

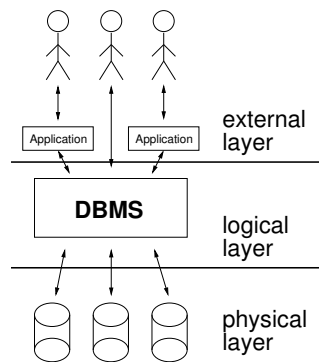
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## 1 Relational Databases

### 1.1 Relational Model

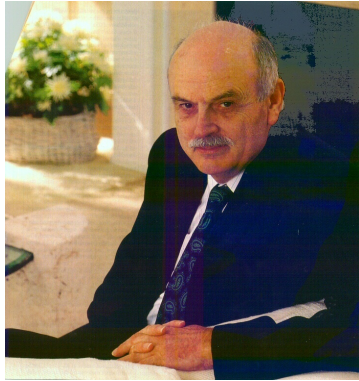
#### Three Layer Model



### Three Layer Model

- *External Layer* How external users view the database.
- *Logical/Conceptual Layer* Logical, holistic view of the database.
- *Physical/Internal Layer* Organisation on the physical media.

### Relational Model – Codd 1970



Edgar Frank Codd (1923–2003)

### Relations

- Schema:
  - Domain (denumerable set)
  - Attributes (denumerable set)
  - Relations (subset of attributes)
- Instances:
  - Relation instances: Sets of tuples.
  - Each tuple is a function from the relation's attributes to domain elements.
  - Database instance: Collection of relation instances.

### Relations: Example

$$A = \{X, Y\}, D = \{a, b, c, d\}$$

$$R = \{X, Y\}, S = \{Y\}$$

$$I(R) = \{t_1, t_2\}$$

$$t_1(X) = a, t_1(Y) = b, t_2(X) = c, t_2(Y) = d$$

$$I(S) = \{t_3\}, t_3(Y) = b$$

$$I(R) = \{\langle a, b \rangle, \langle c, d \rangle\}, I(S) = \{\langle b \rangle\}$$

### Relations: Example

$R$	$X$	$Y$
	$a$	$b$
	$c$	$d$

$S$	$Y$
	$d$

## 1.2 Relational Algebra

### Relational Algebra

Basic Operators:

- $\sigma$  Selection
- $\pi$  Projection
- $\times$  Cartesian Product
- $\cup$  Union
- $-$  Difference

Definable using Basic Operators:

- $\bowtie$  Join [  $R \bowtie S = \sigma_F(R \times S)$  ]
- $\ltimes$  Semijoin [  $R \ltimes S = \pi_{Schema(R)}(R \bowtie S)$  ]
- $\cap$  Intersection

### Relational Algebra Example

$R \times S$	$X_R$	$Y_R$	$Y_S$
	$a$	$b$	$d$
	$c$	$d$	$d$

$\sigma_{2=3}(R \times S)$	$X_R$	$Y_R$	$Y_S$
	$c$	$d$	$d$

$\pi_{1,2}(\sigma_{2=3}(R \times S))$	$X$	$Y$
	$c$	$d$

## 1.3 Relational Model – Logical View

### Relations – Logical View

- Schema:
  - Domain – Constant symbols (denumerable set)
  - Relations – Predicate symbols (attributes are not explicitly named)
  - Attributes – implicit by predicate arity
- Instances:
  - Relation instances: Subset of ground instances for relation predicate.
  - Database instance: Subset of Herbrand Base.

### Relations: Example

$$D = \{a, b, c, d\}$$

$$R/2, S/1$$

$$I(R) = \{R(a, b), R(c, d)\}, I(S) = \{S(d)\}$$

$$I = \{R(a, b), R(c, d), S(d)\}$$

## 2 Relational Calculus

### Relational Calculus

- Based on First-Order Logic
- Atomic formulas  $r(X_1, \dots, X_n)$
- Comparison formulas  $X = 2$  or  $X = Y$  (pre-interpreted predicate)
- Composed formulas using  $\neg, \wedge, \exists$
- $\rightarrow, \leftrightarrow, \vee, \forall$  added as “syntactic sugar”

### Relational Calculus

- Relational Algebra expressions represent relation instances
- In Relational Calculus:  $\{e_1, \dots, e_n \mid \phi\}$ 
  - $\phi$  is a Relational Calculus formula
  - $e_1, \dots, e_n$ : terms containing exactly the free variables of  $\phi$
- Collect all substitutions for free variables such that  $\phi$  is true in the interpretation formed by the database.
- The defined relation is obtained by applying all of these substitutions to  $e_1, \dots, e_n$ .

