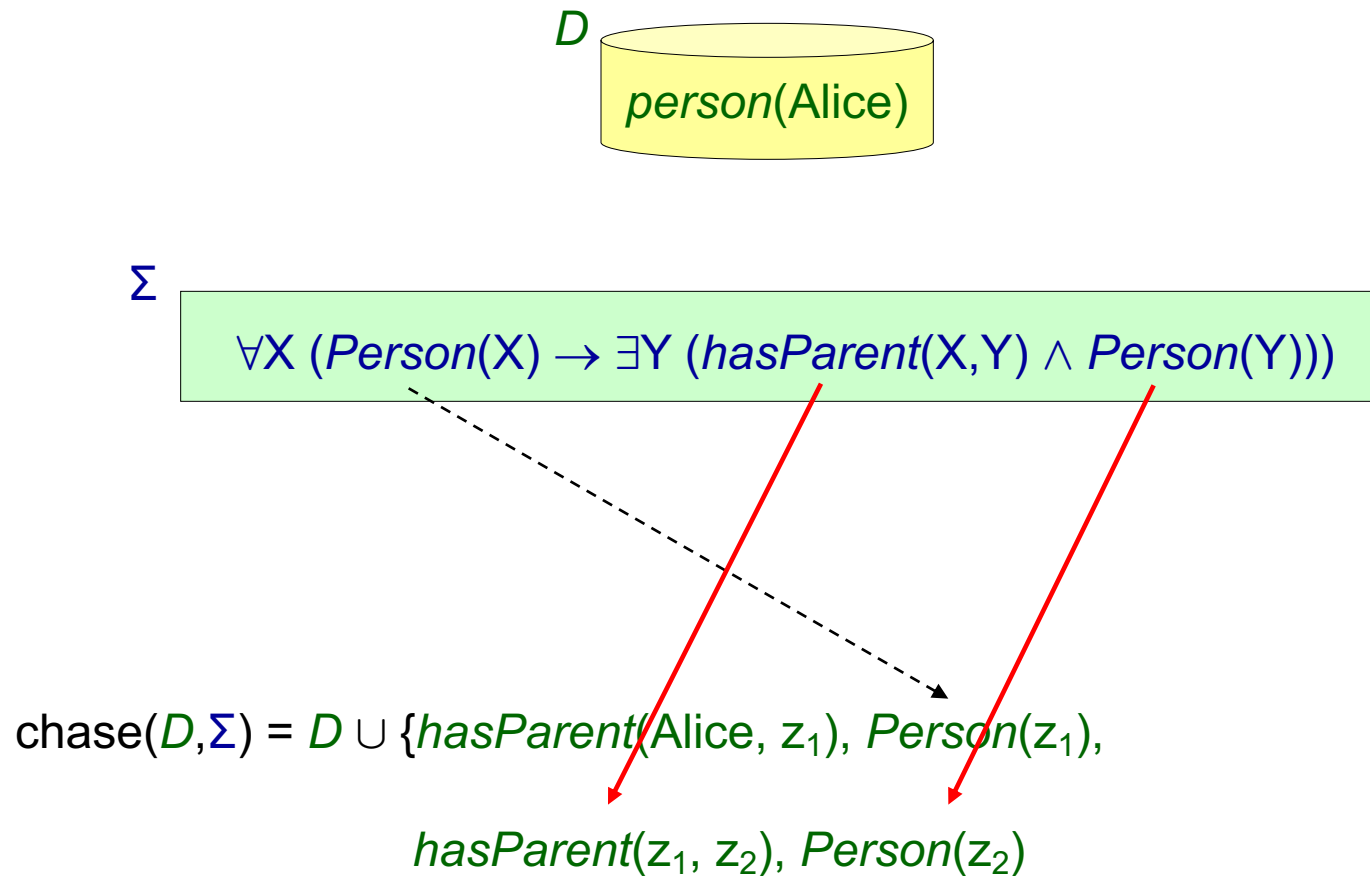
$\forall X (Person(X) \rightarrow \exists Y (hasParent(X,Y) \wedge Person(Y)))$

$chase(D, \Sigma) = D \cup$

The Chase Procedure



The Chase Procedure



The Chase Procedure



Σ

$\forall X (Person(X) \rightarrow \exists Y (hasParent(X,Y) \wedge Person(Y)))$

$chase(D, \Sigma) = D \cup \{hasParent(Alice, z_1), Person(z_1),$

$hasParent(z_1, z_2), Person(z_2),$

$hasParent(z_2, z_3), Person(z_3)\}$



The Chase Procedure



Σ

$\forall X (Person(X) \rightarrow \exists Y (hasParent(X, Y) \wedge Person(Y)))$

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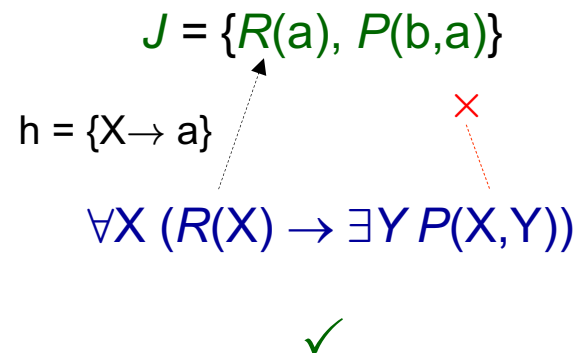
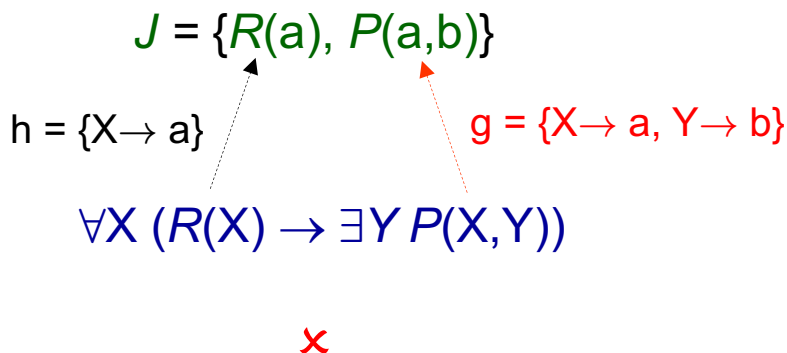
$hasParent(z_2, z_3), Person(z_3), \dots$

infinite instance



The Chase Procedure: Formal Definition

- **Chase rule** - the building block of the chase procedure
- A rule $\sigma = \forall \mathbf{X} \forall \mathbf{Y} (\varphi(\mathbf{X}, \mathbf{Y}) \rightarrow \exists \mathbf{Z} \psi(\mathbf{X}, \mathbf{Z}))$ is **applicable** to instance J if:
 1. There exists a homomorphism h such that $h(\varphi(\mathbf{X}, \mathbf{Y})) \subseteq J$
 2. There is no $g \supseteq h_{|X}$ such that $g(\psi(\mathbf{X}, \mathbf{Z})) \subseteq J$



