



Uncertain Data Management

Relational Probabilistic Database Models

Antoine Amarilli¹, Silviu Maniu²

November 28th, 2017

¹Télécom ParisTech

²LRI

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Uncertain instances

Remember from **last class**:

- Fix a finite set of **possible tuples** of same arity
- A **possible world**: a subset of the **possible tuples**
- A (finite) **uncertain relation**: set of **possible worlds**

Uncertain instances

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U_1			U_2		
date	teacher	room	date	teacher	room
04	Silviu	C017	04	Silviu	C017
04	Antoine	C017	04	Antoine	C017
04	Antoine	C47	04	Antoine	C47
11	Silviu	C017	11	Silviu	C017
11	Silviu	C47	11	Silviu	C47
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Probabilistic instances

- Support \mathcal{U} : uncertain relation

Probabilistic instances

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 - It must sum up to 1: $\sum_{l \in \mathcal{U}} \pi(l) = 1$

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$\pi(U_1) = 0.8$ $\pi(U_2) = 0.2$

What about NULLs?

Remember that last time we saw:

- Codd-tables and v-tables and c-tables, with **NULLs**
- Boolean c-tables, with **NULLs** only in conditions
 - Boolean variables

What about NULLs?

Remember that last time we saw:

- Codd-tables and v-tables and c-tables, with **NULLs**
 - Boolean c-tables, with **NULLs** only in conditions
 - Boolean variables
- We focus for probabilities on models like Boolean c-tables
- Easier to define probabilities on a finite space!

Relational algebra on uncertain instances

Remember from **last class**:

- Extend relational algebra operators to **uncertain instances**
- The **possible worlds** of the **result** should be...
 - take all **possible worlds** in the supports of the inputs
 - apply the operation and get the **possible outputs**

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U_1

04	S.	CO17
11	S.	C47

U_2

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----	----	------

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U_1		V_1
04 S. CO17		
11 S. C47	\cup	
U_2		V_2
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04 S. CO17			
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\cup			$=$
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04 S. Co17				04 S. Co17	
11 S. C47				11 S. C47	
U_2		V_2			
11 A. Co17		11 A. Co17		11 A. Co17	
	\cup		$=$	04 S. Co17	
				11 S. C47	
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Relational algebra on probabilistic instances

- Let's adapt relational algebra to **probabilistic instances**
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Relational algebra on probabilistic instances

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- The **possible worlds** of the **result** should be...
 - take all **possible worlds** of the inputs
 - apply the operation and get a **possible output**
- The **probability** of each possible world should be...
 - consider **all input possible worlds** that give it
 - sum up their **probabilities**

Example of relational algebra on probabilistic instances

Example of relational algebra on probabilistic instances

U_1

04	S.	C017
----	----	------

11	S.	C47
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U_2

11	A.	C017
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\cup

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V_2

11	A.	CO17
----	----	------

Example of relational algebra on probabilistic instances

U_1	
04 S. C017	
11 S. C47	
$\pi(U_1) = 0.8$	

U

V_1	
$\pi(V_1) = 0.9$	

U_2	
11 A. C017	
$\pi(U_2) = 0.2$	

V_2	
11 A. C017	
$\pi(V_2) = 0.1$	

Example of relational algebra on probabilistic instances

$$\begin{array}{c} \hline U_1 \\ \hline 04 \quad S. \quad C017 \\ 11 \quad S. \quad C47 \\ \hline \pi(U_1) = 0.8 \end{array} \cup \begin{array}{c} \hline V_1 \\ \hline \\ \hline \pi(V_1) = 0.9 \end{array} = \begin{array}{c} \hline U_2 \\ \hline 11 \quad A. \quad C017 \\ \hline \pi(U_2) = 0.2 \end{array} \begin{array}{c} \hline V_2 \\ \hline 11 \quad A. \quad C017 \\ \hline \pi(V_2) = 0.1 \end{array}$$

Example of relational algebra on probabilistic instances

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$\pi(U_1) = 0.8$		V_2		$\pi(W_1) = 0.8 \times 0.9$
U_2		11 A. C017		W_2
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U_1		V_1		W_1
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$\pi(U_1) = 0.8$	U	$\pi(V_1) = 0.9$	=	$\pi(W_1) = 0.8 \times 0.9$
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				W_3
				11 A. C017 $\pi(W_3) = 0.2 \times 0.9$

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$\pi(U_2) = 0.2$		11 A. C017		11 A. C017
		$\pi(V_2) = 0.1$		$\pi(W_2) = 0.8 \times 0.1$
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				11 A. C017
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				$+0.2 \times 0.1$

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→ there are 2^N possible instances

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- For probabilistic instances:
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- For probabilistic instances:
 - there are infinitely many possible instances
 - writing out a probabilistic instance is still exponential
- How to represent probabilistic instances?

