

# Efficient Algorithms, Spring 2021

## 3. Functional Aggregate Queries

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**DaST**   
Data • (Systems+Theory)

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Zurich<sup>UZH</sup>

<https://lms.uzh.ch/url/RepositoryEntry/16979230773>

## Notation

- $i \in [n] = \{1, \dots, n\}$
- $X_1, \dots, X_n$  are variables
- $x_i$  are values in discrete domain  $\text{Dom}(X_i)$
- $\mathbf{x} = (x_1, \dots, x_n) \in \text{Dom}(X_1) \times \dots \times \text{Dom}(X_n)$
- For any  $S \subseteq [n]$ ,

$$\mathbf{x}_S = (x_i)_{i \in S} \in \prod_{i \in S} \text{Dom}(X_i)$$

$$\text{e.g., } \mathbf{x}_{\{2,5,8\}} = (x_2, x_5, x_8) \in \text{Dom}(X_2) \times \text{Dom}(X_5) \times \text{Dom}(X_8)$$

# Hypergraphs

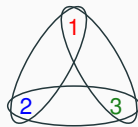
$\mathcal{H} = (\mathcal{V}, \mathcal{E})$  is a (multi)hypergraph

- $\mathcal{V}$  is the set of nodes, with one node per variable
- $\mathcal{E}$  is the (multi)set of hyperedges, with each hyperedge  $S \in \mathcal{E}$  a subset of  $\mathcal{V}$

Examples:

- $\mathcal{V} = \{1, 2, 3\}$
- $\mathcal{E} = \{\{1, 2\}, \{1, 3\}, \{2, 3\}\}$

The hyperedges could be the edge relation of a graph



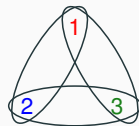
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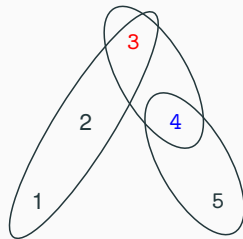
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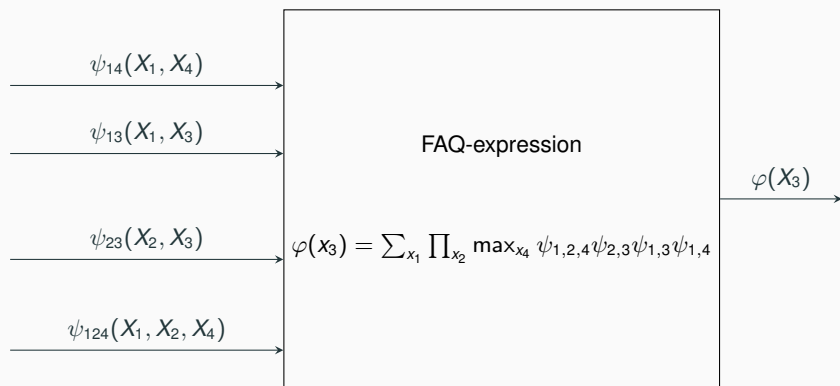
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- $\mathcal{V} = \{1, 2, 3, 4, 5\}$
- $\mathcal{E} = \{\{1, 2, 3\}, \{3, 4\}, \{4, 5\}\}$

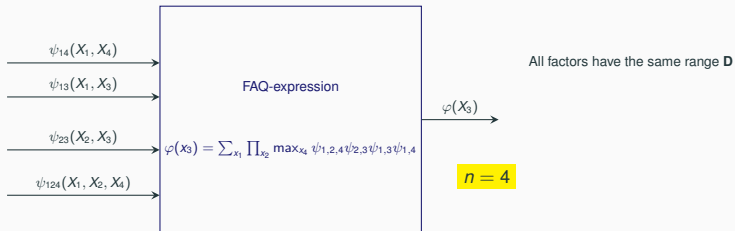
The hyperedges could be: Orders(customer, day, dish),  
Dishes(dish, item), Items(item, price)



## Functional Aggregate Queries: The Problem



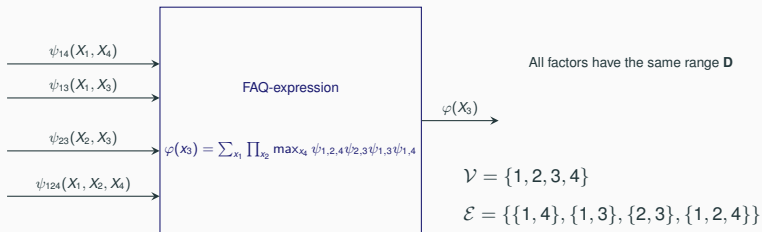
# Functional Aggregate Queries: The Input



