### Lecture 6

# Relational

# Algebra

COMP3278A

Introduction to Database Management Systems

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Acknowledgement: Dr Chui Chun Kit

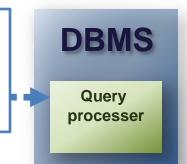
# Motivation

Find the dept. names where employees named Smith work.

**SELECT** D.name

FROM Employees E, Works\_in W, Departments D
WHERE E.name = 'Smith' AND

E.employee\_id = W.employee\_id AND
W.department\_id = D.department\_id;



How does the DBMS execute this SQL query?

**e.g.** Join which two tables first?

Which constraint is applied

first?

### **Employees**

employee_id	name	salary
1	Jones	26000
2	Smith	28000
3	Parker	35000
4	Smith	24000

### Works\_in

_		
employee_id	department_id	since
1	1	2001-1-1
2	1	2002-4-1
2	2	2005-2-2
3	3	2003-1-1
4	3	2005-1-1

### **Departments**

department_id	name	budget
1	Toys	122000
2	Tools	239000
3	Food	100000

# Relational Algebra

- Relational Algebra is similar to normal algebra (as in 2+3\*x-y), except it uses relations (tables) as operand, and a new set of operators.
- The inner, lower-level operations of a relational DBMS are (or are similar to) relational algebra operations.
- We need to know about relational algebra to understand query execution and optimization in a relational DBMS.

# Section 1

# Basic operators

# **Basic operators**

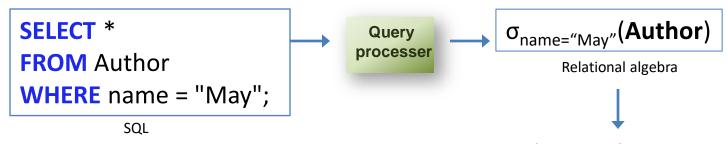
- Select (σ)
- Project (π)
- $\bigcirc$  Union ( $\cup$ )
- Set difference (-)
- Cartesian product (x)
- Rename (ρ)

That is to say, there should be programs (or functions) implemented in the DBMS for each of these relational operators



### Selection

- $\bigcirc$   $\sigma_p(R) = \{ t \mid t \in R \land p(t) \}$
- Example
  - Consider the relation Author (<u>authorID</u>, name, date of birth)
  - Select all authors called "May".



### **Author**

authorID	name	date of birth
101	May	Nov 16
102	Bonnie	Jan 15
103	May	Jul 11
104	Raymond	Apr 30
105	Tiffany	Oct 10

### $\sigma_{\text{name}=\text{"May"}}(\text{Author})$

authorID	name	date of birth
101	May	Nov 16
103	May	Jul 11

# Projection

The result = relation over the k attributes A1, A2, ..., Ak obtained from R by erasing the columns that are not listed and eliminating duplicate rows

117

118

Little girl

Myr

- A copy of R with only listed attributes A1 to Ak.
- Example

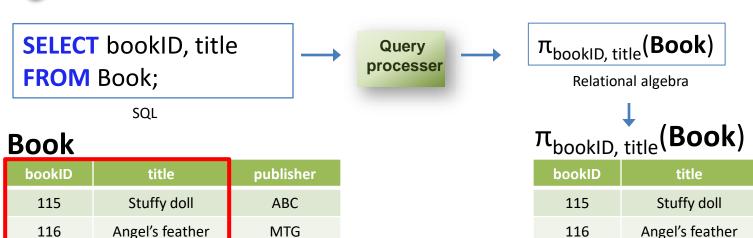
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118

Little girl

Myr

- Consider the relation Book (bookID, title, publisher)
- Report only the bookID and title of all the books.