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Tuple-independent databases

- The **simplest** model: tuple-independent databases
- Annotate each **instance fact** with a **probability**

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Tuple-independent databases

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Tuple-independent databases

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- Assume **independence** between tuples
(Silviu and Antoine may teach at the same time)

Semantics of TID

- Each tuple is **kept** or **discarded** with the probability
- Probabilistic choices are **independent** across tuples

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date	teacher	room	
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What's the **probability** of this outcome?

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0.8

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What's the **probability** of this outcome?

$$0.8 \times$$

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What's the **probability** of this outcome?

$$0.8 \times (1 - 0.2)$$

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What's the **probability** of this outcome?

$$0.8 \times (1 - 0.2) \times 1$$

Getting a probability distribution

The semantics of a TID instance is a probabilistic instance...

→ the possible worlds are the subsets

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Formally, for a TID instance I , the probability of J :

Getting a probability distribution

The semantics of a TID instance is a probabilistic instance...

- the possible worlds are the subsets
 - always keeping tuples with probability 1

Formally, for a TID instance I , the probability of J :

- we must have $J \subseteq I$
- product of $p_{\mathbf{t}}$ for each tuple \mathbf{t} kept in J
- product of $1 - p_{\mathbf{t}}$ for each tuple \mathbf{t} not kept in J

Is it a probability distribution?

Do the probabilities always sum to 1?

- Let N be the number of tuples
 - There are 2^N possible worlds
 - They are all products of p_i or $1 - p_i$ for each $1 \leq i \leq N$
- This is the result of expanding the expression:
- $$(p_1 + (1 - p_1)) \times \cdots \times (p_n + (1 - p_n))$$
- All factors are equal to 1, so the probabilities sum to 1

Strong representation system

Remember from **last class**:

Uncertain instance: set of possible worlds

Strong representation system

Remember from **last class**:

Uncertain instance: set of possible worlds

Uncertainty framework: concise way to represent uncertain instances

Strong representation system

Remember from **last class**:

Uncertain instance: set of possible worlds

Uncertainty framework: concise way to represent
uncertain instances

Query language: here, relational algebra

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Definition (Strong representation system)

For any query in the language,

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Uncertainty framework: concise way to represent uncertain instances

Query language: here, relational algebra

Definition (Strong representation system)

*For any query in the language,
on uncertain instances represented in the framework,*

Strong representation system

Remember from **last class**:

Uncertain instance: set of possible worlds

Uncertainty framework: concise way to represent uncertain instances

Query language: here, relational algebra

Definition (Strong representation system)

*For any query in the language,
on uncertain instances represented in the framework,
the uncertain instance obtained by evaluating the query*

Strong representation system

Remember from **last class**:

Uncertain instance: set of possible worlds

Uncertainty framework: concise way to represent uncertain instances

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Definition (Strong representation system)

*For any query in the language,
on uncertain instances represented in the framework,
the uncertain instance obtained by evaluating the query
can also be represented in the framework*

Strong representation system

Remember from **last class**:

Uncertain instance: set of possible worlds

Uncertainty framework: concise way to represent uncertain instances

Query language: here, relational algebra

Definition (Strong representation system)

*For any query in the language,
on uncertain instances represented in the framework,
the uncertain instance obtained by evaluating the query
can also be represented in the framework*

→ Are TID instances a **strong representation system**?