The Chase Procedure: Formal Definition

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- A rule $\sigma = \forall \mathbf{X} \forall \mathbf{Y} (\varphi(\mathbf{X}, \mathbf{Y}) \rightarrow \exists \mathbf{Z} \psi(\mathbf{X}, \mathbf{Z}))$ is **applicable** to instance J if:
 1. There exists a homomorphism h such that $h(\varphi(\mathbf{X}, \mathbf{Y})) \subseteq J$
 2. There is no $g \supseteq h|_{\mathbf{X}}$ such that $g(\psi(\mathbf{X}, \mathbf{Z})) \subseteq J$
- Let $J_+ = J \cup \{g(\psi(\mathbf{X}, \mathbf{Z}))\}$, where $g \supseteq h|_{\mathbf{X}}$ and $g(\mathbf{Z})$ are “fresh” nulls not in J
- The result of applying σ to J is J_+ , denoted $J \langle \sigma, h \rangle J_+$ - **single chase step**



The Chase Procedure: Formal Definition

- A **finite chase** of D w.r.t. Σ is a finite sequence

$$D \langle \sigma_1, h_1 \rangle J_1 \langle \sigma_2, h_2 \rangle J_2 \langle \sigma_3, h_3 \rangle J_3 \dots \langle \sigma_n, h_n \rangle J_n$$

where no rule from Σ is applicable in J_n .

Then, $\text{chase}(D, \Sigma)$ is defined as the instance J_n

all applicable rules will eventually be applied

- An **infinite chase** of D w.r.t. Σ is a **fair** infinite sequence

$$D \langle \sigma_1, h_1 \rangle J_1 \langle \sigma_2, h_2 \rangle J_2 \langle \sigma_3, h_3 \rangle J_3 \dots \langle \sigma_n, h_n \rangle J_n \dots$$

and $\text{chase}(D, \Sigma)$ is **defined** as the instance $\bigcup_{k \geq 0} J_k$ (with $J_0 = D$)

least fixpoint of a monotonic operator - chase step



Chase: A Universal Model

Theorem: $\text{chase}(D, \Sigma)$ is a universal model of $D \wedge \Sigma$

the result of the chase after k applications of the chase step

Proof:

- By construction, $\text{chase}(D, \Sigma) \in \text{models}(D \wedge \Sigma)$
- It remains to show that $\text{chase}(D, \Sigma)$ can be homomorphically embedded into every other model of $D \wedge \Sigma$
- Fix an arbitrary instance $J \in \text{models}(D \wedge \Sigma)$. We need to show that there exists h such that $h(\text{chase}(D, \Sigma)) \subseteq J$
- By induction on the number of applications of the chase step, we show that for every $k \geq 0$, there exists h_k such that $h_k(\text{chase}^{[k]}(D, \Sigma)) \subseteq J$, and h_k is compatible with h_{k-1}
- Clearly, $\cup_{k \geq 0} h_k$ is a well-defined homomorphism that maps $\text{chase}(D, \Sigma)$ to J
- The claim follows with $h = \cup_{k \geq 0} h_k$



Chase: Uniqueness Property

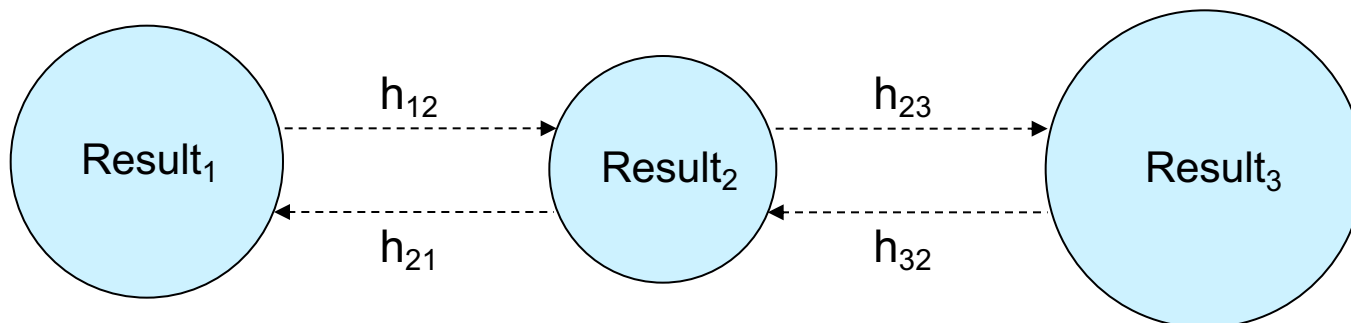
- The result of the chase is **not unique** - depends on the order of rule application

$$D = \{P(a)\} \quad \sigma_1 = \forall X (P(X) \rightarrow \exists Y R(Y)) \quad \sigma_2 = \forall X (P(X) \rightarrow R(X))$$

$$\text{Result}_1 = \{P(a), R(z), R(a)\} \quad \sigma_1 \text{ then } \sigma_2$$

$$\text{Result}_2 = \{P(a), R(a)\} \quad \sigma_2 \text{ then } \sigma_1$$

- But, it is **unique up to homomorphic equivalence**



- Thus, it is **unique** for query answering purposes



Query Answering via the Chase

Theorem: $D \wedge \Sigma \models Q$ iff $U \models Q$, where U is a universal model of $D \wedge \Sigma$

+

Theorem: $\text{chase}(D, \Sigma)$ is a universal model of $D \wedge \Sigma$

=

Corollary: $D \wedge \Sigma \models Q$ iff $\text{chase}(D, \Sigma) \models Q$

- We can tame the first dimension of infinity by exploiting the chase procedure
- But, **what about the second dimension of infinity?** - the chase may be infinite



Rest of the Lecture

- Undecidability of BCQ-Answering
- Gaining decidability - terminating chase
- Full Existential Rules
- Acyclic Existential Rules