- $n$  variables  $X_1, \dots, X_n$
- a multi-hypergraph  $\mathcal{H} = (\mathcal{V}, \mathcal{E})$ 
  - Each vertex is a variable:  $\mathcal{V} = [n]$
  - To each hyperedge  $S \in \mathcal{E}$  there corresponds a factor  $\psi_S$

$$\psi_S : \prod_{i \in S} \text{Dom}(X_i) \rightarrow \mathbf{D}$$

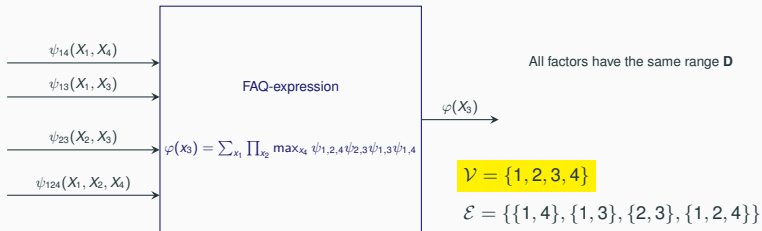
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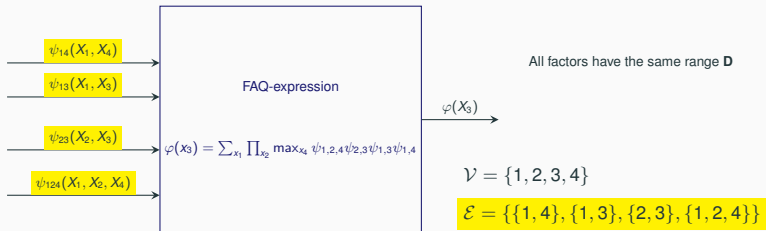


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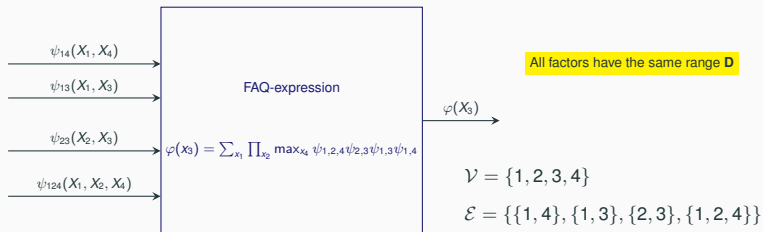
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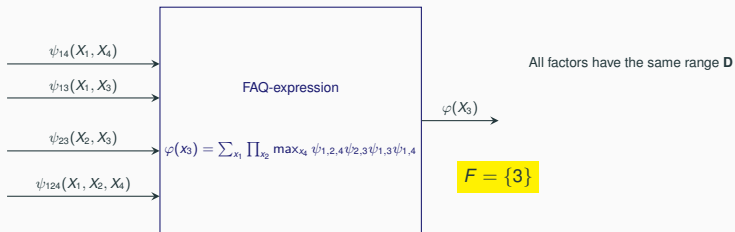
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↑

$\mathbb{R}, \{\text{true}, \text{false}\}, \{0, 1\}, 2^{\mathcal{U}}, \text{etc.}$

# Functional Aggregate Queries: The Input



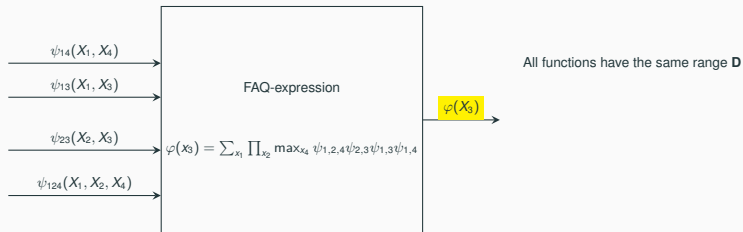
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$\uparrow$   
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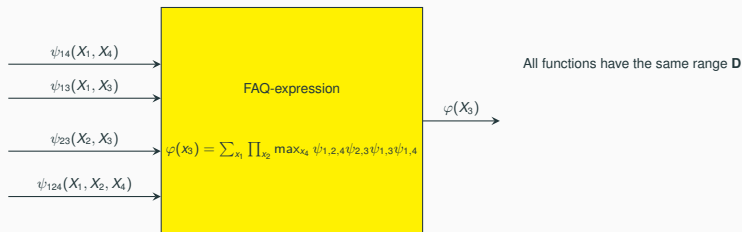
- a set  $F \subseteq \mathcal{V}$  of free variables; w.l.o.g.,  $F = [f] = \{1, \dots, f\}$

# Functional Aggregate Queries: The Output



- Compute the function  $\varphi : \prod_{i \in F} \text{Dom}(X_i) \rightarrow \mathbf{D}$ .

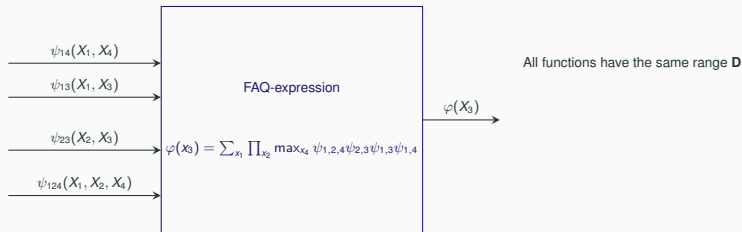
# Functional Aggregate Queries: The Output



- Compute the function  $\varphi : \prod_{i \in F} \text{Dom}(X_i) \rightarrow \mathbf{D}$ .
- $\varphi$  defined by the FAQ-expression

$$\varphi(\mathbf{x}_{[f]}) = \bigoplus_{x_{f+1} \in \text{Dom}(X_{f+1})}^{(f+1)} \cdots \bigoplus_{x_{n-1} \in \text{Dom}(X_{n-1})}^{(n-1)} \bigoplus_{x_n \in \text{Dom}(X_n)}^{(n)} \bigotimes_{S \in \mathcal{E}} \psi_S(\mathbf{x}_S)$$