### Relational Calculus Examples

$$\{X, Y, Z \mid R(X, Y) \wedge S(Z)\} = \{T(a, b, d), T(c, d, d)\} = R \times S$$

$$\{X, Y, Y \mid R(X, Y) \wedge S(Y)\} = \{T(c, d, d)\} = \sigma_{2=3}(R \times S)$$

$$\{X, Y \mid R(X, Y) \wedge S(Y)\} = \{T(c, d)\} = \pi_{1,2}(\sigma_{2=3}(R \times S))$$

### Algebra as Calculus

- $\sigma_S r$   $\{X_1, \dots, X_n \mid r(X_1, \dots, X_n) \wedge S\}$
- $\pi_i r$   $\{X_i \mid \exists X_1, \dots, X_{i-1}, X_{i+1}, \dots, X_n : r(X_1, \dots, X_n)\}$
- $r \times s$   $\{X_1, \dots, X_n, Y_1, \dots, Y_m \mid r(X_1, \dots, X_n) \wedge s(Y_1, \dots, Y_m)\}$
- $r \cup s$   $\{X_1, \dots, X_n \mid r(X_1, \dots, X_n) \vee s(X_1, \dots, X_n)\}$
- $r - s$   $\{X_1, \dots, X_n \mid r(X_1, \dots, X_n) \wedge \neg s(X_1, \dots, X_n)\}$

## 3 Domain Independence

### 3.1 Domain Dependent Queries

#### Calculus: More than Algebra

Problematic expressions:

$$\begin{aligned} &\{X \mid \neg R(a, X)\} \\ &\{X, Y \mid R(a, X) \vee R(Y, b)\} \\ &\{X \mid \forall Y : R(X, Y)\} \end{aligned}$$

#### Calculus: More than Algebra

Using the *domain* of the database:

- $\{X \mid \neg R(a, X)\}$ 
  - all constants  $c$  of the domain such that  $(a, c)$  is no tuple in  $R$
  - will be infinite if the domain is infinite
- $\{X, Y \mid R(a, X) \vee R(Y, b)\}$ 
  - if  $R$  contains some tuple  $(a, b)$ ,  $(b, c)$  for all constants  $c$  in the domain
  - will be infinite if the domain is infinite
- $\{X \mid \forall Y : R(X, Y)\}$ 
  - this will be always empty if the domain is infinite, because relations are finite

### Calculus: More than Algebra

Using the *active domain* of the database (only constants appearing in the database and the query):

- $\{X \mid \neg R(a, X)\}$ 
  - all constants  $c$  in the database such that  $(a, c)$  is no tuple in  $R$
  - will change if some unrelated constant is added
- $\{X, Y \mid R(a, X) \vee R(Y, b)\}$ 
  - if  $R$  contains some tuple  $(a, b)$ ,  $(b, c)$  for all constants  $c$  in the database
  - will change if some unrelated constant is added
- $\{X \mid \forall Y : R(X, Y)\}$ 
  - will unintuitively become empty if an unrelated constant is added

### Natural versus Active Domain Semantics

1. *Natural Semantics*: Interpretations from Database Domain
  - pro: Classical First-Order theory
  - contra: Produces infinite relations
  - contra: Quantification over infinite sets
2. *Active Domain Semantics*: Interpretations from Active Domain
  - pro: Always finite
  - contra: Frequently gives unintuitive results
  - contra: Active Domain not always available

## 3.2 Domain Independent Queries

### Domain Independent Queries

*Idea*: Consider only those queries for which Natural and Active Domain Semantics coincide.

**Definition 1.** A query in the relational calculus is *domain independent*, if it yields the same answer using the natural (full) domain and the active domain.

### Domain Independent Queries

**Theorem 2.** Any query of the Relational Algebra can be written as a domain independent query of Relational Calculus, and vice versa.

Great, let's use only domain independent queries of Relational Calculus.

## Domain Independent Queries

**Theorem 3.** *Deciding whether a query of Relational Calculus is domain independent, is undecidable.*

### 3.3 Safe Range Queries

#### Safe Range Queries

Define a syntactically restricted fragment of Relational Calculus queries, which is guaranteed to be domain independent.