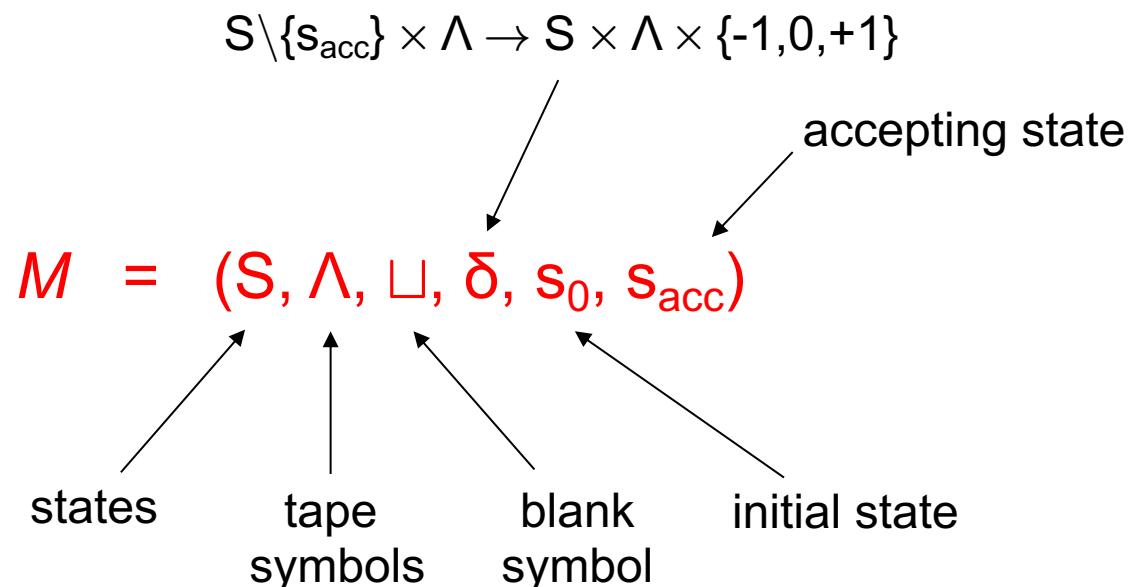
Undecidability of BCQ-Answering

Theorem: BCQ-Answering is **undecidable**

Proof : By simulating a deterministic Turing machine with an empty tape



Deterministic Turing Machine (DTM)



$$\delta(s_1, \alpha) = (s_2, \beta, +1)$$

IF at some time instant τ the machine is in state s_1 , the cursor points to cell κ , and this cell contains α

THEN at instant $\tau+1$ the machine is in state s_2 , cell κ contains β , and the cursor points to cell $\kappa+1$



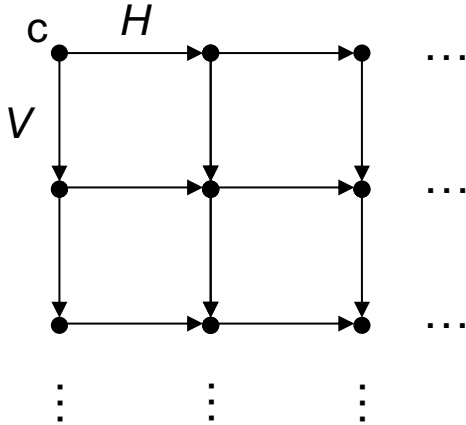
Undecidability of BCQ-Answering

Our Goal: Encode the computation of a DTM M with an empty tape using a database D , a set Σ of existential rules, and a BCQ Q such that

$$D \wedge \Sigma \models Q \text{ iff } M \text{ accepts}$$



Build an Infinite Grid



k -th horizontal line represents the
 k -th configuration of the machine

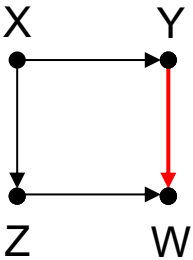
$$\forall X (Start(X) \rightarrow Node(X) \wedge Initial(X))$$

$$D = \{Start(c)\}$$

fixes the origin of the grid

$$\forall X (Node(X) \rightarrow \exists Y (H(X,Y) \wedge Node(Y)))$$

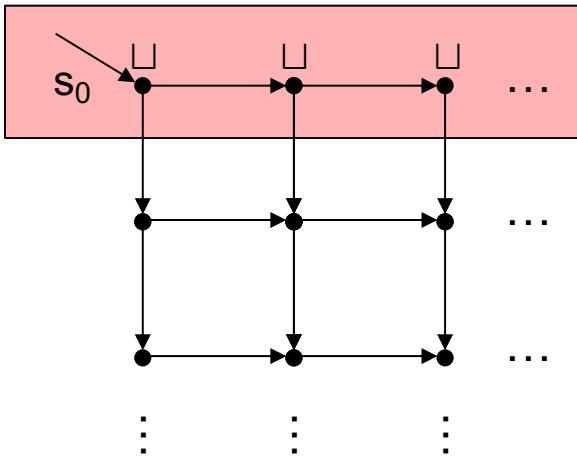
$$\forall X (Node(X) \rightarrow \exists Y (V(X,Y) \wedge Node(Y)))$$



$$\forall X \forall Y \forall Z \forall W (H(X,Y) \wedge H(Z,W) \wedge V(X,Z) \rightarrow V(Y,W))$$



Initialization Rules

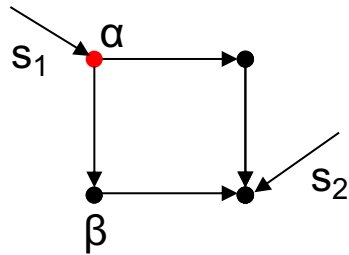


$$\forall X \forall Y (Initial(X) \wedge H(X, Y) \rightarrow Initial(Y))$$

$$\forall X (Start(X) \rightarrow Cursor[s_0](X))$$

$$\forall X (Initial(X) \rightarrow Symbol[\square](X))$$

Transition Rules

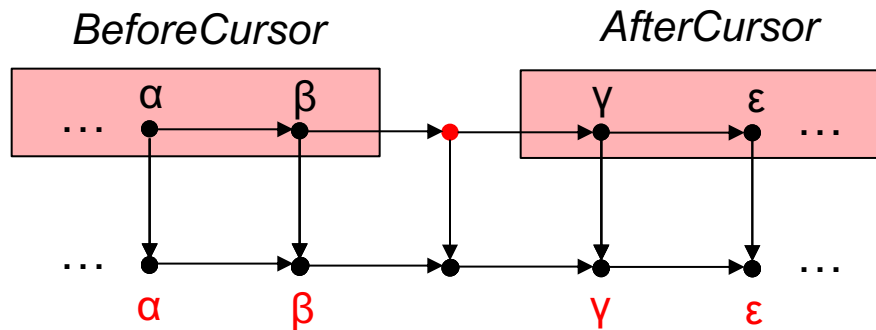


$$\delta(s_1, \alpha) = (s_2, \beta, +1)$$

$\forall X \forall Y \forall Z (Cursor[s_1](X) \wedge Symbol[\alpha](X) \wedge V(X, Y) \wedge H(Y, Z) \rightarrow$