Implementing select

U

date	teacher	room	
04	Silviu	C47	0.8
04	Antoine	C47	0.2
11	Silviu	C47	1

Implementing select

U

date	teacher	room	
04	Silviu	C47	0.8
04	Antoine	C47	0.2
11	Silviu	C47	1

$\sigma_{\text{teacher}=\text{"Silviu"}}(U)$

date	teacher	room
-------------	----------------	-------------

Implementing select

U

date	teacher	room	
04	Silviu	C47	0.8
04	Antoine	C47	0.2
11	Silviu	C47	1

$\sigma_{\text{teacher}=\text{"Silviu"}}(U)$

date	teacher	room	
04	Silviu	C47	0.8
11	Silviu	C47	1

Implementing select

U

date	teacher	room	
04	Silviu	C47	0.8
04	Antoine	C47	0.2
11	Silviu	C47	1

$\sigma_{\text{teacher}=\text{"Silviu"}}(U)$

date	teacher	room	
04	Silviu	C47	0.8
11	Silviu	C47	1

→ Is this correct? ...

Implementing select

U

date	teacher	room	
04	Silviu	C47	0.8
04	Antoine	C47	0.2
11	Silviu	C47	1

$\sigma_{\text{teacher}=\text{"Silviu"}}(U)$

date	teacher	room	
04	Silviu	C47	0.8
11	Silviu	C47	1

→ Is this correct? ... So far, so good.

Implementing project

U

date	teacher	room	
04	Silviu	C47	0.8
04	Antoine	C47	0.2
11	Silviu	C47	1
11	Antoine	C47	0.1
18	Silviu	C47	0.9

Implementing project

U

date	teacher	room	
04	Silviu	C47	0.8
04	Antoine	C47	0.2
11	Silviu	C47	1
11	Antoine	C47	0.1
18	Silviu	C47	0.9

$\pi_{\text{date}}(U)$

date

Implementing project

U

date	teacher	room	
04	Silviu	C47	0.8
04	Antoine	C47	0.2
11	Silviu	C47	1
11	Antoine	C47	0.1
18	Silviu	C47	0.9

$\pi_{\text{date}}(U)$

date
04
11
18

Implementing project

U

date	teacher	room	
04	Silviu	C47	0.8
04	Antoine	C47	0.2
11	Silviu	C47	1
11	Antoine	C47	0.1
18	Silviu	C47	0.9

$\pi_{\text{date}}(U)$

date	
04	
11	
18	0.9

Implementing project

U

date	teacher	room	
04	Silviu	C47	0.8
04	Antoine	C47	0.2
11	Silviu	C47	1
11	Antoine	C47	0.1
18	Silviu	C47	0.9

$\pi_{\text{date}}(U)$

date	
04	
11	1
18	0.9

Implementing project

U

date	teacher	room	
04	Silviu	C47	0.8
04	Antoine	C47	0.2
11	Silviu	C47	1
11	Antoine	C47	0.1
18	Silviu	C47	0.9

$\pi_{\text{date}}(U)$

date	
04	$1 - (1 - 0.2) \times (1 - 0.8)$
11	1
18	0.9

Implementing project

U

date	teacher	room	
04	Silviu	C47	0.8
04	Antoine	C47	0.2
11	Silviu	C47	1
11	Antoine	C47	0.1
18	Silviu	C47	0.9

$\pi_{\text{date}}(U)$

date	
04	$1 - (1 - 0.2) \times (1 - 0.8)$
11	1
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→ Is this correct? ...

Implementing project

U

date	teacher	room	
04	Silviu	C47	0.8
04	Antoine	C47	0.2
11	Silviu	C47	1
11	Antoine	C47	0.1
18	Silviu	C47	0.9

$\pi_{\text{date}}(U)$

date	
04	$1 - (1 - 0.2) \times (1 - 0.8)$
11	1
18	0.9

→ Is this correct? ... So far, so good.