1. Transform formula into a normal form (SRNF).
2. Determine range restricted variables of the SRNF formula.
3. Check whether the range restricted variables are exactly the free variables.

#### SRNF

- Normalize variables: Rename variables, so that each quantifier binds a distinct variable and free and bound variables are different.
- Remove  $\forall$ :  $\forall X : \phi \Rightarrow \neg \exists X : \neg \phi$
- Remove  $\rightarrow$ :  $\phi \rightarrow \psi \Rightarrow \neg \phi \vee \psi$
- Remove  $\neg\neg$ :  $\neg\neg\phi \Rightarrow \phi$
- Push  $\neg$ :  $\neg(\phi \wedge \psi) \Rightarrow (\neg\phi \vee \neg\psi)$
- Push  $\neg$ :  $\neg(\phi \vee \psi) \Rightarrow (\neg\phi \wedge \neg\psi)$

Apply these rules as until none is applicable.

#### Range Restricted Variables

Intuition: Variables, for which the value is determined by the database.

- Variables in relational atoms are range restricted.
- Variables in equality comparisons with a constant are range restricted.
- Variables in conjunctions are range restricted if they are range restricted in the subformulas.
- Variables in disjunctions are only range restricted if they occur range restricted in both subformulas.
- Variables in negated formulas are never range restricted.
- Variables in existentially quantified subformulas (without the quantified variable) are range restricted if the quantified variable is range restricted in the subformula.

### Range Restriction Algorithm

Function  $rr$

Input: Formula  $\phi$  in SRNF

Output: Subset of free variables of  $\phi$  or  $\perp$

case  $\phi$  of

- $R(t_1, \dots, t_n)$ :  $rr(\phi) = \text{all variables in } t_1, \dots, t_n$ ;
- $X = a$  or  $a = X$ :  $rr(\phi) = \{X\}$ ;
- $\phi_1 \wedge \phi_2$ :  $rr(\phi) = rr(\phi_1) \cup rr(\phi_2)$ ;
- $\phi_1 \wedge X = Y$ :  $rr(\phi) = \begin{cases} rr(\phi_1) & \text{if } \{X, Y\} \cap rr(\phi_1) = \emptyset; \\ rr(\phi_1) \cup \{X, Y\} & \text{otherwise;} \end{cases}$
- $\phi_1 \vee \phi_2$ :  $rr(\phi) = rr(\phi_1) \cap rr(\phi_2)$ ;
- $\neg\phi_1$ :  $rr(\phi) = \emptyset$ ;
- $\exists X : \psi$ : if  $X \in rr(\psi)$  then  $rr(\phi) = rr(\psi) \setminus \{X\}$  else return  $\perp$ ;

Assumption: Set operations with  $\perp$  always result in  $\perp$ .

### Safe Range Queries

**Definition 4.** A Relational Calculus query  $\{e_1, \dots, e_n \mid \phi\}$  is *safe range*, if  $rr(SRNF(\phi))$  is equal to the free variables in  $\phi$ .

**Theorem 5.** Each safe range query is domain independent.

**Theorem 6.** Any safe range query can be written as query of Relational Algebra, and vice versa.

## 3.4 SQL

### SQL

- Exists since 1974 (developed by IBM).
- ISO/ANSI standardization 1986/87.
- First extension 1989.
- Second extension 1992 (SQL-92/SQL-2).
- Last (up to now) extension 1999/2000 (SQL-99/SQL-3)
- SQL combines query and manipulation languages.



## SQL

SELECT  $P$  FROM  $C$  WHERE  $S$

- $P$  — Projections
- $C$  — Cartesian Product
- $S$  — Selections

## SQL

*Union:*

SELECT ... UNION SELECT ...

*Nesting:*

SELECT  $P$  FROM  $C$  WHERE  $A$  [NOT] IN (SELECT ...)

## SQL

**Theorem 7.** *The query portion of SQL-92 basically corresponds to Relational Algebra, and hence to safe range Relational Calculus.*

# 4 Datalog

## 4.1 Motivation

### Recursion

- Some simple problems cannot be represented in relational calculus.
- Example: Reachability on deterministic graphs.
- Prototypical for LOGSPACE!
- Holds also for relational algebra, SQL-92 etc.