$Cursor[s_2](Z) \wedge Symbol[\beta](Y) \wedge Mark(X))$

Inertia Rules



$$\forall X \forall Y (Mark(X) \wedge H(X, Y) \rightarrow AfterCursor(Y))$$

$$\forall X \forall Y (AfterCursor(X) \wedge H(X, Y) \rightarrow AfterCursor(Y))$$

$$\forall X \forall Y (AfterCursor(X) \wedge Symbol[\alpha](X) \wedge V(X, Y) \rightarrow Symbol[\alpha](Y))$$

...we have similar rules for the **cells before the cursor**

Accepting Rule

Once we reach the accepting state we accept

$$\forall X (\text{Cursor}[s_{\text{acc}}](X) \rightarrow \text{Accept}(X))$$

$D \wedge \Sigma \models \exists X \text{Accept}(X)$ iff the DTM M accepts



Undecidability of BCQ-Answering

Theorem: BCQ-Answering is **undecidable**

Proof : By simulating a deterministic Turing machine with an empty tape

...syntactic restrictions are needed!!!



Gaining Decidability

By restricting the database

- $\{Start(c)\} \wedge \Sigma \models Q$ iff the DTM M accepts
- The problem is undecidable already for singleton databases
- No much to do in this direction

By restricting the query language

- $D \wedge \Sigma \models \exists X Accept(X)$ iff the DTM M accepts
- The problem is undecidable already for atomic queries
- No much to do in this direction

By restricting the ontology language

- Achieve a good trade-off between expressive power and complexity
- Field of intense research
- Any ideas?

... force the chase to terminate



What is the Source of Non-termination?



Σ

$\forall X (Person(X) \rightarrow \exists Y (hasParent(X, Y) \wedge Person(Y)))$

$chase(D, \Sigma) = D \cup \{hasParent(Alice, z_1), Person(z_1),$

$hasParent(z_1, z_2), Person(z_2),$

$hasParent(z_2, z_3), Person(z_3), \dots$

1. Existential quantification
2. Recursive definitions



Termination of the Chase

- Drop the existential quantification
 - We obtain the class of **full** existential rules
 - Very close to Datalog

- Drop the recursive definitions
 - We obtain the class of **acyclic** existential rules
 - A.k.a. non-recursive existential rules



Full Existential Rules

- A **full existential rule** is an existential rule of the form

$$\forall X \forall Y (\varphi(X, Y) \rightarrow \psi(X))$$

- We denote **FULL** the class of full existential rules
- A **local property** - we can inspect one rule at a time
 - ⇒ given Σ , we can decide in linear time whether $\Sigma \in \mathbf{FULL}$
 - ⇒ closed under union - $\Sigma_1 \in \mathbf{FULL}, \Sigma_2 \in \mathbf{FULL} \Rightarrow (\Sigma_1 \cup \Sigma_2) \in \mathbf{FULL}$
- Why does the chase terminate?



Full Existential Rules

- Consider a database D and a set $\Sigma \in \text{FULL}$
- $\text{chase}(D, \Sigma) \subseteq \{P(c_1, \dots, c_n) \mid \langle c_1, \dots, c_n \rangle \in \text{adom}(D)^n \text{ and } P \in \text{sch}(\Sigma)\}$

active domain - constants occurring in D

schema - predicates occurring in Σ

maximum number of tuples
with terms of $\text{adom}(D)$

$$|\text{chase}(D, \Sigma)| \leq |\text{sch}(\Sigma)| \cdot (|\text{adom}(D)|)^{\text{maxarity}}$$

$$\text{maxarity} = \max_{P \in \text{sch}(\Sigma)} \{\text{arity}(P)\}$$

maximum number of atoms with predicates of
 $\text{sch}(\Sigma)$ and terms of $\text{adom}(D)$

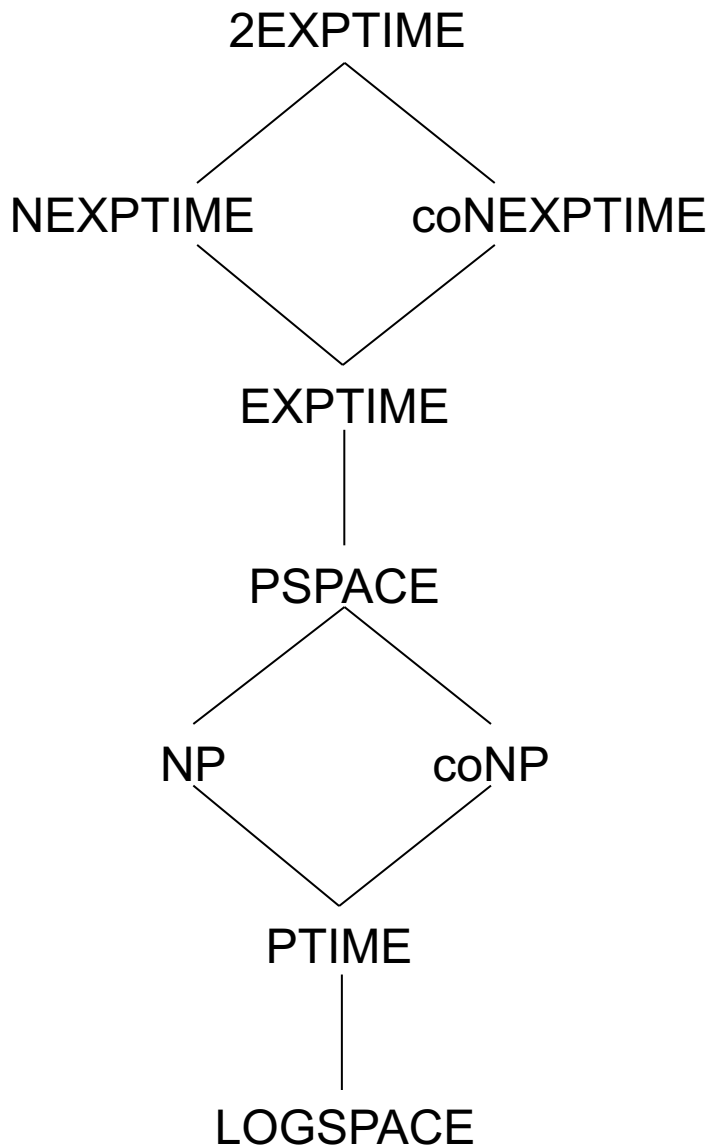


Complexity Measures for Query Answering

- **Data complexity:** is calculated by considering only the database as part of the input, while the ontology and the query are fixed
- **Combined complexity:** is calculated by considering, apart from the database, also the ontology and the query as part of the input
- Data complexity vs. Combined complexity
 - Data complexity tends to be a more meaningful measure - ontologies and queries tend to be small; databases tend to be large
 - Nevertheless, the combined complexity is a relevant measure - identifies the real source of complexity