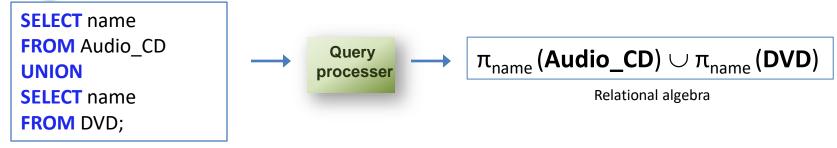


MGH

ABC

### Union

- $\bigcirc$  R $\cup$ S = { t | t  $\in$  R $\vee$ t  $\in$  S }
  - R and S must have the same number of attributes and attribute data types are compatible.
- Example
  - Find the name of all products in Audio\_CD and DVD tables.



### **Audio CD**

name	#tracks
One Heart	14
Miracle	14

SQL

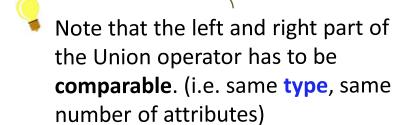
#### **DVD**

name	length	subtitle
Prince of Persia	110	English, Chinese
Villon's Wife	90	Japanese
Legend is born: Ip Man	90	Chinese

# Union



name
One Heart
Miracle
Prince of Persia
Villon's Wife
Legend is born: Ip Man





 $\frac{\pi_{\text{name}}}{\pi_{\text{name}}}$  (Audio\_CD)

name

One Heart

Miracle



### Audio\_CD

name	#tracks
One Heart	14
Miracle	14

### $\frac{\pi_{\text{name}}}{\pi_{\text{name}}}$

#### name

Prince of Persia

Villon's Wife

Legend is born: Ip Man

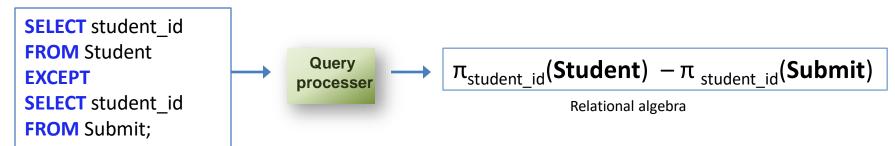


#### **DVD**

name	length	subtitle
Prince of Persia	110	English, Chinese
Villon's Wife	90	Japanese
Legend is born: Ip Man	90	Chinese

### Set difference

- $\bigcirc$  R-S = { t | t  $\in$  R  $\land$  t  $\notin$  S }
  - R and S must have the same number of attributes and attribute data types are compatible.
- Example
  - Find the ID of the students who haven't submitted the assignment.



#### **Student**

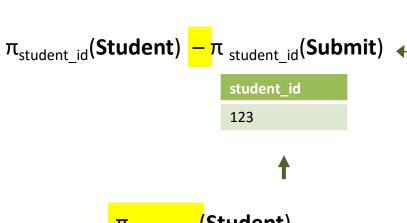
SQL

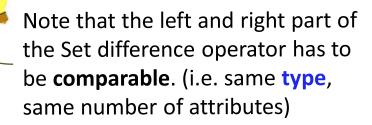
student_id	name	gender	major
123	Kit	М	CS
456	Yvonne	F	CS
789	Paul	М	CS

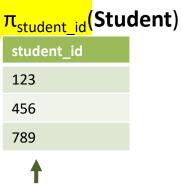
#### Submit

student_id	assignment_id	date
456	1	28/9
789	1	25/9

# Set difference







$\pi_{\text{student\_id}}(\text{Subi}$	mit)
student_id	
456	
789	



### Student

student_id	name	gender	major
123	Kit	М	CS
456	Yvonne	F	CS
789	Paul	М	CS

#### **Submit**

student_id	assignment_id	date
456	1	28/9
789	1	25/9

- $\bigcirc$  R  $\times$  S = { t q | t  $\in$  R  $\land$  q  $\in$  S }
  - No attributes with a common name in R and S.
- Example
  - Display the date of the tutorials of the course "Introduction to Database Management Systems".

```
SELECT Tutorial.date
FROM Course, Tutorial
WHERE
Coruse.name="Introduction to
Database Management Systems" AND
Course.course_id = Tutorial.course_id;
```



```
\pi_{\text{Tutorial.date}} \ ( \sigma_{\text{Course.name="Introduction to Database Management Systems"}} \ ( \sigma_{\text{Course.course\_id=Tutorial.course\_id"}} \ (\text{\textbf{Course}} \times \text{\textbf{Tutorial}}) )
```