### Transitive Closure

Key notion: Transitive Closure

**Definition 8.** Given graph  $G = \langle V, E \rangle$ ,  $E \subseteq V \times V$ , and  $a, b \in V$ , the *transitive closure*  $TC(G) \subseteq V \times V$  is:

$$TC(G) = \{(x, y) \mid (x, y) \in E\} \cup \{(x, y) \mid (x, z) \in TC(G) \wedge (z, y) \in TC(G)\}$$

*Note:*  $TC(G)$  appears in its own definition.

In relational calculus we cannot refer to what we define.

- Idea: Use Horn clauses for named definitions.
- It is then possible to write definitions using the concept being defined.
- Positive Datalog

## 4.2 Syntax

### Language Elements

- Set of extensional predicate symbols  $\mathbf{PS}_{\text{EDB}}$
- Set of intensional predicate symbols  $\mathbf{PS}_{\text{IDB}}$
- $\mathbf{PS}_{\text{EDB}} \cap \mathbf{PS}_{\text{IDB}} = \emptyset$
- Each predicate symbol has an associated arity  $ar : \mathbf{PS}_{\text{EDB}} \cup \mathbf{PS}_{\text{IDB}} \rightarrow \mathbb{N}_0$
- Set of constant symbols  $\mathbf{CS}$
- Set of variable symbols  $\mathbf{VS}$

### Syntax

A Datalog rule is of the form:

$$r_1(t_{1_1}, \dots, t_{n_1}) \leftarrow r_2(t_{1_2}, \dots, t_{n_2}), \dots, r_m(t_{1_m}, \dots, t_{n_m}).$$

- $m \geq 1$
- $r_1 \in \mathbf{PS}_{\text{IDB}}$
- $r_2, \dots, r_m \in \mathbf{PS}_{\text{EDB}} \cup \mathbf{PS}_{\text{IDB}}$
- $t_{1_1}, \dots, t_{n_m} \in \mathbf{CS} \cup \mathbf{VS}$
- $\forall i \ 1 \leq i \leq m : ar(r_i) = n_i$
- $((t_{1_1} \cup \dots \cup t_{n_1}) \cap \mathbf{VS}) \subseteq ((t_{1_2} \cup \dots \cup t_{n_m}) \cap \mathbf{VS})$

### Syntax

$$r_1(t_{1_1}, \dots, t_{n_1}) \leftarrow r_2(t_{1_2}, \dots, t_{n_2}), \dots, r_m(t_{1_m}, \dots, t_{n_m}).$$

- $H(r) = \{r_1(t_{1_1}, \dots, t_{n_1})\}$
- $B(r) = \{r_2(t_{1_2}, \dots, t_{n_2}), \dots, r_m(t_{1_m}, \dots, t_{n_m})\}$
- $V(r) = \{t_{1_1}, \dots, t_{n_m}\} \cap \mathbf{VS}$
- $C(r) = \{t_{1_1}, \dots, t_{n_m}\} \cap \mathbf{CS}$

- $H(r)$  is the *head* of  $r$ .
- $B(r)$  is the *body* of  $r$ .
- A *Datalog program* is a set of rules.

### 4.3 Semantics

#### Semantics

Intuitively: For each rule  $r$ , whenever  $B(r)$  is true,  $H(r)$  should also be true.  $B(r) = \emptyset$  is considered to be true.

Different ways for defining the semantics:

- *model theory*
- *fixpoint theory*
- *proof theory*

#### 4.3.1 Model Theory

##### Model Theory

**Definition 9** (Herbrand Universe).

$$\mathbf{HU}(\mathcal{P}) = \bigcup_{r \in \mathcal{P}} C(r)$$

**Definition 10** (Herbrand Base).

$$\mathbf{HB}(\mathcal{P}) = \{r(t_1, \dots, t_n) \mid r \in \mathbf{PS}_{\text{EDB}} \cup \mathbf{PS}_{\text{IDB}}, \\ t_1, \dots, t_n \in \mathbf{HU}(\mathcal{P}), ar(r) = n\}$$

- $\mathbf{HU}(\mathcal{P})$ : Constants of the program (active domain!)
- $\mathbf{HB}(\mathcal{P})$ : Ground atoms constructible from  $\mathbf{HU}(\mathcal{P})$