Some Important Complexity Classes



Problems that can be solved by an algorithm that runs in **double-exponential time**

We need the power of non-determinism

Problems that can be solved by an algorithm that runs in **exponential time**

Problems that can be solved by an algorithm that uses a **polynomial amount of memory**

We need the power of non-determinism

Problems that can be solved by an algorithm that runs in **polynomial time**

Problems that can be solved by an algorithm that uses a **logarithmic amount of memory**



Data Complexity of FULL

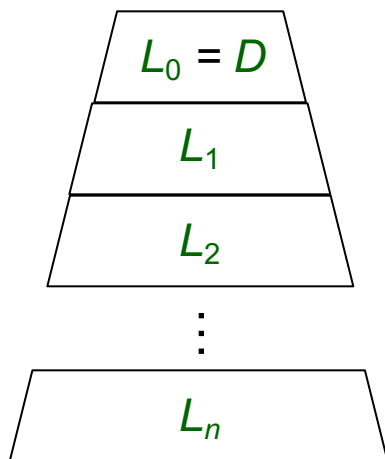
Theorem: BCQ-Answering under FULL is in PTIME w.r.t. the data complexity

Proof: Consider a database D , a set $\Sigma \in \text{FULL}$, and a BCQ Q

We apply the naïve algorithm:

1. Construct $\text{chase}(D, \Sigma)$
2. Check for the existence of a homomorphism h such that $h(Q) \subseteq \text{chase}(D, \Sigma)$

Step 1: We construct the chase level-by-level



- From L_k to L_{k+1} : for each $\sigma \in \Sigma$, find all the homomorphisms h such that $h(\text{body}(\sigma)) \subseteq L_k$, and add to L_k the set of atoms $h(\text{head}(\sigma))$
- Stop when $L_k = L_{k+1}$

$$|\Sigma| \cdot (|\text{adom}(D)|)^{\max \text{variables}(\Sigma)} \cdot \max \text{body}(\Sigma) \cdot |L_k|$$



Data Complexity of FULL

Theorem: BCQ-Answering under FULL is in PTIME w.r.t. the data complexity

Proof: Consider a database D , a set $\Sigma \in \text{FULL}$, and a BCQ Q

We apply the naïve algorithm:

1. Construct $\text{chase}(D, \Sigma)$
2. Check for the existence of a homomorphism h such that $h(Q) \subseteq \text{chase}(D, \Sigma)$

Step 1: We construct the chase level-by-level in time

$$(k-1) \cdot |\Sigma| \cdot (|\text{adom}(D)|)^{\max \text{variables}(\Sigma)} \cdot \max \text{body}(\Sigma) \cdot |L|$$

$$\text{where } k, |L| \leq |\text{chase}(D, \Sigma)| \leq |\text{sch}(\Sigma)| \cdot (|\text{adom}(D)|)^{\max \text{arity}}$$



Data Complexity of FULL

Theorem: BCQ-Answering under FULL is in PTIME w.r.t. the data complexity

Proof: Consider a database D , a set $\Sigma \in \text{FULL}$, and a BCQ Q

We apply the naïve algorithm:

1. Construct $\text{chase}(D, \Sigma)$
2. Check for the existence of a homomorphism h such that $h(Q) \subseteq \text{chase}(D, \Sigma)$

Step 2: By applying similar analysis, we can show that the existence of h can be checked in time

$$(|\text{adom}(D)|)^{\#\text{variables}(Q)} \cdot |Q| \cdot |\text{chase}(D, \Sigma)|$$

$$\text{where } |\text{chase}(D, \Sigma)| \leq |\text{sch}(\Sigma)| \cdot (|\text{adom}(D)|)^{\text{maxarity}}$$



Data Complexity of FULL

Theorem: BCQ-Answering under FULL is in PTIME w.r.t. the data complexity

Proof: Consider a database D , a set $\Sigma \in \text{FULL}$, and a BCQ Q

We apply the naïve algorithm:

1. Construct $\text{chase}(D, \Sigma)$
2. Check for the existence of a homomorphism h such that $h(Q) \subseteq \text{chase}(D, \Sigma)$

Consequently, in the worst case, the naïve algorithm runs in time

$$\begin{aligned} & (|\text{sch}(\Sigma)| \cdot (|\text{adom}(D)|)^{\text{maxarity}})^2 \cdot |\Sigma| \cdot (|\text{adom}(D)|)^{\text{maxvariables}(\Sigma)} \cdot \text{maxbody}(\Sigma) \\ & \quad + \\ & (|\text{adom}(D)|)^{\#\text{variables}(Q)} \cdot |Q| \cdot |\text{sch}(\Sigma)| \cdot (|\text{adom}(D)|)^{\text{maxarity}} \end{aligned}$$



Data Complexity of FULL

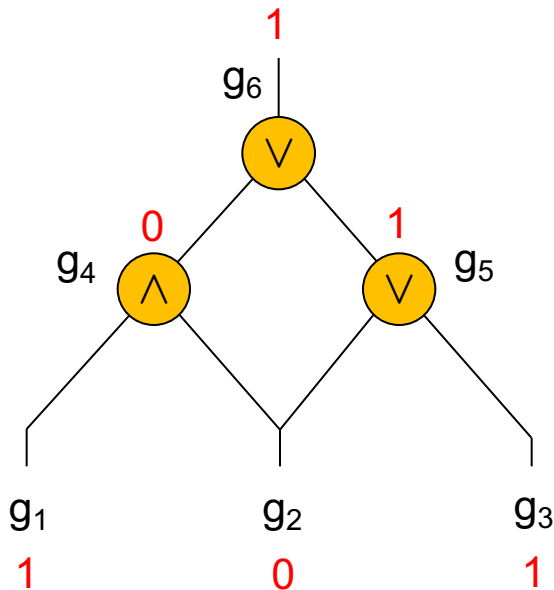
We cannot do better than the naïve algorithm

Theorem: BCQ-Answering under FULL is PTIME-hard w.r.t. the data complexity

Proof : By a LOGSPACE reduction from Monotone Circuit Value problem



Data Complexity of FULL



Does the circuit evaluate to *true*?