SQL

#### Relational algebra

#### Course

course_id	name
c1119	Data Structures and Algorithms
c0278a	Introduction to Database Management Systems

### **Tutorial**

tutorial_id	course_id	date
1	c1119	5/9
1	c0278a	7/9
2	c0278a	15/9

### **Course** × **Tutorial**

Course.course_id	Course.name	Tutorial .tutorial_id	Tutorial .course_id	Tutorial .date
c1119	Data Structures and Algorithms	1	c1119	5/9
c1119	Data Structures and Algorithms	1	c0278a	7/9
c1119	Data Structures and Algorithms	2	c0278a	15/9
c0278a	Introduction to Database Management Systems	1	c1119	5/9
c0278a	Introduction to Database Management Systems	1	c0278a	7/9
c0278a	Introduction to Database Management Systems	2	c0278a	15/9



### **Course**

# course\_id name c1119 Data Structures and Algorithms c0278a Introduction to Database Management Systems

### **Tutorial**

tutorial_id	course_id	date
1	c1119	5/9
1	c0278a	7/9
2	c0278a	15/9

### **Course** × **Tutorial**

Course.course_id	Course.name	Tutorial .tutorial_id	Tutorial .course_id	Tutorial .date
c1119	Data Structures and Algorithms	1	c1119	5/9
c1119	Data Structures and Algorithms	1	c0278a	7/9
c1119	Data Structures and Algorithms	2	c0278a	15/9
c0278a	Introduction to Database Management Systems	1	c1119	5/9
c0278a	Introduction to Database Management Systems	1	c0278a	7/9
c0278a	Introduction to Database Management Systems	2	c0278a	15/9



### $\sigma_{\text{Course.course\_id}=\text{Tutorial.course\_id}}$ (Course $\times$ Tutorial)

Course.course_id	Course.name	Tutorial .tutorial_id	Tutorial .course_id	Tutorial .date
c1119	Data Structures and Algorithms	1	c1119	5/9
c0278a	Introduction to Database Management Systems	1	c0278a	7/9
c0278a	Introduction to Database Management Systems	2	c0278a	15/9

```
π<sub>Tutorial.date</sub> (
                                                                                                             Tutorial.date
          σ<sub>Course.name="Introduction to Database Management Systems"</sub>(
                   \sigma_{Course\_id=Tutorial.course\_id"} \text{ (Course} \times \text{Tutorial)}
                                                                                                             7/9
                                                                                                             15/9
```



 $\sigma_{\text{Course.name}=\text{"Introduction to Database Management Systems"}}(\sigma_{\text{Course.course}\_\text{id}=\text{Tutorial.course}\_\text{id}"}(\text{Course}\times\text{Tutorial}))$ 

Course.course_id	Course.name	Tutorial .tutorial_id	Tutorial .course_id	Tutorial .date
c0278a	Introduction to Database Management Systems	1	c0278a	7/9
c0278a	Introduction to Database Management Systems	2	c0278a	15/9



### $\sigma_{Course\_id=Tutorial.course\_id} \text{ (Course} \times \text{Tutorial)}$

Course.course_id	Course.name	Tutorial .tutorial_id	Tutorial .course_id	Tutorial .date
c1119	Data Structures and Algorithms	1	c1119	5/9
c0278a	Introduction to Database Management Systems	1	c0278a	7/9
c0278a	Introduction to Database Management Systems	2	c0278a	15/9