Implementing join

U

date	teacher	room	
04	Silviu	C47	0.8
04	Antoine	C47	0.2

Implementing join

U

date	teacher	room	
04	Silviu	C47	0.8
04	Antoine	C47	0.2

Repair

room	cause	
C47	leopard	0.1

Implementing join

U

date	teacher	room	
04	Silviu	C47	0.8
04	Antoine	C47	0.2

Repair

room	cause
C47	leopard 0.1

$U \bowtie \text{Repair}$

date	teacher	room	cause
-------------	----------------	-------------	--------------

Implementing join

U

date	teacher	room	
04	Silviu	C47	0.8
04	Antoine	C47	0.2

Repair

room	cause
C47	leopard 0.1

$U \bowtie \text{Repair}$

date	teacher	room	cause
04	Silviu	C47	leopard
04	Antoine	C47	leopard

Implementing join

U

date	teacher	room	
04	Silviu	C47	0.8
04	Antoine	C47	0.2

Repair

room	cause
C47	leopard 0.1

$U \bowtie \text{Repair}$

date	teacher	room	cause	
04	Silviu	C47	leopard	0.8×0.1
04	Antoine	C47	leopard	0.2×0.1

Implementing join

U

date	teacher	room	
04	Silviu	C47	0.8
04	Antoine	C47	0.2

Repair

room	cause
C47	leopard 0.1

$U \bowtie \text{Repair}$

date	teacher	room	cause	
04	Silviu	C47	leopard	0.8×0.1
04	Antoine	C47	leopard	0.2×0.1

→ Is this correct?

Implementing join ... **OR NOT!**

U				Repair	
date	teacher	room		room	cause
04	Silviu	C47	0.8	C47	leopard 0.1
04	Antoine	C47	0.2		

$U \bowtie \text{Repair}$

date	teacher	room	cause	
04	Silviu	C47	leopard	0.8×0.1
04	Antoine	C47	leopard	0.2×0.1

→ Is this correct?

→ It's **WRONG!**

Why is it wrong?

U

date	teacher	room	
04	Silviu	C47	1
04	Antoine	C47	1

Why is it wrong?

U

date	teacher	room	
04	Silviu	C47	1
04	Antoine	C47	1

Repair

room	cause	
C47	leopard	1/2

Why is it wrong?

U

date	teacher	room	
04	Silviu	C47	1
04	Antoine	C47	1

Repair

room	cause	
C47	leopard	1/2

$U \bowtie \text{Repair}$

date	teacher	room	cause
04	Silviu	C47	leopard
04	Antoine	C47	leopard

Why is it wrong?

U

date	teacher	room	
04	Silviu	C47	1
04	Antoine	C47	1

Repair

room	cause	
C47	leopard	1/2

$U \bowtie \text{Repair}$

date	teacher	room	cause	
04	Silviu	C47	leopard	1/2
04	Antoine	C47	leopard	1/2

Why is it wrong?

U				Repair		
date	teacher	room		room	cause	
04	Silviu	C47	1	C47	leopard	1/2
04	Antoine	C47	1			

$U \bowtie \text{Repair}$

date	teacher	room	cause	
04	Silviu	C47	leopard	1/2
04	Antoine	C47	leopard	1/2

- The two tuples are **not independent!**
- The first is there **iff** the second is there.

Why does it matter?

$U \bowtie \text{Repair}$

date	teacher	room	cause	
04	Silviu	C47	leopard	1/2
04	Antoine	C47	leopard	1/2

Why does it matter?

$U \bowtie \text{Repair}$

date	teacher	room	cause	
04	Silviu	C47	leopard	1/2
04	Antoine	C47	leopard	1/2

$\pi_{\text{room}}(U \bowtie \text{Repair})$

room

C47

Why does it matter?

$U \bowtie \text{Repair}$

date	teacher	room	cause	
04	Silviu	C47	leopard	1/2
04	Antoine	C47	leopard	1/2

$\pi_{\text{room}}(U \bowtie \text{Repair})$

room	
C47	$1 - (1 - 1/2) \times (1 - 1/2)$

Why does it matter?

$U \bowtie \text{Repair}$

date	teacher	room	cause	
04	Silviu	C47	leopard	1/2
04	Antoine	C47	leopard	1/2

$\pi_{\text{room}}(U \bowtie \text{Repair})$

room	
C47	$1 - (1 - 1/2) \times (1 - 1/2)$

→ Probability of 3/4...

Why does it matter?

$U \bowtie \text{Repair}$

date	teacher	room	cause	
04	Silviu	C47	leopard	1/2
04	Antoine	C47	leopard	1/2

$\pi_{\text{room}}(U \bowtie \text{Repair})$

room	
C47	$1 - (1 - 1/2) \times (1 - 1/2)$

→ Probability of 3/4...