encoding of the circuit as a database D

$T(g_1)$ $T(g_3)$

$AND(g_4, g_1, g_2)$ $OR(g_5, g_2, g_3)$ $OR(g_6, g_4, g_5)$

evaluation of the circuit via a *fixed* set Σ

$\forall X \forall Y \forall Z (T(X) \wedge OR(Z, X, Y) \rightarrow T(Z))$

$\forall X \forall Y \forall Z (T(Y) \wedge OR(Z, X, Y) \rightarrow T(Z))$

$\forall X \forall Y \forall Z (T(X) \wedge T(Y) \wedge AND(Z, X, Y) \rightarrow T(Z))$

Circuit evaluates to *true* iff $D \wedge \Sigma \models T(g_6)$

Combined Complexity of FULL

Theorem: BCQ-Answering under FULL is in EXPTIME w.r.t. the combined complexity

Proof: Consider a database D , a set $\Sigma \in \text{FULL}$, and a BCQ Q

We apply the naïve algorithm:

1. Construct $\text{chase}(D, \Sigma)$
2. Check for the existence of a homomorphism h such that $h(Q) \subseteq \text{chase}(D, \Sigma)$

By our previous analysis, in the worst case, the naïve algorithm runs in time

$$\begin{aligned} & (|\text{sch}(\Sigma)| \cdot (|\text{adom}(D)|)^{\text{maxarity}})^2 \cdot |\Sigma| \cdot (|\text{adom}(D)|)^{\text{maxvariables}(\Sigma)} \cdot \text{maxbody}(\Sigma) \\ & + \\ & (|\text{adom}(D)|)^{\#\text{variables}(Q)} \cdot |Q| \cdot |\text{sch}(\Sigma)| \cdot (|\text{adom}(D)|)^{\text{maxarity}} \end{aligned}$$



Combined Complexity of FULL

We cannot do better than the naïve algorithm

Theorem: BCQ-Answering under FULL is EXPTIME-hard w.r.t. the combined complexity

Proof : By simulating a deterministic exponential time Turing machine



EXPTIME-hardness of FULL

Our Goal: Encode the exponential time computation of a DTM M on input

string I using a database D , a set $\Sigma \in \text{FULL}$, and a BCQ Q such that

$D \wedge \Sigma \models Q$ iff M accepts I in at most $N = 2^m$ steps, where $m = |I|^k$



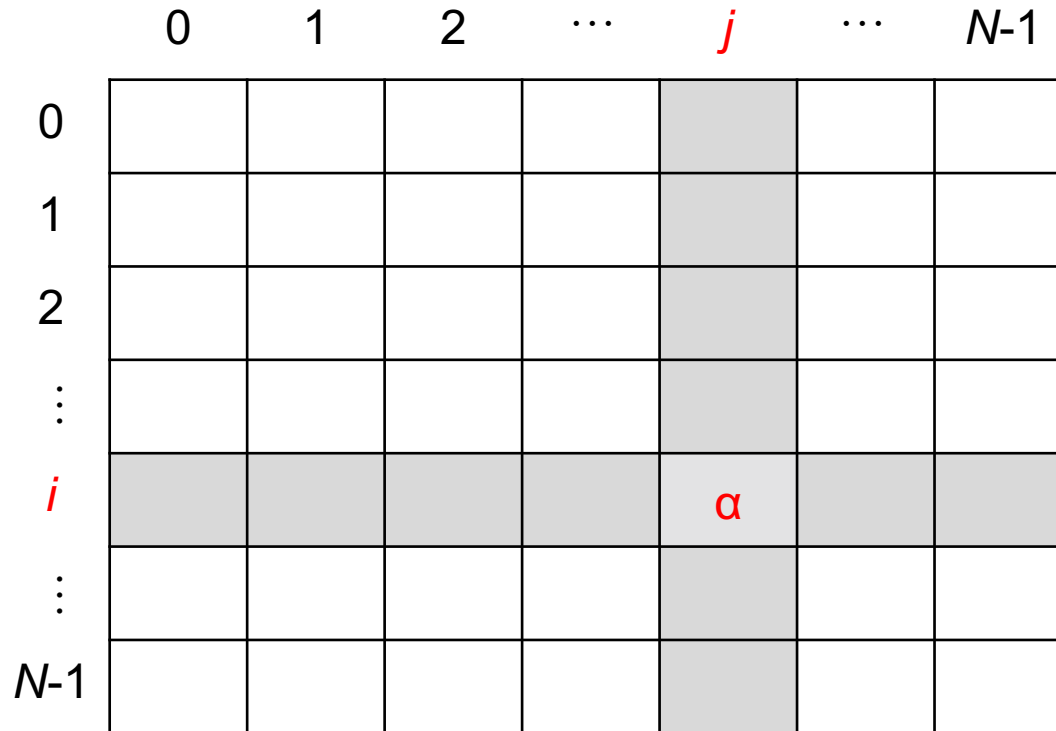
The Schema

	0	1	2	...	j	...	$N-1$
0							
1							
2							
⋮							
i					α		
⋮							
$N-1$							