# Functional Aggregate Queries: The Output

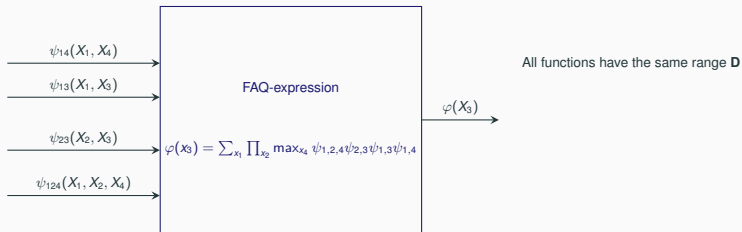


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- For each  $\bigoplus^{(i)}$

# Functional Aggregate Queries: The Output

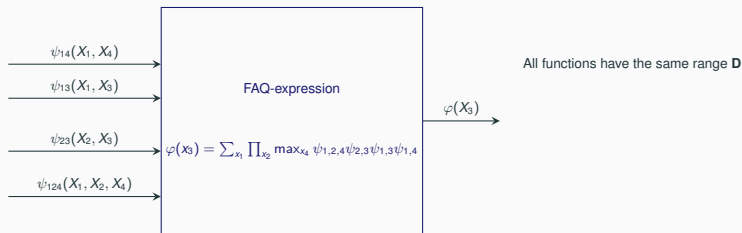


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- For each  $\bigoplus^{(i)}$ 
  - Either  $(\mathbf{D}, \bigoplus^{(i)}, \bigotimes, \mathbf{0}, \mathbf{1})$  is a commutative semiring

# Functional Aggregate Queries: The Output



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- For each  $\bigoplus^{(i)}$ 
  - Either  $(\mathbf{D}, \bigoplus^{(i)}, \bigotimes, \mathbf{0}, \mathbf{1})$  is a **commutative semiring**
  - Or  $\bigoplus^{(i)} = \bigotimes$



# Functional Aggregate Queries: Putting it Together

An FAQ expression has the form

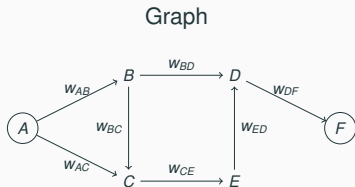
$$\varphi(\mathbf{x}_{[f]}) = \bigoplus_{x_{f+1} \in \text{Dom}(X_{f+1})}^{(f+1)} \cdots \bigoplus_{x_n \in \text{Dom}(X_n)}^{(n)} \bigotimes_{S \in \mathcal{E}} \psi_S(\mathbf{x}_S)$$

where

- $X_1, \dots, X_n$  are variables with domains  $\text{Dom}(X_1), \dots, \text{Dom}(X_n)$ 
  - $X_1, \dots, X_f$  are the **free** variables,  $X_{f+1}, \dots, X_n$  are the **bound** variables
- $\mathcal{H} = (\mathcal{V}, \mathcal{E})$  is a multi-hypergraph, where
  - $\mathcal{V} = [n]$  is the index set of variables
  - $\mathcal{E}$  is a set of subsets  $S$  of  $\mathcal{V}$
- Each  $\psi_S$  is an input function or **factor** over variables with index set  $S \in \mathcal{E}$ 
  - $\psi_S$  maps tuples over  $S$  to elements in a finite set  $\mathbf{D}$
  - Semirings  $(\mathbf{D}, \oplus^{(i)}, \otimes, \mathbf{0}, \mathbf{1})$  have the **same support**  $\mathbf{D}$

## Expressing Problems in FAQ

# PATH: Example



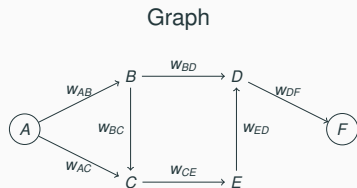
Edge relation  $E$

| start | end | weight   |
|-------|-----|----------|
| A     | B   | $w_{AB}$ |
| A     | C   | $w_{AC}$ |
| B     | D   | $w_{BD}$ |
| B     | C   | $w_{BC}$ |
| C     | E   | $w_{CE}$ |
| E     | D   | $w_{ED}$ |
| D     | F   | $w_{DF}$ |

Vertices  $V$

| Node |
|------|
| A    |
| B    |
| C    |
| D    |
| E    |
| F    |

# PATH: Example



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| C     | E   | $w_{CE}$ |
| E     | D   | $w_{ED}$ |
| D     | F   | $w_{DF}$ |

Vertices  $V$

| Node |
|------|
| A    |
| B    |
| C    |
| D    |
| E    |
| F    |

- The factor  $\psi$  maps each edge  $(x_i, x_j) \in E$  to  $w_{ij}$
- FAQ expresses graph traversal from 1 hop to 5 hops (length of longest path)

$$\begin{aligned}
 \psi(A, F) \oplus & \left( \bigoplus_{x_1 \in V} \psi(A, x_1) \otimes \psi(x_1, F) \right) \oplus \left( \bigoplus_{x_1, x_2 \in V} \psi(A, x_1) \otimes \psi(x_1, x_2) \otimes \psi(x_2, F) \right) \\
 & \oplus \left( \bigoplus_{x_1, x_2, x_3 \in V} \psi(A, x_1) \otimes \psi(x_1, x_2) \otimes \psi(x_2, x_3) \otimes \psi(x_3, F) \right) \\
 & \oplus \left( \bigoplus_{x_1, x_2, x_3, x_4 \in V} \psi(A, x_1) \otimes \psi(x_1, x_2) \otimes \psi(x_2, x_3) \otimes \psi(x_3, x_4) \otimes \psi(x_4, F) \right)
 \end{aligned}$$