### Rename

- Notation:  $\rho_x(E)$
- Rename operator allows us to name and refer to the results of relational-algebra expressions
- $\bigcirc$   $\rho_X$  (E) returns the expression E under the name X

```
SELECT Tutorial.date
FROM Course, Tutorial
WHERE
Coruse.name="Introduction to Database
Management Systems" AND
Course.course_id = Tutorial.course_id;
```

SQL

```
\pi_{Tutorial.date} \ ( \sigma_{Course.name="Introduction to Database Management Systems"} \ ( \sigma_{Course.course\_id=Tutorial.course\_id"} \ (\textbf{Course} \times \textbf{Tutorial}) )
```

Relational algebra (Without rename)

```
SELECT T.date
FROM Course C, Tutorial T
WHERE
C.name="Introduction to Database
Management Systems" AND
C.course_id = T.course_id;
```

```
\pi_{\text{T.date}} (
\sigma_{\text{C.name="Introduction to Database Management Systems"}} (
\sigma_{\text{C.course\_id=T.course\_id"}} (\rho_{\text{C}} (\text{Course}) \times \rho_{\text{T}} (\text{Tutorial}))
)
```

# Section 2

# Exercises

- Given the following relational schema:
  - Student (UID, name, age).
  - Course (CID, title).
  - Enroll (<u>UID</u>, <u>CID</u>) with <u>UID</u> referencing **Student** and <u>CID</u> referencing **Course**.
  - \*UID and CID, age are interger; name and title are varchar.
- Which of the following is (are) valid Relational Algebra expression(s)?

  - **Ourse**- $\pi_{UID}$ (Enroll)



The left and right parts of Set difference have to be **comparable** (same number of attributes).

- Given the following relational schema:
  - Student (UID, name, age).
  - Course (CID, title).
  - Enroll (UID, CID) with UID referencing Student and **CID** referencing **Course**.

- Which of the following is (are) valid Relational Algebra expression(s)?
  - $\odot$   $\sigma_{age<18}$ (Student  $\cup$  Course)
  - $\sigma_{\text{age}<18}(\pi_{\text{UID, name}}(\text{Student}))$



Student and Course have different

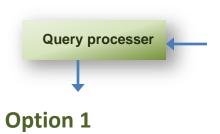


No attribute "age" for selection.

<sup>\*</sup>UID and CID, age are interger; name and title are varchar.

**SELECT** employee id,name **FROM** Employee WHERE employee id < 100; SQL

Find the employeeIDs and names of employees whose employeeID < 100.



### σ<sub>employeeID<100</sub>(Employee)

employeeID	name	start_date
97	May	Nov 16, 1997
98	Felix	Jun 30, 2003
99	May	Sep 18, 2007

### **Employee**

employeeID	name	start_date
97	May	Nov 16, 1997
98	Felix	Jun 30, 2003
99	May	Sep 18, 2007
100	George	Jan 1, 2008

### **Option 2**

Query processer

 $\pi_{\text{name, employeeID}}(\text{Employee})$ 

employeeID	name
97	May
98	Felix
99	May
100	George

99

 $\frac{\pi_{\text{name, employeeID}}}{\sigma_{\text{employeeID}}} (\sigma_{\text{employeeID} < 100} (\text{Employee}))$ 

employeeID	name
97	May
98	Felix
99	May

Which one is better?

σ <sub>employeeID</sub>	<mark>0&lt;100</mark> (π <sub>name, 0</sub>	<sub>employeeID</sub> (Er	nployee))
	employeeID	name	
tter?	97	May	
itei:	98	Felix	

May

Find the dept. id(s) where employees named Smith work.

### **Option 1**

 $\pi_{\text{W.department\_id}}(\sigma_{\text{E.name="Smith"}}(\sigma_{\text{E.employee\_id=W.employee\_id}}(\rho_{\text{E}}(\text{Employees}) \times \rho_{\text{W}}(\text{Works\_in}))))$ 



Now we compute the Cartesian product between **Employees** and **Works\_in** first, which creates an **intermediate relation** with 10,000\*5 = 50,000 tuples! Can we reduce the size of intermediate relation?