**Example: Herbrand Base**

$$\begin{aligned} \mathcal{P}_r &= \{ \text{arc}(a, b). \\ &\quad \text{arc}(b, c). \\ &\quad \text{reachable}(a). \\ &\quad \text{reachable}(Y) \leftarrow \text{arc}(X, Y), \text{reachable}(X). \} \\ \mathbf{HU}(\mathcal{P}_r) &= \{a, b, c\} \\ \mathbf{HB}(\mathcal{P}_r) &= \{ \text{arc}(a, a), \text{arc}(a, b), \text{arc}(a, c), \\ &\quad \text{arc}(b, a), \text{arc}(b, b), \text{arc}(b, c), \\ &\quad \text{arc}(c, a), \text{arc}(c, b), \text{arc}(c, c), \\ &\quad \text{reachable}(a), \text{reachable}(b), \text{reachable}(c) \} \end{aligned}$$

### Instantiation

**Definition 12.** Valuation  $v_{\mathcal{P}}(r)$  of a rule  $r$ : Set of all substitutions  $V(r) \rightarrow \mathbf{HU}(\mathcal{P})$

**Definition 13** (Instantiation of a rule  $r$ ).  $Ground_{\mathcal{P}}(r) = \bigcup_{v \in v_{\mathcal{P}}(r)} v(r)$

**Definition 14** (Instantiation of a program  $\mathcal{P}$ ).  $Ground(\mathcal{P}) = \bigcup_{r \in \mathcal{P}} Ground_{\mathcal{P}}(r)$

### Example: Instantiation

*Example 15.*  $\mathcal{P}_r = \{ \text{arc}(a, b), \text{arc}(b, c), \text{reachable}(a),$   
 $\text{reachable}(Y) \leftarrow \text{arc}(X, Y), \text{reachable}(X). \}$

$Ground(\mathcal{P}_r) = \{ \text{arc}(a, b), \text{arc}(b, c), \text{reachable}(a),$   
 $\text{reachable}(a) \leftarrow \text{arc}(a, a), \text{reachable}(a),$   
 $\text{reachable}(b) \leftarrow \text{arc}(a, b), \text{reachable}(a),$   
 $\text{reachable}(c) \leftarrow \text{arc}(a, c), \text{reachable}(a),$   
 $\text{reachable}(a) \leftarrow \text{arc}(b, a), \text{reachable}(b),$   
 $\text{reachable}(b) \leftarrow \text{arc}(b, b), \text{reachable}(b),$   
 $\text{reachable}(c) \leftarrow \text{arc}(b, c), \text{reachable}(b),$   
 $\text{reachable}(a) \leftarrow \text{arc}(c, a), \text{reachable}(c),$   
 $\text{reachable}(b) \leftarrow \text{arc}(c, b), \text{reachable}(c),$   
 $\text{reachable}(c) \leftarrow \text{arc}(c, c), \text{reachable}(c). \}$

### Herbrand Models

**Definition 16** ((Herbrand-) Interpretations  $I$  for  $\mathcal{P}$ ).  $I \subseteq \mathbf{HB}(\mathcal{P})$

**Definition 17** ((Herbrand-) Models for  $\mathcal{P}$ ).  $M \subseteq \mathbf{HB}(\mathcal{P})$  such that  $\forall r \in Ground(\mathcal{P}) :$   
 $(H(r) \subseteq M) \vee (B(r) \not\subseteq M)$

“If the body is true, the head must be true.”

**Definition 18** ((Herbrand-) Models for  $\mathcal{P}$ ).  $M \subseteq \mathbf{HB}(\mathcal{P})$  such that  $\forall r \in Ground(\mathcal{P}) :$   
 $(B(r) \subseteq M) \rightarrow (H(r) \subseteq M)$

### Example: Herbrand Models

*Example 19.*  $\mathcal{P}_r = \{ \text{arc}(a, b), \text{arc}(b, c), \text{reachable}(a),$   
 $\text{reachable}(Y) \leftarrow \text{arc}(X, Y), \text{reachable}(X). \}$

$M_1 = \{ \text{arc}(a, b), \text{arc}(b, c),$   
 $\text{reachable}(a), \text{reachable}(b), \text{reachable}(c) \}$

$M_2 = \mathbf{HB}(\mathcal{P}_r)$

All  $M : M_1 \subseteq M \subseteq M_2$  are models and only these.