Compute path problem over vertices  $V$  and weighted edge relation  $E$

FAQ encoding over the **semiring**  $(\mathbf{D}, \oplus, \otimes, \mathbf{0}, \mathbf{1})$ :

$$\begin{aligned}\Phi() &= \left( \bigoplus_{i \in [2]: x_i \in V} \psi_{12}(x_1, x_2) \right) \oplus && // \text{ 1 hop} \\ &\left( \bigoplus_{i \in [3]: x_i \in V} \psi_{12}(x_1, x_2) \otimes \psi_{23}(x_2, x_3) \right) \oplus \cdots \oplus && // \text{ 2 hops} \\ &\left( \bigoplus_{i \in [n+1]: x_i \in V} \psi_{12}(x_1, x_2) \otimes \cdots \otimes \psi_{n,n+1}(x_n, x_{n+1}) \right) && // \text{ n hops}\end{aligned}$$

$\psi_{i,i+1}(x_i, x_{i+1})$  maps each edge  $(x_i, x_{i+1}) \in E$  to a semiring-dependent weight

- **min-sum** over reals for shortest distance
- **max-min** over  $\{0, 1\}$  for connectivity and over reals for largest capacity
- **max-product** over  $[0, 1]$  for maximum reliability
- **concatenate-union** over strings for language accepted by automaton

## SAT: Example

$$(R_{CH} \vee G_{CH} \vee B_{CH})$$

$$\wedge$$

$$(\neg R_{CH} \vee \neg G_{CH}) \wedge (\neg R_{CH} \vee \neg B_{CH}) \wedge$$

$$(\neg G_{CH} \vee \neg B_{CH})$$

$$\wedge$$

$$(\neg R_{CH} \vee \neg R_{DE}) \wedge (\neg G_{CH} \vee \neg G_{DE}) \wedge$$

$$(\neg B_{CH} \vee \neg B_{DE})$$

$$\dots$$

“Switzerland has *at least* one colour.”

“Switzerland has *at most* one colour.”

“Switzerland and Germany have different colours.”

$$\dots$$

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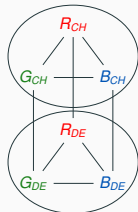
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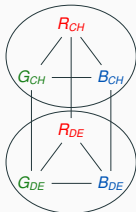
$$(\neg R_{CH} \vee \neg R_{DE}) \wedge (\neg G_{CH} \vee \neg G_{DE}) \wedge$$

“Switzerland and Germany have different colours.”

$$(\neg B_{CH} \vee \neg B_{DE})$$

...

...



...

There is a factor encoding each clause.

$r_{ch}$  is a value of  $R_{CH}$  (true or false).

$$\bigvee_{\substack{r_{ch}, g_{ch}, b_{ch} \\ r_{de}, g_{de}, b_{de}}} \psi_{R_{CH}, G_{CH}, B_{CH}}(r_{ch}, g_{ch}, b_{ch}) \wedge$$

$$\psi_{R_{CH}, B_{CH}}(r_{ch}, b_{ch}) \wedge \psi_{G_{CH}, B_{CH}}(g_{ch}, b_{ch}) \wedge \psi_{R_{CH}, G_{CH}}(r_{ch}, g_{ch}) \wedge$$

$$\psi_{R_{CH}, R_{DE}}(r_{ch}, r_{de}) \wedge \psi_{G_{CH}, G_{DE}}(g_{ch}, g_{de}) \wedge \psi_{B_{CH}, B_{DE}}(b_{ch}, b_{de}) \wedge$$

...



## SAT: Satisfiability

Check satisfiability of CNF formula  $\bigwedge_{i \in [m]} c_i$  with hypergraph  $\mathcal{H}$

FAQ encoding over the **Boolean semiring** ( $\{\text{true}, \text{false}\}, \vee, \wedge, \text{false}, \text{true}$ ):

$$\Phi() = \bigvee_{\mathbf{x}} \bigwedge_{S \in \mathcal{E}} \psi_S(\mathbf{x}_S)$$

where  $\Phi$  has the same hypergraph  $\mathcal{H} = (\mathcal{V}, \mathcal{E})$  as the CNF formula

- $\mathcal{V} = [n]$  is the set of indices of variables  $X_1, \dots, X_n$  in the CNF formula
- Each clause  $c_i$  over variables with indices in  $S$  defines a factor  $\psi_S$ 
  - For variable assignment  $\mathbf{x}_S$ ,  $\psi_S(\mathbf{x}_S)$  returns the truth of  $c_i$
  - Naïve  $O(2^{|S|})$  representation of  $\psi_S$  is the truth table of  $c_i$  over the variables in  $S$
  - Alternative  $O(|S|)$  representation is just the clause

## #SAT: Counting the Number of Satisfying Assignments

Count the satisfying assignments of CNF formula  $\bigwedge_{i \in [m]} c_i$  with hypergraph  $\mathcal{H}$

FAQ encoding over the **sum-product semiring**  $(\mathbb{N}, +, *, 0, 1)$ :

$$\Phi() = \sum_{\mathbf{x}} \prod_{S \in \mathcal{E}} \psi_S(\mathbf{x}_S)$$

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- Each clause  $c_i$  over variables with indices in  $S$  defines a factor  $\psi_S$

$$\psi_S(\mathbf{x}_S) = \begin{cases} 1 & \text{if } \mathbf{x}_S \text{ satisfies } c_i \\ 0 & \text{otherwise.} \end{cases}$$

### 3-Colorability: A Different Approach

Recall again the 3-colorability problem instance.