### **Employees**

### Works\_in

employee_id	name	salary	employee_id	department_id	since
1	Jones	26000	1	1	2001-1-1
2	Smith	28000	2	1	2002-4-1
		•••	2	2	2005-2-2
10000		•••	3	3	2003-1-1
Employees	(10000 tu	ples)	4	3	2005-1-1

Find the dept. id(s) where employees named Smith work.

### **Option 1**

 $\pi_{\text{W.department\_id}}(\sigma_{\text{E.name="Smith"}}(\sigma_{\text{E.employee\_id=W.employee\_id}}(\rho_{\text{E}}(\text{Employees}) \times \rho_{\text{W}}(\text{Works\_in}))))$ 

### Option 2

$$\pi_{\text{W.department\_id}}(\sigma_{\text{E.employee\_id=W.employee\_id}}(\sigma_{\text{E.name="Smith"}}(\rho_{\text{E}}(\text{Employees})) \times \rho_{\text{W}}(\text{Works\_in})))$$



We apply selection on Employees before the Cartesian product. If there is only two employee named Smith, the **intermediate relation** can be reduced to 2\*5 = 10 tuples!

### **Employees**

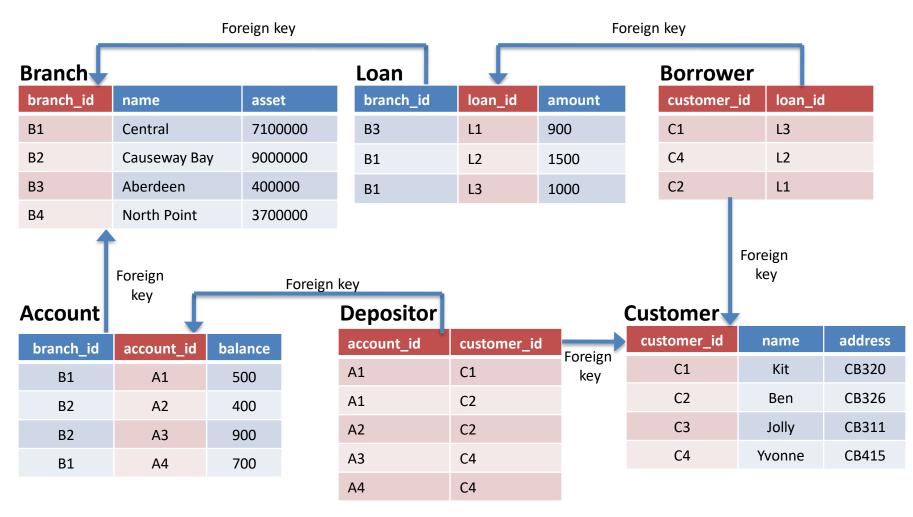
### Works\_in

employee_id	name	salary		
1	Jones	26000		
2	Smith	28000		
10000				
Employees (10000 tuples)				

employee_id	department_id	since
1	1	2001-1-1
2	1	2002-4-1
2	2	2005-2-2
3	3	2003-1-1
4	3	2005-1-1

SELECT W.department\_id
FROM Employees E, Works\_in W
WHERE E.name = 'Smith' AND
E.employee\_id = W.employee\_id;

# Banking example



Find the names of all customers who have a loan, an account, or both in a bank.

FROM Borrower
UNION
SELECT customer\_id
FROM Depositor

 $\pi_{ ext{customer_id}}$  (Borrower)  $\cup \pi_{ ext{customer_id}}$  (Depositor)

Find the names of all customers who have both a loan and an account in a bank.

SELECT customer\_id FROM Borrower INTERSECT SELECT customer\_id FROM Depositor

```
\pi_{\text{customer id}} (Borrower) \cap \pi_{\text{customer id}} (Depositor)
```

Wait! Do we have set intersection in relational algebra ©?