→ But the leopard had probability 1/2!

TID are not a strong representation system

- Remember how **Codd tables** required **named nulls**?
 - The result of a query on TID may **not** be a TID
- We will see that the correlations can be **complex**

TID are not a strong representation system

- Remember how **Codd tables** required **named nulls**?
- The result of a query on TID may **not** be a TID
- We will see that the correlations can be **complex**

- How to **evaluate** queries on a TID then?
- List all **possible worlds** and count the probabilities

Query evaluation done right

U

date	teacher	room	
04	Silviu	C47	0.8
04	Antoine	C47	0.2

Query evaluation done right

U

date	teacher	room	
04	Silviu	C47	0.8
04	Antoine	C47	0.2

Repair

room	cause
C47	leopard 0.1

Query evaluation done right

U				Repair	
date	teacher	room		room	cause
04	Silviu	C47	0.8	C47	leopard 0.1
04	Antoine	C47	0.2		

$\pi_{\text{room}}(U \times \text{Repair})$

room

C47

Query evaluation done right

U				Repair	
date	teacher	room		room	cause
04	Silviu	C47	0.8	C47	leopard 0.1
04	Antoine	C47	0.2		

$\pi_{\text{room}}(U \times \text{Repair})$

room
C47 ???

Query evaluation done right

U				Repair	
date	teacher	room		room	cause
04	Silviu	C47	0.8	C47	leopard 0.1
04	Antoine	C47	0.2		

$$\pi_{\text{room}}(U \times \text{Repair})$$

room
C47 ???

- Either there is no leopard and then no result...

Query evaluation done right

U				Repair	
date	teacher	room		room	cause
04	Silviu	C47	0.8	C47	leopard 0.1
04	Antoine	C47	0.2		

$$\pi_{\text{room}}(U \times \text{Repair})$$

room
C47 ???

- Either there is no leopard and then no result...
- Or there is a leopard and then...

Query evaluation done right

U				Repair	
date	teacher	room		room	cause
04	Silviu	C47	0.8	C47	leopard 0.1
04	Antoine	C47	0.2		

$$\pi_{\text{room}}(U \times \text{Repair})$$

room
C47 ???

- Either there is no leopard and then no result...
- Or there is a leopard and then...
 - Non-empty result:

Query evaluation done right

U				Repair	
date	teacher	room		room	cause
04	Silviu	C47	0.8	C47	leopard 0.1
04	Antoine	C47	0.2		

$$\pi_{\text{room}}(U \times \text{Repair})$$

room
C47 ???

- Either there is no leopard and then no result...
- Or there is a leopard and then...
 - Non-empty result: $1 - (1 - 0.8) \times (1 - 0.2)$

Query evaluation done right

U				Repair	
date	teacher	room		room	cause
04	Silviu	C47	0.8	C47	leopard 0.1
04	Antoine	C47	0.2		

$$\pi_{\text{room}}(U \times \text{Repair})$$

room
C47 ???

- Either there is no leopard and then no result...
- Or there is a leopard and then...
 - Non-empty result: $1 - (1 - 0.8) \times (1 - 0.2) = 0.84$

Query evaluation done right

U				Repair	
date	teacher	room		room	cause
04	Silviu	C47	0.8	C47	leopard 0.1
04	Antoine	C47	0.2		

$$\pi_{\text{room}}(U \times \text{Repair})$$

room
C47

- Either there is no leopard and then no result...
- Or there is a leopard and then...
 - Non-empty result: $1 - (1 - 0.8) \times (1 - 0.2) = 0.84$
- The query probability is:

Query evaluation done right

U				Repair	
date	teacher	room		room	cause
04	Silviu	C47	0.8	C47	leopard 0.1
04	Antoine	C47	0.2		

$$\pi_{\text{room}}(U \times \text{Repair})$$

$\pi_{\text{room}}(U \times \text{Repair})$	
room	
C47	0.084

- Either there is no leopard and then no result...
- Or there is a leopard and then...
 - Non-empty result: $1 - (1 - 0.8) \times (1 - 0.2) = 0.84$
- The query probability is: 0.1×0.84

Expressiveness of TID

Can we represent **all** probabilistic instances with TID?