We use the map graph as the hypergraph of the problem:

- $\mathcal{V} = \{CH, DE, FR, IT, AT, LI, \dots\}$
- $\mathcal{E} = \{(CH, DE), (CH, FR), (CH, IT), (CH, AT), (CH, LI), \dots\}$

For each edge  $(u, v)$  there is a factor  $\psi_{uv}(c_1, c_2) = (c_1 \neq c_2)$

- Each variable can take value 1, 2, or 3 representing one of the three colours
- For a pair of colours  $(c_1, c_2)$  for nodes  $(u, v)$ ,  $\psi_{uv}$  is true if  $c_1 \neq c_2$

The FAQ is (the part for CH):

$$\Phi() = \bigvee_{\substack{c_{ch}, c_{de}, c_{fr}, \\ c_{it}, c_{at}, c_{li}, \dots}} \psi_{CH,DE}(c_{ch}, c_{de}) \wedge \psi_{CH,FR}(c_{ch}, c_{fr}) \wedge \psi_{CH,IT}(c_{ch}, c_{it}) \wedge \\ \psi_{CH,AT}(c_{ch}, c_{at}) \wedge \psi_{CH,LI}(c_{ch}, c_{li}) \wedge \dots$$

Check  $k$ -colorability for a graph  $G = (V, E)$

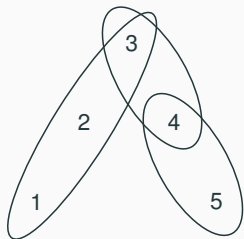
FAQ encoding over the **Boolean semiring** ( $\{\text{true}, \text{false}\}, \vee, \wedge, \text{false}, \text{true}$ ):

$$\Phi() = \bigvee_{\mathbf{x}} \bigwedge_{(u,v) \in E} \psi_{uv}(x_u, x_v), \text{ where}$$

- Every edge  $(u, v) \in E$  defines a factor  $\psi_{uv}(c_1, c_2) = (c_1 \neq c_2)$
- Every node  $v \in V$  defines a variable  $X_v$  with domain  $\text{Dom}(X_v) = [k]$
- $\Phi$  has the hypergraph with vertices  $\mathcal{V} = \{X_v | v \in V\}$  and edges  $\mathcal{E} = E$

## DB: Example (1/4)

We map our previous DB example with customers ordering dishes to FAQ:



Orders(customer, day, dish)

$\psi_{123}(x_1, x_2, x_3)$

Dishes(dish, item)

$\psi_{34}(x_3, x_4)$

Items (item, price)

$\psi_{45}(x_4, x_5)$

The FAQ over the **union-intersection semiring** to capture the join:

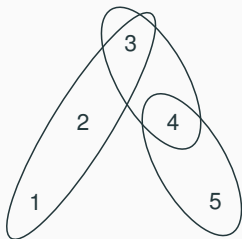
$$\Phi() = \bigcup_{x_1, x_2, x_3, x_4, x_5} \psi_{123}(x_1, x_2, x_3) \cap \psi_{34}(x_3, x_4) \cap \psi_{45}(x_4, x_5)$$

$\Phi$  maps the empty tuple to the join result.

In SQL, this join is expressed as follows:

```
SELECT * FROM Orders NATURAL JOIN Dishes NATURAL JOIN Items
```

## DB: Example (2/4)



Orders(customer, day, dish)

$\psi_{123}(x_1, x_2, x_3)$

Dishes(dish, item)

$\psi_{34}(x_3, x_4)$

Items (item, price)

$\psi_{45}(x_4, x_5)$

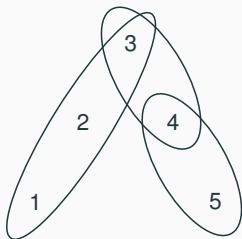
The FAQ over the **Boolean semiring** to capture the Boolean conjunctive query:

$$\Phi() = \bigvee_{x_1, x_2, x_3, x_4, x_5} \psi_{123}(x_1, x_2, x_3) \wedge \psi_{34}(x_3, x_4) \wedge \psi_{45}(x_4, x_5)$$

In SQL, this is expressed as follows:

```
SELECT true FROM Orders NATURAL JOIN Dishes NATURAL JOIN Items
```

## DB: Example (3/4)



Orders(customer, day, dish)

$\psi_{123}(x_1, x_2, x_3)$

Dishes(dish, item)

$\psi_{34}(x_3, x_4)$

Items (item, price)

$\psi_{45}(x_4, x_5)$

FAQ over the **sum-product semiring** to express a COUNT query:

$$\Phi(x_1) = \sum_{x_2, x_3, x_4, x_5} \psi_{123}(x_1, x_2, x_3) \cdot \psi_{34}(x_3, x_4) \cdot \psi_{45}(x_4, x_5)$$

Each factor maps tuples from corresponding relation to 1.