 $\pi_{\text{customer_id}}$  (Borrower) – ( $\pi_{\text{customer_id}}$  (Borrower) –  $\pi_{\text{customer_id}}$  (Depositor))

# Additional operators

- These operations do not add any power to relational algebra, but simplify common queries.
  - Set intersection ( ∩ )
  - Natural join (⋈)
  - Assignment (←)
  - Left outer join (⋈), Right outer join (⋈)
  - Division (÷)

# Section 3

# Additional operators

# Motivation

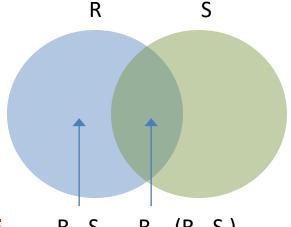
- The fundamental operators of the relational algebra introduced are sufficient to express any relationalalgebra query.
- However, if we restrict ourselves to just the fundamental operators, certain common queries are tedious to express.
  - Therefore, we define additional operators that do not add any power to the algebra, but simplify common queries

That is to say, for each additional operator, we can give an equivalent expression that use only the fundamental operators.

# Additional operators

- $\bigcirc$  Set intersection ( $\cap$ )
- Natural join (⋈)
- Assignment (←)
- Left outer join (⋈), Right outer join (⋈)
- Division (÷)

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- $\bigcirc R \cap S = R (R S)$ 
  - R and S must have the same number of R-S R-(R-S) attributes and attribute data types are compatible.
- Example
  - Query: Find the employee\_id of employees who work in department 1 and department 3.

```
SELECT employee_id
FROM Works_in
WHERE department_id=1

INTERSECT

SELECT employee_id
FROM Works_in
WHERE department_id=3
```

For your reference, there is another way to answer the same query by joining the **Works\_in** table with itself.



FROM Works\_in W1, Works\_in W2
WHERE

W1.employee\_id = W2.employee\_id AND

W1.department\_id=1 AND

W2. department\_id = 3

Query processer

W	or	·ks	i	n
			_	

employee_id	department_id
1	1
2	1
1	3

SQL

For your reference, there is another way to answer the same query by joining the **Works\_in** table with itself.



SELECT DISTINCT employee\_id
FROM Works\_in W1, Works\_in W2
WHERE

W1.employee\_id = W2.employee\_id AND

W1.department\_id=1 AND

W2. department id = 3

 $\rho_{W1}$  (Works\_in)  $\times \rho_{W2}$  (Works\_in)

W1.employee_id	W1.department_id	W2.employee_id	W2.department_id
1	1	1	1
1	1	2	1
1	1	1	3
2	1	1	1
2	1	2	1
2	1	1	3
1	3	1	1
1	3	2	1
1	3	1	3

Query processer Works\_in

employee_id	department_id
1	1
2	1
1	3

SQL

For your reference, there is another way to answer the same query by joining the **Works\_in** table with itself.



**SELECT DISTINCT** employee\_id **FROM** Works\_in **W1**, Works\_in **W2 WHERE** 

W1.employee\_id = W2.employee\_id AND

W1.department\_id=1 AND

W2. department\_id = 3

 $\sigma_{\text{W1.employee\_id=W2.employee\_id}}(\rho_{\text{W1}}(\text{Works\_in}) \times \rho_{\text{W2}}(\text{Works\_in}))$ 

W1.employee_id	W1.department_id	W2.employee_id	W2.department_id
1	1	1	1
1	1	1	3
2	1	2	1
1	3	1	1
1	3	1	3

 $\rho_{\text{W1}}\left(\text{Works\_in}\right) \times \rho_{\text{W2}}\left(\text{Works\_in}\right)$ 