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*“The class is taught by Antoine or Silviu or no one but **not both**”*

Expressiveness of TID

Can we represent **all** probabilistic instances with TID?

*“The class is taught by Antoine or Silviu or no one but **not both**”*

U_1

teacher

Silviu

$\pi(U_1) = 0.8$

Expressiveness of TID

Can we represent **all** probabilistic instances with TID?

*“The class is taught by Antoine or Silviu or no one but **not both**”*

U_1	U_2
<hr/>	<hr/>
teacher	teacher
<hr/>	<hr/>
Silviu	Antoine
<hr/>	<hr/>
$\pi(U_1) = 0.8$	$\pi(U_2) = 0.1$

Expressiveness of TID

Can we represent **all** probabilistic instances with TID?

*“The class is taught by Antoine or Silviu or no one but **not both**”*

U_1	U_2	U_3
teacher	teacher	teacher
Silviu	Antoine	
$\pi(U_1) = 0.8$	$\pi(U_2) = 0.1$	$\pi(U_3) = 0.1$

Expressiveness of TID

Can we represent **all** probabilistic instances with TID?

*“The class is taught by Antoine or Silviu or no one but **not both**”*

U_1	U_2	U_3
<hr/>	<hr/>	<hr/>
teacher	teacher	teacher
<hr/>	<hr/>	<hr/>
Silviu	Antoine	
<hr/>	<hr/>	<hr/>
$\pi(U_1) = 0.8$	$\pi(U_2) = 0.1$	$\pi(U_3) = 0.1$

U
<hr/>
teacher
<hr/>
Antoine
Silviu
<hr/>

Expressiveness of TID

Can we represent **all** probabilistic instances with TID?

*“The class is taught by Antoine or Silviu or no one but **not both**”*

U_1	U_2	U_3
<hr/>	<hr/>	<hr/>
teacher	teacher	teacher
<hr/>	<hr/>	<hr/>
Silviu	Antoine	
<hr/>	<hr/>	<hr/>
$\pi(U_1) = 0.8$	$\pi(U_2) = 0.1$	$\pi(U_3) = 0.1$

U
<hr/>
teacher
<hr/>
Antoine 0.1
Silviu
<hr/>

Expressiveness of TID

Can we represent **all** probabilistic instances with TID?

*“The class is taught by Antoine or Silviu or no one but **not both**”*

U_1	U_2	U_3
teacher	teacher	teacher
Silviu	Antoine	
$\pi(U_1) = 0.8$	$\pi(U_2) = 0.1$	$\pi(U_3) = 0.1$

U	
teacher	
Antoine	0.1
Silviu	0.8

Expressiveness of TID

Can we represent **all** probabilistic instances with TID?

*“The class is taught by Antoine or Silviu or no one but **not both**”*

U_1	U_2	U_3
teacher	teacher	teacher
Silviu	Antoine	
$\pi(U_1) = 0.8$	$\pi(U_2) = 0.1$	$\pi(U_3) = 0.1$

U	
teacher	
Antoine	0.1
Silviu	0.8

→ We **cannot** forbid that both teach the class!

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Conclusion

Block-independent disjoint instances

- A more expressive framework than TID
- Call some attributes the **key** (underlined)

Block-independent disjoint instances

- A more expressive framework than TID
- Call some attributes the **key** (underlined)

U

<u>mon</u>	<u>day</u>	teacher	room
Jan	04	Silviu	Co17
Jan	04	Antoine	Co17
Jan	11	Silviu	C47
Jan	11	Antoine	Co17

Block-independent disjoint instances

- A more expressive framework than TID
- Call some attributes the **key** (underlined)

U

<u>mon</u>	<u>day</u>	teacher	room
Jan	04	Silviu	Co17
Jan	04	Antoine	Co17
Jan	11	Silviu	C47
Jan	11	Antoine	Co17

- The **blocks** are the sets of tuples with the same key

Block-independent disjoint instances

- A more expressive framework than TID
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Jan	04	Antoine	Co17
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Jan	11	Antoine	Co17

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- Each **tuple** has a probability

Block-independent disjoint instances

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<u>mon</u>	<u>day</u>	teacher	room	
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- Each **tuple** has a probability