In SQL, this FAQ is expressed as follows:

```
SELECT customer, COUNT(*)  
FROM Orders NATURAL JOIN Dishes NATURAL JOIN Items  
GROUP BY customer
```

## DB: Example (4/4)

More interesting aggregates captured by appropriately defining the factors

Query: Total price per customer and day

In FAQ:

- Let  $\psi_{45}$  map  $(x_4, x_5)$  to  $x_5$  (price), all other factors map tuples to 1

$$\Phi(x_1, x_2) = \sum_{x_3, x_4, x_5} \psi_{123}(x_1, x_2, x_3) \cdot \psi_{34}(x_3, x_4) \cdot \psi_{45}(x_4, x_5)$$

In SQL:

```
SELECT customer, day, SUM(price)
FROM Orders NATURAL JOIN Dishes NATURAL JOIN Items
GROUP BY customer, day
```



## BCQ: Boolean Conjunctive Queries

Compute the Boolean query  $\exists X_1 \dots \exists X_n : \bigwedge_{R \in \text{atoms}} R(\text{vars}(R))$

- atoms is the set of relation symbols in the query, e.g., Dishes(dish, item)
- Each relation symbol  $R \in \text{atoms}$  has variables (attributes)  $\text{vars}(R)$

FAQ encoding over the **Boolean semiring** ( $\{\text{true}, \text{false}\}, \vee, \wedge, \text{false}, \text{true}$ ):

$$\Phi() = \bigvee_{\mathbf{x}} \bigwedge_{S \in \mathcal{E}} \psi_S(\mathbf{x}_S)$$

- $\Phi$  has the hypergraph  $(\mathcal{V}, \mathcal{E})$
- $\mathcal{V} = \bigcup_{R \in \text{atoms}(\Phi)} \text{vars}(R)$  and  $\mathcal{E} = \{\text{vars}(R) \mid R \in \text{atoms}(\Phi)\}$
- For  $S \in \mathcal{E}$  corresponding to relation  $R$ , there is a factor  $\psi_S$  such that

$$\psi_S(\mathbf{x}_S) = (\mathbf{x}_S \in R)$$

## Join: Natural Join Queries

Compute the natural join query  $\bowtie_{R \in \text{atoms}} R$

FAQ encoding over the **set semiring**  $(2^{\mathcal{U}}, \cup, \cap, \emptyset, \mathcal{U})$ :

$$\Phi() = \bigcup_{\mathbf{x}} \bigcap_{S \in \mathcal{E}} \psi_S(\mathbf{x}_S)$$

- $\Phi$  has the hypergraph  $(\mathcal{V}, \mathcal{E})$  and maps the empty tuple to the join result
- $\mathcal{U} = \prod_{i=1}^n \text{Dom}(X_i)$  is the set of all possible tuples
- $2^{\mathcal{U}}$  is the powerset of  $\mathcal{U}$ , i.e., the set of all possible subsets of  $\mathcal{U}$
- $\mathcal{V}$  is the set of variables (attributes) in the atoms of the join query
- For  $S \in \mathcal{E}$  corresponding to relation  $R$ , there is a factor  $\psi_S$  such that

$$\psi_S(\mathbf{x}_S) = \begin{cases} \{\mathbf{t} \mid \pi_S(\mathbf{t}) = \mathbf{x}_S\} & \text{if } \mathbf{x}_S \in R \\ \emptyset & \text{if } \mathbf{x}_S \notin R \end{cases}$$

# MCM: Matrix Chain Multiplication

Compute the matrix product  $\mathbf{A} = \mathbf{A}_1 \cdots \mathbf{A}_n$ , where  $\forall i \in [n] : \mathbf{A}_i \in \mathbb{R}^{p_i \times p_{i+1}}$

FAQ encoding over the **real sum-product semiring**  $(\mathbb{R}, +, *, 0, 1)$ :

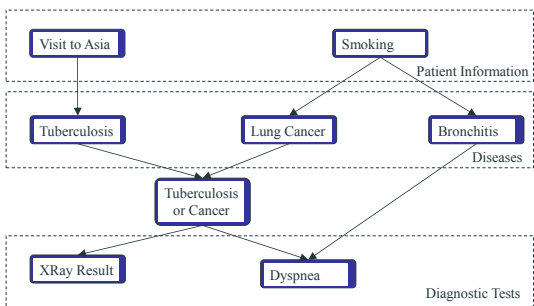
$$\Phi(x_1, x_{n+1}) = \sum_{x_2 \in \text{Dom}(X_2)} \cdots \sum_{x_n \in \text{Dom}(X_n)} \prod_{i \in [n]} \psi_{i,i+1}(x_i, x_{i+1}).$$

- We use  $n + 1$  variables  $X_1, \dots, X_{n+1}$  with domains  $\text{Dom}(X_i) = [p_i]$
- Each matrix  $\mathbf{A}_i$  can be viewed as a function of two variables:

$$\psi_{i,i+1} : \text{Dom}(X_i) \times \text{Dom}(X_{i+1}) \rightarrow \mathbb{R}, \text{ where } \psi_{i,i+1}(x_i, x_{i+1}) = (\mathbf{A}_i)_{x_i x_{i+1}}$$