Symbol $[\alpha](i,j)$ - at time instant i , cell j contains α



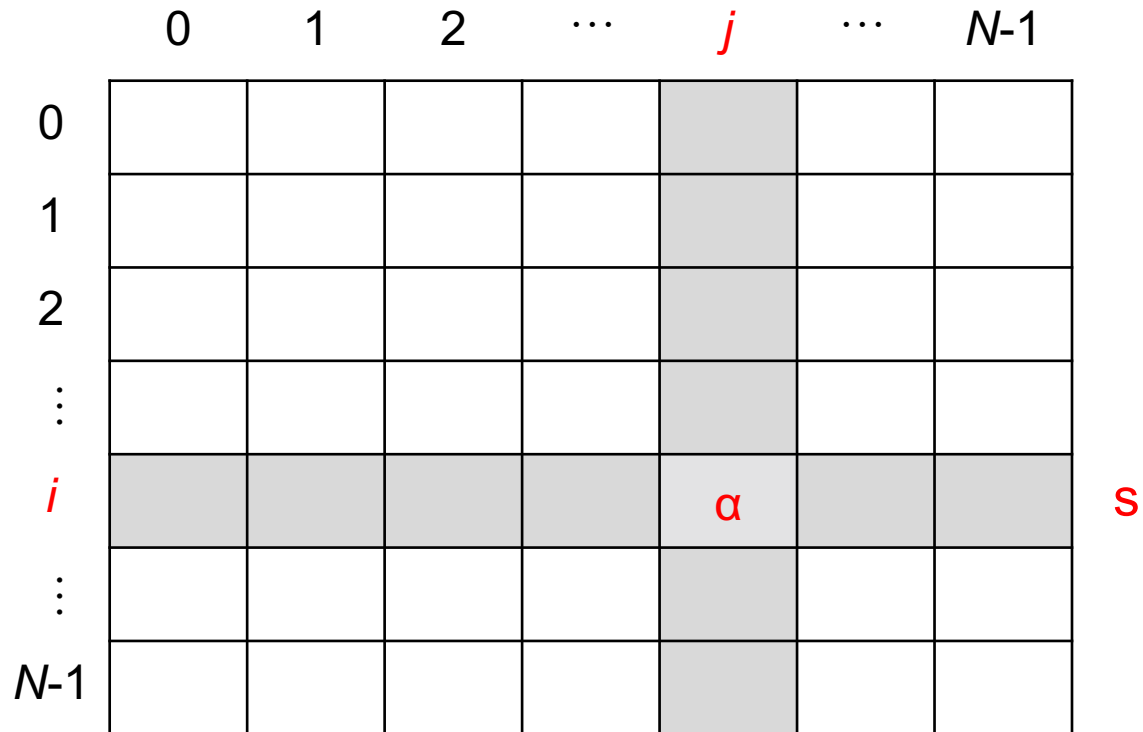
The Schema



Cursor(i,j) - at time instant i , cursor points to cell j



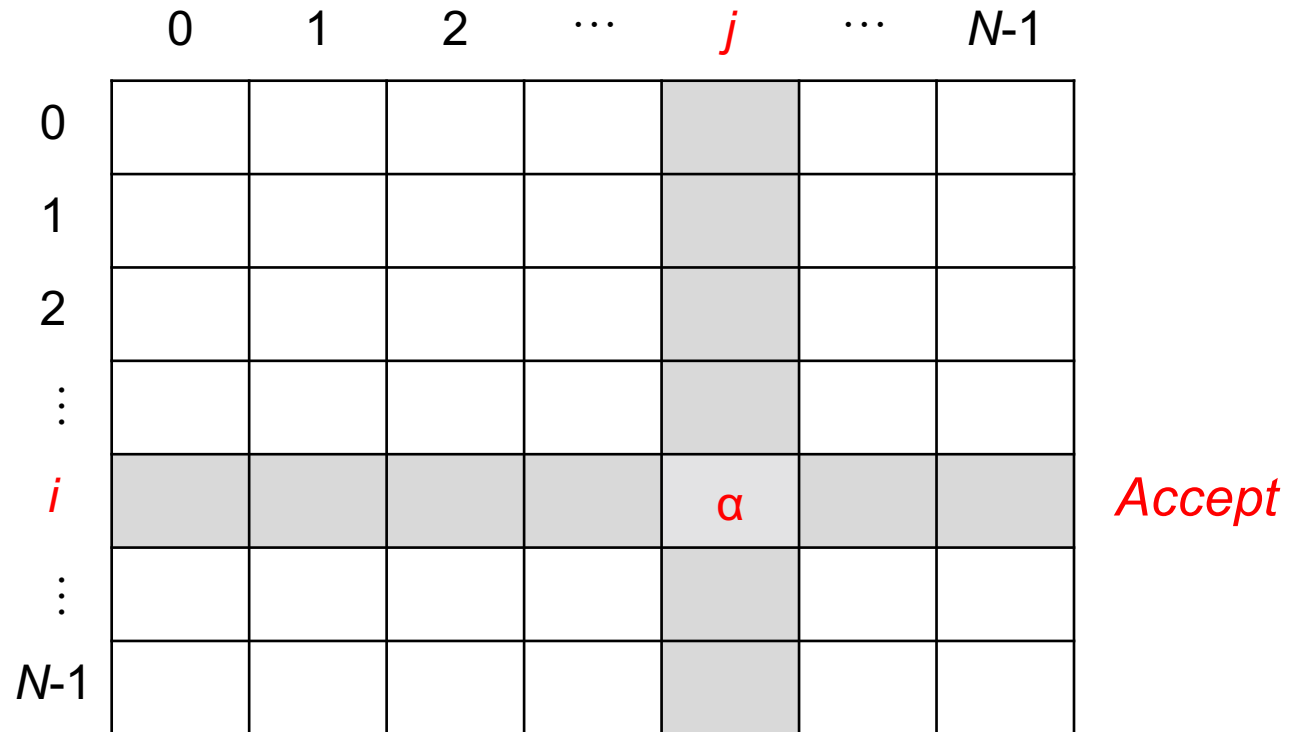
The Schema



State $[s](i)$ - at time instant i , the machine is in state s



The Schema



Accept(i) - at time instant i , the machine accepts



The Schema

	0	1	2	...	j	...	$N-1$
0							
1							
2							
⋮							
i							
⋮							
$N-1$							

$First(0), Succ(0,1), Succ(1,2), Succ(2,3), \dots, Succ(N-2,N-1)$

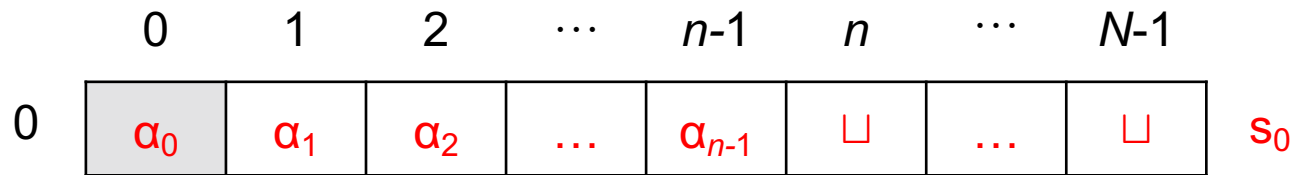
} will be defined later

\prec - transitive closure of $Succ$



Initialization Rules

Assume that $I = \alpha_0 \dots \alpha_{n-1}$



$$\forall T (First(T) \rightarrow Symbol[\alpha_i](T,i) \wedge Cursor(T,T) \wedge State[s_0](T))$$

$$\forall T \forall C (First(T) \wedge \prec(n-1,C) \rightarrow Symbol[\sqcup](T,C))$$



Transition Rules

$$\delta(s_1, \alpha) = (s_2, \beta, +1)$$

	j	$j+1$	$j+2$	
i	x	α	y	s_1
$i+1$	x	β	y	s_2

$$\forall T \forall T_1 \forall C \forall C_1 (State[s_1](T) \wedge Cursor(T, C) \wedge Symbol[\alpha](T, C) \wedge Succ(T, T_1) \wedge Succ(C, C_1) \rightarrow \\ Symbol[\beta](T_1, C_1) \wedge Cursor(T_1, C_1) \wedge State[s_2](T_1))$$