W1.employee_id	W1.department_id	W2.employee_id	W2.department_id
1	1	1	1
1	1	2	1
1	1	1	3
2	1	1	1
2	1	2	1
2	1	1	3
1	3	1	1
1	3	2	1
1	3	1	3
1	3	1	3

Query processer

Works\_in

employee_id	department_id
1	1
2	1
1	3

SQL

 $\sigma_{\text{W1.department\_id=1}} \wedge \sigma_{\text{W2.department\_id=3}} (\sigma_{\text{W1.employee\_id=W2.employee\_id}} (\rho_{\text{W1}} (\text{Works\_in}) \times \rho_{\text{W2}} (\text{Works\_in})))$ 

For your reference, there is another way to answer the same query by joining the **Works\_in** table with itself.



**SELECT DISTINCT** employee\_id **FROM** Works\_in **W1**, Works\_in **W2 WHERE** 

W1.employee\_id = W2.employee\_id AND W1.department id=1 AND

W2. department id = 3

SQL

W1.employee_id	W1.department_id	W2.employee_id	W2.department_id
1	1	1	3

 $\sigma_{\text{W1.employee\_id=W2.employee\_id}}\left(\rho_{\textbf{W1}}\left(\textbf{Works\_in}\right) \times \rho_{\textbf{W2}}\left(\textbf{Works\_in}\right)\right)$ 

W1.employee_id	W1.department_id	W2.employee_id	W2.department_id
1	1	1	1
1	1	1	3
2	1	2	1
1	3	1	1
1	3	1	3

 $\rho_{\text{W1}}\left(\text{Works\_in}\right) \times \rho_{\text{W2}}\left(\text{Works\_in}\right)$ 

. ***	VV 2 · /		
W1.employee_id	W1.department_id	W2.employee_id	W2.department_id
1	1	1	1
1	1	2	1
1	1	1	3
2	1	1	1
2	1	2	1
2	1	1	3
1	3	1	1
1	3	2	1
1	3	1	3

Query processer

Works_in	4
Works_in	

employee_id	department_id
1	1
2	1
1	3

# Natural join

- Usually, a query that involves a Cartesian product includes a selection operation on the result of the Cartesian product.
  - The selection operation most often requires that all attributes that are common to the relations that are involved in the Cartesian product be equated.

Requires the common attributes to be equated

where  $R \cap S = \{A1, A2, ..., An\}$ Commutative:  $r \bowtie s = s \bowtie r$ 

# Natural join

- The schema of R ⋈ S is R-schema ∪ S-schema (repeated attributes are removed)
  - $\bigcirc$  For each pair of tuples  $t_r$  from R and  $t_s$  from S,
  - If  $t_r$  and  $t_s$  share the same value over each of the common attributes in R and S,
  - $\bigcirc$  Tuple t will be added to the result of R  $\bowtie$  S.

# Natural join

R

· \		
Α	В	
1	1	
1	2	
2	3	

S

Α	С
1	2
2	1

- Ommon attributes:  $R \cap S = \{A\}$
- $\bigcirc$  Attributes of the resulting relation: R  $\cup$  S = {A,B,C}

 $R \times S$ 

R.A	R.B	S.A	s.c	
1	1	1	2	
1	1	2	1	<b>→</b>
1	2	1	2	
1	2	2	1	
2	3	1	2	_
2	3	2	1	

 $\sigma_{R.A=S.A}(R\times S)$ 

R.A	R.B	S.A	S.C
1	1	1	2
1	2	1	2
2	3	2	1

 $\pi_{A,B,C}(\sigma_{R.A=S.A}(\mathbf{R}\times\mathbf{S}))$ 

Α	В	С
1	1	2
1	2	2
2	3	1

equivalent to:



Α	В	С
1	1	2
1	2	2
2	3	1

## Assignment

- It is convenient to write a relational-algebra expression by assigning parts of it to temporary relation variables.
- The assignment operator, denoted by " $\leftarrow$ ", works like assignment operator "=" in programming language.
  - $\bullet$  temp1  $\leftarrow \pi_a(R)$
  - $\bigcirc$  temp2  $\leftarrow \pi_a(S)$
  - result ← temp1 temp2

With the assignment operator, a query can be written as a sequential program.

temp1, temp2, result are called "relation variable".