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Jan	04	Antoine	Co17	0.1
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- The **blocks** are the sets of tuples with the same key
- Each **tuple** has a probability
- Probabilities must **sum** to ≤ 1 in each **block**

BID semantics

U

<u>mon</u>	<u>day</u>	teacher	room	
Jan	04	Silviu	C017	0.9
Jan	04	Antoine	C017	0.1
Jan	11	Silviu	C47	0.8
Jan	11	Antoine	C017	0.1

BID semantics

U

<u>mon</u>	<u>day</u>	teacher	room	
Jan	04	Silviu	C017	0.9
Jan	04	Antoine	C017	0.1
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Jan	11	Antoine	C017	0.1

- For each **block**:

BID semantics

U

<u>mon</u>	<u>day</u>	teacher	room	
Jan	04	Silviu	C017	0.9
Jan	04	Antoine	C017	0.1
Jan	11	Silviu	C47	0.8
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- For each **block**:
 - Pick **one** tuple according to probabilities

BID semantics

U

<u>mon</u>	<u>day</u>	teacher	room	
Jan	04	Silviu	C017	0.9
Jan	04	Antoine	C017	0.1
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- For each **block**:
 - Pick **one** tuple according to probabilities
 - Possibly **no** tuple if probabilities are < 1

BID semantics

U

<u>mon</u>	<u>day</u>	teacher	room	
Jan	04	Silviu	C017	0.9
Jan	04	Antoine	C017	0.1
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- For each **block**:
 - Pick **one** tuple according to probabilities
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- Do choices **independently** in each block

BID semantics

U

<u>mon</u>	<u>day</u>	teacher	room	
Jan	04	Silviu	C017	0.9
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Jan	11	Antoine	C017	0.1

U

<u>mon</u>	<u>day</u>	teacher	room	
------------	------------	---------	------	--

- For each **block**:
 - Pick **one** tuple according to probabilities
 - Possibly **no** tuple if probabilities are < 1

→ Do choices **independently** in each block

BID semantics

U

<u>mon</u>	<u>day</u>	teacher	room	
Jan	04	Silviu	C017	0.9
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Jan	11	Antoine	C017	0.1

U

<u>mon</u>	<u>day</u>	teacher	room	
Jan	04	Silviu	C017	
Jan	04	Antoine	C017	

- For each **block**:
 - Pick **one** tuple according to probabilities
 - Possibly **no** tuple if probabilities are < 1

→ Do choices **independently** in each block

BID semantics

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<u>mon</u>	<u>day</u>	teacher	room	
Jan	04	Silviu	Co17	0.9
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Jan	11	Silviu	C47	0.8
Jan	11	Antoine	Co17	0.1

U

<u>mon</u>	<u>day</u>	teacher	room	
Jan	04	Silviu	Co17	
Jan	04	Antoine	Co17	
Jan	11	Silviu	C47	
Jan	11	Antoine	Co17	

- For each **block**:
 - Pick **one** tuple according to probabilities
 - Possibly **no** tuple if probabilities are < 1

→ Do choices **independently** in each block

BID captures TID

- Each TID can be expressed as a BID...

BID captures TID

- Each TID can be expressed as a BID...
 - Take all attributes as key
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Expressiveness of BID

Can we represent **all** probabilistic instances with BID?

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U_1
teacher
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Fabian
$\pi(U_1) = 0.8$

Expressiveness of BID

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U_1	U_2
teacher	teacher
Silviu	Antoine
Fabian	Fabian
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Expressiveness of BID

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U_1	U_2	U_3
teacher	teacher	teacher
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- If **teacher** is a key teacher, then TID
- If **teacher** is not a key, then **only one tuple**
- We **cannot represent** this probabilistic instance as a BID
- It is not a **strong representation system** either
 - Same counterexample as for TID

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Boolean c-tables

Remember Boolean c-tables:

- Set of Boolean variables x_1, x_2, \dots
- Each tuple has a condition: Variables, Boolean operators

Boolean c-tables

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x_1 Silviu is sick

x_2 Projector in C017 is working

pc-tables

A (Boolean) *pc-table* is a Boolean *c-table* plus a *probability* p_i for each x_i indicating the *independent probability* that x_i is true.