One variable is the row index, the other variable is the column index

# PGM: Example



Network represents a knowledge structure that models the relationship between diseases, their causes and effects, patient information and diagnostic tests

| $A$ | $P(A)$ |
|-----|--------|
| T   | .01    |
| F   | .99    |

| $S$ | $P(S)$ |
|-----|--------|
| T   | .4     |
| F   | .6     |

| $AT$ | $P(T A)$ | $SB$ | $P(B S)$ | $SL$ | $P(L S)$ |
|------|----------|------|----------|------|----------|
| TT   | .05      | TT   | .6       | TT   | .1       |
| TF   | .95      | TF   | .4       | TF   | .9       |
| FT   | .01      | FT   | .3       | FT   | .01      |
| FF   | .99      | FF   | .7       | FF   | .99      |

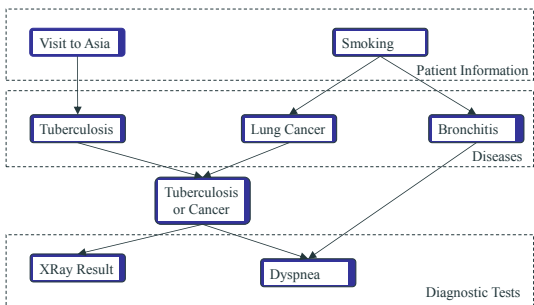
| $TLO$ | $P(O T, L)$ | $OBD$ | $P(D O, B)$ |
|-------|-------------|-------|-------------|
| TTT   | 1           | TTT   | .9          |
| TTF   | 0           | TTF   | .1          |
| TFT   | 1           | TFT   | .7          |
| TFB   | 0           | TFB   | .3          |
| FTT   | 1           | FTT   | .8          |
| FTB   | 0           | FTB   | .2          |
| FFT   | 0           | FFT   | .1          |
| FFB   | 1           | FFB   | .9          |

| $O$ | $X$ | $P(X O)$ |
|-----|-----|----------|
| T   | T   | .98      |
| T   | F   | .02      |
| F   | T   | .05      |
| F   | F   | .95      |

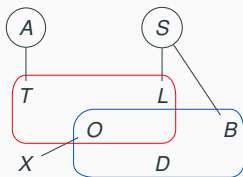
Variable  $O$ :  
Tuberculosis or Cancer

## PGM: Example

Hypergraph: one node per random variable and one hyperedge per conditional probability table



Network represents a knowledge structure that models the relationship between diseases, their causes and effects, patient information and diagnostic tests



One factor  $\psi_S$  for each conditional probability table over variables  $S$

- Maps tuple  $x_S$  of values of variables  $S$  to the associated probability
- Factors:  $\psi_A, \psi_{A,T}, \psi_S, \psi_{S,L}, \psi_{S,B}, \psi_{T,L,O}, \psi_{O,X}, \psi_{O,B,D}$

## PGM: Example

The original joint probability distribution

$$P(A, T, S, L, B, O, X, D) = P(A) \cdot P(T|A) \cdot P(S) \cdot P(L|S) \cdot P(B|S) \cdot P(O|T, L) \cdot P(X|O) \cdot P(D|O, B)$$

## PGM: Example

The original joint probability distribution and the corresponding FAQ encoding

$$\underbrace{P(A, T, S, L, B, O, X, D)}_{\phi} = \underbrace{P(A)}_{\psi_A} \cdot \underbrace{P(T|A)}_{\psi_{A,T}} \cdot \underbrace{P(S)}_{\psi_S} \cdot \underbrace{P(L|S)}_{\psi_{S,L}} \cdot \underbrace{P(B|S)}_{\psi_{S,B}} \cdot \underbrace{P(O|T,L)}_{\psi_{T,L,O}} \cdot \underbrace{P(X|O)}_{\psi_{O,X}} \cdot \underbrace{P(D|O,B)}_{\psi_{O,B,D}}$$

The original joint probability distribution and the corresponding FAQ encoding

$$\underbrace{P(A, T, S, L, B, O, X, D)}_{\Phi} = \underbrace{P(A)}_{\psi_A} \cdot \underbrace{P(T|A)}_{\psi_{A,T}} \cdot \underbrace{P(S)}_{\psi_S} \cdot \underbrace{P(L|S)}_{\psi_{S,L}} \cdot \underbrace{P(B|S)}_{\psi_{S,B}} \cdot \underbrace{P(O|T, L)}_{\psi_{T,L,O}} \cdot \underbrace{P(X|O)}_{\psi_{O,X}} \cdot \underbrace{P(D|O, B)}_{\psi_{O,B,D}}$$

Marginal Distribution  $P(A, B, D)$  using the sum-product semiring:

$$\begin{aligned} \Phi(a, b, d) = \sum_{t,s,l,o,x} & \psi_A(a) \cdot \psi_{A,T}(a, t) \cdot \psi_S(s) \cdot \psi_{S,L}(s, l) \cdot \psi_{S,B}(s, b) \cdot \\ & \psi_{T,L,O}(t, l, o) \cdot \psi_{O,X}(o, x) \cdot \psi_{O,B,D}(o, b, d) \end{aligned}$$