### Minimal Models

**Theorem 20.**  $\mathbf{HB}(\mathcal{P})$  is always a model for any Datalog program  $\mathcal{P}$ .

**Theorem 21.** Each Datalog program  $\mathcal{P}$  has a unique subset minimal model  $MM(\mathcal{P})$ .

**Definition 22.** The semantics of a Datalog program  $\mathcal{P}$  is given by  $MM(\mathcal{P})$

*Note:* Each element of  $MM(\mathcal{P})$  is a logical consequence of  $\mathcal{P}$ .

### 4.3.2 Fixpoint Theory

#### Concept: Operator

“If we assume that all atoms in  $I$  are true, which other atoms must be true in order to satisfy the program?”

- Start with  $I = \emptyset$  (nothing is true).
- Define operator  $\mathbf{T}_{\mathcal{P}}$ .
- Apply  $\mathbf{T}_{\mathcal{P}}$ , until there are no further additions.
- The obtained result (fixpoint) defines the semantics.

#### Immediate Consequences

**Definition 23** (Operator  $\mathbf{T}_{\mathcal{P}}$  for Datalog program  $\mathcal{P}$ ). Given an interpretation  $I$ ,

$$\mathbf{T}_{\mathcal{P}}(I) = \{h \mid r \in \text{Ground}(\mathcal{P}), B(r) \subseteq I, h \in H(r)\}$$

- $\mathbf{T}_{\mathcal{P}}(I)$  extends  $I$ , such that unsatisfied rules (w.r.t.  $I$ ) become satisfied.
- Other rules may become unsatisfied w.r.t.  $\mathbf{T}_{\mathcal{P}}(I)$ .
- $\Rightarrow$  Iterative application.

#### Example: Immediate Consequences

*Example 24.*  $\mathcal{P}_r = \{ \text{arc}(a, b), \text{arc}(b, c), \text{reachable}(a), \\ \text{reachable}(Y) \leftarrow \text{arc}(X, Y), \text{reachable}(X). \}$

1.  $\mathbf{T}_{\mathcal{P}_r}(\emptyset) = \{\text{arc}(a, b), \text{arc}(b, c), \text{reachable}(a)\}$
2.  $\mathbf{T}_{\mathcal{P}_r}(\mathbf{T}_{\mathcal{P}_r}(\emptyset)) = \mathbf{T}_{\mathcal{P}_r}(\emptyset) \cup \{\text{reachable}(b)\}$
3.  $\mathbf{T}_{\mathcal{P}_r}(\mathbf{T}_{\mathcal{P}_r}(\mathbf{T}_{\mathcal{P}_r}(\emptyset))) = \mathbf{T}_{\mathcal{P}_r}(\mathbf{T}_{\mathcal{P}_r}(\emptyset)) \cup \{\text{reachable}(c)\}$
4.  $\mathbf{T}_{\mathcal{P}_r}(\mathbf{T}_{\mathcal{P}_r}(\mathbf{T}_{\mathcal{P}_r}(\mathbf{T}_{\mathcal{P}_r}(\emptyset)))) = \mathbf{T}_{\mathcal{P}_r}(\mathbf{T}_{\mathcal{P}_r}(\mathbf{T}_{\mathcal{P}_r}(\emptyset)))$
5.  $\{\text{arc}(a, b), \text{arc}(b, c), \text{reachable}(a), \text{reachable}(b), \text{reachable}(c)\}$

#### Properties of $\mathbf{T}_{\mathcal{P}}$

Lattice:  $V = (P(\mathbf{HB}(\mathcal{P})), \subseteq)$ [0.5cm]

$\forall X \subseteq V : \exists \text{inf}(X) \wedge \exists \text{sup}(X)$ [0.5cm]

$\text{inf}(V) = \emptyset, \text{sup}(V) = \mathbf{HB}(\mathcal{P})$ [0.5cm]

**Monotony:**  $X \subseteq Y \rightarrow \mathbf{T}_{\mathcal{P}}(X) \subseteq \mathbf{T}_{\mathcal{P}}(Y)$ [0.5cm]

**Continuity:**  $\forall X \subseteq V : \mathbf{T}_{\mathcal{P}}(\text{sup}(X)) = \text{sup}(\mathbf{T}_{\mathcal{P}}(X))$

## Tarski, Kleene



Bronisław Knaster

Alfred Tarski  
(1902–1983)

Stephen Kleene (1893–1990)  
(1909–1994)

### Existence of Fixpoints

**Theorem 25.**  $\mathbf{T}_{\mathcal{P}}$  is monotone und continuous on the lattice of interpretations and subset relations.