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 - Yeah, it's like TID instances!

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→ Probability 0.1

x_2 Projector in C017 is working

→ Probability 0.2

pc-table semantics example

date	teacher	room	$x_1 : 0.1, x_2 : 0.2$
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 - **Probability** of ν : $(1 - 0.1) \times 0.2 = 0.18$
 - Evaluate the **conditions**
- Probability of possible world: **sum** over the valuations
- Here: **only** this valuation, 0.18

pc-tables capture TID

Give each tuple its **own** variable:

U

date	teacher	room
04	Silviu	Co17
04	Antoine	Co17
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date	teacher	room	
04	Silviu	Co17	x_1
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date	teacher	room	
04	Silviu	CO17	x_1
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→ Give each **variable** the **probability** of the tuple

pc-tables capture mutually exclusive

- Remember **non-Boolean** c-tables

pc-tables capture mutually exclusive

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U

mon	day	teacher	room	
Jan	04	Silviu	CO17	$x = 1$
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- Give a **probability** to each value of x , summing up to 1

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 - **Example**: x has probability:
 - 0.8 to be 1
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 - 0.1 to be 3
- Remember our **rewriting** from non-Boolean to Boolean...

Reminder: rewriting non-Boolean to Boolean

U

mon	day	teacher	room	
Jan	04	Silviu	Co17	$x = 00$
Jan	04	Antoine	Co17	$x = 01$
Jan	04	Fabian	Co17	$x = 10$

Reminder: rewriting non-Boolean to Boolean

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→ How to choose the **probabilities**?

Choosing the probabilities

- We start with the probabilities:
 - $x = 00$ has probability 0.8
 - $x = 01$ has probability 0.1
 - $x = 10$ has probability 0.1
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Choosing the probabilities

- We start with the **probabilities**:
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Converting mutually exclusive to pc-tables

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→ Rewriting:

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→ x_1 has proba $1/9$, x_2 has proba $1/2$, x_2' has proba 0

Capturing BID with pc-tables

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x_1 has probability 0.1

y_1 has probability 0.1

y_2 has probability 1/9

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 - assuming that variables in the input relations are different
 - this preserves probabilities
- pc-tables are a strong representation system

Capturing all probabilistic instances

- Remember:
 - **Support** \mathcal{U} : uncertain relation
 - Here, set of subsets of a **finite** set of tuples
 - **Probability distribution** π on \mathcal{U}

Capturing all probabilistic instances

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→ Can **any** probabilistic instance be **represented** by a pc-table?

Capturing uncertain instances with Boolean c-tables (1)

Remember from last time:

- Number the possible worlds in binary
- For each tuple, write the possible worlds where it appears

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00		01		10		11	
v	w	v	w	v	w	v	w
a	d	a	d	a	d	a	d
b	e	b	e	b	e	b	e
c	f	c	f	c	f	c	f

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v	w
----------	----------

a	d	$x = 00 \vee x = 01 \vee x = 10 \vee x = 11$
---	---	--

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→ We can **also** do this with pc-tables

Capturing uncertain instances with Boolean c-tables (2)

Remember: the **second step** was to **reduce** to binary:

v	w	
a	d	$x = 00 \vee x = 01 \vee x = 10 \vee x = 11$
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v	w	
a	d	$\neg x_1 \wedge \neg x_2 \vee \neg x_1 \wedge x_2 \vee x_1 \wedge \neg x_2 \vee x_1 \wedge x_2$
b	e	$\neg x_1 \wedge x_2$
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- For pc-instances, how to **choose the probabilities**?

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- For pc-instances, how to **choose the probabilities**?
- We have **seen this**: this is encoding a **mutually exclusive choice**_{43/45}

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Conclusion

Summary

We have seen **relational** formalisms for **probabilistic** instances:

- **TID**, a simple model with **independent probabilities** on tuples
- **BID**, adding **blocks** with mutually exclusive choices
- **pc-tables**, i.e., **Boolean c-tables** with probabilities on variables
 - pc-tables can capture **any** probabilistic instance

Summary

We have seen **relational** formalisms for **probabilistic** instances:

- **TID**, a simple model with **independent probabilities** on tuples
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- In the next class: how to evaluate **queries** efficiently

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