Maximum A-Posteriori  $P(A, B, D)$  using the max-product semiring:

$$\begin{aligned} \Phi(a, b, d) = \max_{t,s,l,o,x} & \psi_A(a) \cdot \psi_{A,T}(a, t) \cdot \psi_S(s) \cdot \psi_{S,L}(s, l) \cdot \psi_{S,B}(s, b) \cdot \\ & \psi_{T,L,O}(t, l, o) \cdot \psi_{O,X}(o, x) \cdot \psi_{O,B,D}(o, b, d) \end{aligned}$$



# MAP-PGM: Maximum A-Posteriori in Probabilistic Graphical Models

Compute the MAP estimate over variables  $X_1, \dots, X_f$

FAQ encoding over the **max-product semiring** ( $[0, \infty)$ ,  $\max, *, 0, 1$ )

$$\Phi(x_1, \dots, x_f) = \max_{x_{f+1} \in \text{Dom}(X_{f+1})} \cdots \max_{x_n \in \text{Dom}(X_n)} \prod_{S \in \mathcal{E}} \psi_S(\mathbf{x}_S)$$

where  $(\mathcal{V}, \mathcal{E})$  is the hypergraph of the (undirected) probabilistic graphical model

- $\mathcal{V} = [n]$ : indices of  $n$  discrete random variables  $X_1, \dots, X_n$
- There is a factor  $\psi_S : \prod_{i \in S} \text{Dom}(X_i) \rightarrow [0, \infty)$  for each edge  $S \in \mathcal{E}$

# MD-PGM: Marginal Distribution in Probabilistic Graphical Models

Compute the marginal distribution of the set of variables  $X_1, \dots, X_f$

FAQ encoding over the **sum-product semiring**  $(\mathbb{R}_+, +, *, 0, 1)$

$$\Phi(x_1, \dots, x_f) = \sum_{x_{f+1} \in \text{Dom}(X_{f+1})} \cdots \sum_{x_n \in \text{Dom}(X_n)} \prod_{S \in \mathcal{E}} \psi_S(\mathbf{x}_S)$$

where  $(\mathcal{V}, \mathcal{E})$  is the hypergraph of the (undirected) probabilistic graphical model

- $\mathcal{V} = [n]$ : indices of  $n$  discrete random variables  $X_1, \dots, X_n$
- There is a factor  $\psi_S : \prod_{i \in S} \text{Dom}(X_i) \rightarrow \mathbb{R}_+$  for each edge  $S \in \mathcal{E}$

For conditional distributions  $P(\mathbf{X}_A \mid \mathbf{X}_B = \mathbf{x}_B)$ , we set  $\mathbf{X}_B$  to  $\mathbf{x}_B$ .

## Sample of Problems Expressible in FAQ

## Boolean Semiring

$(\{\text{true}, \text{false}\}, \vee, \wedge, \text{false}, \text{true})$

- Constraint satisfaction problems (CSP) [FAQ]
- Boolean conjunctive query evaluation (BCQ) [FAQ]
- Conjunctive query evaluation (CQ)\* [FAQ]
- Join evaluation [FAQ]
- Satisfiability (SAT) [FAQ]
- $k$ -colorability [FAQ]
- List recovery problem (coding theory) [FAQ]

(\*) also expressible using the set semiring

## Set and Natural Sum-Product Semirings

$(2^{\mathcal{U}}, \cup, \cap, \emptyset, \mathcal{U})$

- Conjunctive query evaluation (CQ)\* [FAQ]
- Join evaluation [FAQ]

$(\mathbb{N}, +, *, 0, 1)$

- Complex network analysis [FAQ]
- Count constraint satisfaction problems ( $\#$ CSP) [FAQ]
- Count satisfiability ( $\#$ SAT) [FAQ]

(\*) also expressible using the Boolean semiring

## Real Sum-Product Semiring

$(\mathbb{R}, +, *, 0, 1)$

- Permanent [FAQ]
- Discrete Fourier transform [FAQ,AjiMcEI]
- Hadamard transform [AjiMcEI]
- Inference in probabilistic graphical models [FAQ]
- Probability propagation in AI [AjiMcEI]
- Matrix chain multiplication [FAQ,AjiMcEI]
- Graph homomorphism [FAQ]
- BCJR decoding (Bahl, Cocke, Jelinek, Raviv) [AjiMcEI]
- Holant problem [FAQ]

## Max-Product and Min-Sum Semirings

$([0, \infty), \max, *, 0, 1)$

- MAP queries in probabilistic graphical models [FAQ]
- Quantified conjunctive query evaluation (QCQ)\* [FAQ]

$((-\infty, \infty], \min, +, \infty, 0)$

- Gallager-Tanner-Wiberg decoding [AjiMcEl]
- Viterbi decoding [AjiMcEl]
- Trellis path problem [AjiMcEl]
- Graph optimization [KohIWils]
- Queuing systems [KohIWils]
- Discrete event systems [KohIWils]
- Optimization for weighted CSPs [KohIWils]

(\*) also expressible using the max-product, min-product semirings

## Problems Expressible With Two Semirings

$([0, \infty), \max, *, 0, 1)$ ,  $((0, \infty], \min, *, \infty, 1)$

- Quantified conjunctive query evaluation (QCQ)\* [FAQ]

$(\mathbb{N}, \max, *, 0, 1)$ ,  $(\mathbb{N}, +, *, 0, 1)$

- Count conjunctive query evaluation (#CQ) [FAQ]
- Count quantified conjunctive query evaluation (#QCQ) [FAQ]

(\*) also expressible using the max-product semiring