**Theorem 26 (Knaster-Tarski).** For monotone operators on lattices a least fixpoint exists, and it is  $\inf(\{X \mid \mathbf{T}_{\mathcal{P}}(X) \subseteq X\})$

### Construction of Fixpoints

**Theorem 27 (Kleene).** For continuous operators on lattices the least fixpoint can be computed by iteration starting from the infimum.  $\mathbf{T}_{\mathcal{P}}^{\omega} = \sup(\{\mathbf{T}_{\mathcal{P}}^i \mid i \geq 0\})$ ,  $\mathbf{T}_{\mathcal{P}}^0 = \inf(V)$ ,  $\mathbf{T}_{\mathcal{P}}^i = \mathbf{T}_{\mathcal{P}}(\mathbf{T}_{\mathcal{P}}^{i-1})$

**Corollary 28.** Our lattice is finite, therefore the least fixpoint of  $\mathbf{T}_{\mathcal{P}}$  can be computed by a finite number of iterations starting from  $\emptyset$ .

### $\mathbf{T}_{\mathcal{P}}^{\omega}$ – Minimal Model

**Theorem 29.** For all Datalog programs  $\mathcal{P}$ , we can show  $\mathbf{T}_{\mathcal{P}}^{\omega} = MM(\mathcal{P})$ .

*Note:* All consequences of a program can be computed by iteration over the immediate consequences.

### 4.3.3 Proof Theory

#### Reminder: Horn and Goal Clauses, SLD Resolution

- A *Horn clause* is a clause containing at most one positive literal.
- A *Goal clause* is a clause containing no positive literal.
- *SLD Resolution:* Linear resolution, where at each step only goal clauses and (instances of) input clauses are used.

**Theorem 30.** SLD resolution is refutation complete for Horn clauses.

### SLD Resolution for Datalog

- We can view each rule as a Horn clause.
- So SLD Resolution can be applied.
- Unification is simpler for Datalog because of absence of function symbols.

**Definition 31** (SLD Resolution Semantics). Let  $SLD(\mathcal{P})$  denote the set of ground atoms, for which an SLD refutation w.r.t.  $\mathcal{P}$  exists.

### Equivalence

**Theorem 32.** For all Datalog programs  $\mathcal{P}$ , we can show  $SLD(\mathcal{P}) = \mathbf{T}_{\mathcal{P}}^{\omega} = MM(\mathcal{P})$ .

### SLD Tree

**Top-down** and **bottom-up** views:[0.5cm]

1. *Top-down*: Start at the root.
2. *Bottom-up*: Start at leaves.

$\mathbf{T}_{\mathcal{P}}^{\omega}$ : Is like SLD bottom-up (on finite branches).

### SLD Resolution – Termination

```
child_of(charles, francis).
child_of(francis, frida).
successor_of(X, Y) ← child_of(X, Y).
successor_of(X, Y) ← child_of(X, Z), successor_of(Z, Y).
← successor_of(charles, X).
```

### SLD Resolution – Termination

```
child_of(charles, francis).
child_of(francis, frida).
successor_of(X, Y) ← child_of(X, Y).
successor_of(X, Y) ← successor_of(X, Z), child_of(Z, Y).
← successor_of(charles, X).
```

### SLD Resolution – Termination

```
child_of(charles, francis).
child_of(francis, frida).
successor_of(X, Y) ← child_of(X, Y).
successor_of(X, Y) ← successor_of(X, Z), successor_of(Z, Y).
← successor_of(charles, X).
```

## 4.4 Computation

### Simple Algorithms

From the semantic definitions, we can produce simple algorithms:

- *Model Theory*: Enumerate all subsets of  $\mathbf{HB}(\mathcal{P})$ , test whether they are models and take the minimal one.
- *Fixpoint Theory*: Extend  $\emptyset$  by applying  $\mathbf{T}_{\mathcal{P}}$  until a fixpoint is reached.
- *Proof Theory*: Use SLD Resolution bottom-up.

### Simple Algorithms 2

Also for query answering we can find simple algorithms:

- Straightforward: Compute model and test whether query is true.
- Better: Use SLD resolution top-down.

Termination?