Inertia Rules

Cells that are not changed during the transition **keep their old values**

	j	$j+1$	$j+2$	
i	x	α	y	s_1
$i+1$	x	β	y	s_2

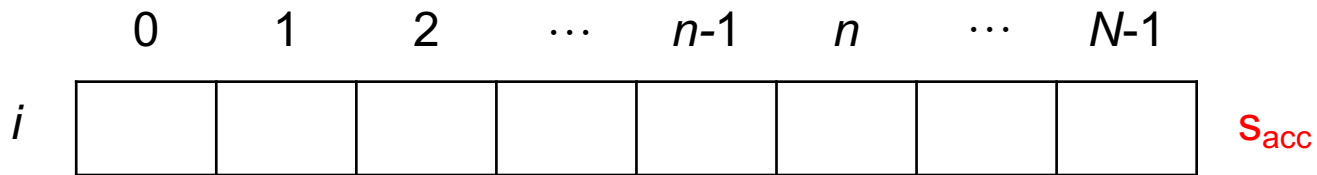
$$\forall T \forall T_1 \forall C \forall C_1 (Symbol[\alpha](T, C) \wedge Cursor(T, C_1) \wedge \prec(C, C_1) \wedge Succ(T, T_1) \rightarrow Symbol[\alpha](T_1, C))$$

$$\forall T \forall T_1 \forall C \forall C_1 (Symbol[\alpha](T, C) \wedge Cursor(T, C_1) \wedge \prec(C_1, C) \wedge Succ(T, T_1) \rightarrow Symbol[\alpha](T_1, C))$$



Accepting Rule

Once we reach the **accepting state** we accept



$$\forall T (State[s_{acc}](T) \rightarrow Accept(T))$$



Defining *First*, *Succ* and \prec

- $First(0), Succ(0,1), Succ(1,2), Succ(2,3), \dots, Succ(N-2,N-1)$
- In fact, $0, \dots, N-1$ are in **binary form** - assume the $N = 2^m$, where $m = 3$
 $First(0,0,0), Succ(0,0,0,0,0,1), Succ(0,0,1,0,1,0), \dots, Succ(1,1,0,1,1,1)$
- **Inductive definition** of $First_i$ and $Succ_i$

$$D = \{First_1(0), Last_1(1), Succ_1(0,1)\}$$

$$First_2(0,0), Last_2(1,1), Succ_2(0,0,0,1), Succ_2(0,1,1,0), Succ(1,0,1,1)$$

$$\forall X (First_1(X) \wedge First_1(X) \rightarrow First_2(X,X))$$

$$\forall X (Last_1(X), Last_1(X) \rightarrow Last_2(X,X))$$



Defining *First*, *Succ* and \prec

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$$D = \{First_1(0), Last_1(1), Succ_1(0,1)\}$$

$$First_2(0,0), Last_2(1,1), Succ_2(0,0,0,1), Succ_2(0,1,1,0), Succ(1,0,1,1)$$

$$\forall X \forall Y \forall Z (First_1(X), Succ_1(Y,Z) \rightarrow Succ_2(X,Y,X,Z))$$

$$\forall X \forall Y \forall Z (Last_1(X), Succ_1(Y,Z) \rightarrow Succ_2(X,Y,X,Z))$$



Defining *First*, *Succ* and \prec

- $First(0), Succ(0,1), Succ(1,2), Succ(2,3), \dots, Succ(N-2,N-1)$
- In fact, $0, \dots, N-1$ are in **binary form** - assume the $N = 2^m$, where $m = 3$
 $First(0,0,0), Succ(0,0,0,0,0,1), Succ(0,0,1,0,1,0), \dots, Succ(1,1,0,1,1,1)$
- **Inductive definition** of $First_i$ and $Succ_i$

$$D = \{First_1(0), Last_1(1), Succ_1(0,1)\}$$

$$First_2(0,0), Last_2(1,1), Succ_2(0,0,0,1), Succ_2(0,1,1,0), Succ(1,0,1,1)$$

$$\forall X \forall Y \forall Z \forall W (First_1(X), Last_1(Y), Succ_1(Z,W) \rightarrow Succ_2(Z,X,W,Y))$$



Defining *First*, *Succ* and \prec

$$D = \{First_1(0), Last_1(1), Succ_1(0,1)\}$$

Inductive definition of $First_{i+1}$ and $Succ_{i+1}$:

$$\forall \mathbf{X} \forall \mathbf{Y} (Succ_i(\mathbf{X}, \mathbf{Y}) \rightarrow Succ_{i+1}(Z, \mathbf{X}, Z, \mathbf{Y}))$$

$$\forall \mathbf{X} \forall \mathbf{Y} \forall Z \forall W (Succ_1(Z, W) \wedge Last_i(\mathbf{X}) \wedge First_i(\mathbf{Y}) \rightarrow Succ_{i+1}(Z, \mathbf{X}, W, \mathbf{Y}))$$

$$\forall \mathbf{X} \forall Z (First_1(Z) \wedge First_i(\mathbf{X}) \rightarrow First_{i+1}(Z, \mathbf{X}))$$

$$\forall \mathbf{X} \forall Z (Last_1(Z) \wedge Last_i(\mathbf{X}) \rightarrow Last_{i+1}(Z, \mathbf{X}))$$

Definition of \prec_m :

$$\forall \mathbf{X} \forall \mathbf{Y} (Succ_m(\mathbf{X}, \mathbf{Y}) \rightarrow \prec_m(\mathbf{X}, \mathbf{Y}))$$

$$\forall \mathbf{X} \forall \mathbf{Y} \forall Z (Succ_m(\mathbf{X}, Z) \wedge \prec_m(Z, \mathbf{Y}) \rightarrow \prec_m(\mathbf{X}, \mathbf{Y}))$$



Concluding EXPTIME-hardness of FULL

- Several rules but polynomially many \Rightarrow feasible in **polynomial time**
- $D \wedge \Sigma \models \exists X \text{Accept}(X)$ iff M accepts I in at most N steps
- Can be formally shown **by induction** on the time steps

Corollary: BCQ-Answering under FULL is **EXPTIME-complete w.r.t. the combined complexity**



Termination of the Chase

- Drop the existential quantification
 - We obtain the class of **full** existential rules
 - Very close to Datalog ✓

- Drop the recursive definitions
 - We obtain the class of **acyclic** existential rules
 - A.k.a. non-recursive existential rules