### Outer join

The outer-join operator is an extension of the join operation to deal with missing information.

#### Customer

customer_id	name	address
C1	Kit	CB320
C2	Ben	CB326
C3	Jolly	CB311
C4	Yvonne	CB415

### **Depositor**

account_id	customer_id
A1	C1
A1	C2
A2	C2
A3	C4
A4	C4

### **Customer** ⋈ **Depositor**

customer_id	name	address	account_id
C1	Kit	CB320	A1
C2	Ben	CB326	A1
C2	Ben	CB326	A2
C4	Yvonne	CB415	A3
C4	Yvonne	CB415	A4



Natural join (e.g., joining Customer and Depositor) result in a table without the information of customers (e.g., C3 in this case) who has no account.

- Natural join result, plus
- The tuples that do not match any tuples from the other side.

$$R \supset S = (R \bowtie S) \cup (R - \pi_R(R \bowtie S)) \times \{(null, ..., null)\}$$

#### Customer

customer_id	name	address
C1	Kit	CB320
C2	Ben	CB326
C3	Jolly	CB311
C4	Yvonne	CB415

### **Depositor**

customer_id
C1
C2
C2
C4
C4

### **Customer** ⋈ **Depositor**

customer_id	name	address	account_id
C1	Kit	CB320	A1
C2	Ben	CB326	A1
C2	Ben	CB326	A2
C4	Yvonne	CB415	A3
C4	Yvonne	CB415	A4

Let's illustrate why the left outer join is defined as  $(R \bowtie S) \cup (R - \pi_R(R \bowtie S)) \times \{(null, ..., null)\}$  through a step-by-step illustration.

$$R \supset X S = (R \bowtie S) \cup (R - \pi_R(R \bowtie S)) \times \{(null, ..., null)\}$$

#### Customer

customer_id	name	address
C1	Kit	CB320
C2	Ben	CB326
C3	Jolly	CB311
C4	Yvonne	CB415

### **Depositor**

account_id	customer_id
A1	C1
A1	C2
A2	C2
A3	C4
A4	C4

### **Customer** ⋈ **Depositor**

customer_id	name	address	account_id
C1	Kit	CB320	A1
C2	Ben	CB326	A1
C2	Ben	CB326	A2
C4	Yvonne	CB415	A3
C4	Yvonne	CB415	A4

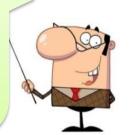
**Customer** -  $\pi_{\text{Customer's attributes}}$  (Customer  $\bowtie$  Depositor)

customer_id	name	address
C3	Jolly	CB311

### Finding missed tuples in the natural join

R -  $\pi_R$ ( R  $\bowtie$  S ) is to generate the tuples in R (i.e., Customer) that are missed in the natural join.

i.e., C3 Jolly in Customer doesn't have any matched records in Depositor, we use this part to recover Jolly's record.



$$R \supset S = (R \bowtie S) \cup (R - \pi_R(R \bowtie S)) \times \{(null, ..., null)\}$$

#### Customer

customer_id	name	address
C1	Kit	CB320
C2	Ben	CB326
C3	Jolly	CB311
C4	Yvonne	CB415

### **Depositor**

account_id	customer_id
A1	C1
A1	C2
A2	C2
A3	C4
A4	C4

#### **Customer** ⋈ **Depositor**

customer_id	name	address	account_id
C1	Kit	CB320	A1
C2	Ben	CB326	A1
C2	Ben	CB326	A2
C4	Yvonne	CB415	A3
C4	Yvonne	CB415	A4

Customer -  $\pi_{\text{Customer's attributes}}$  (Customer  $\bowtie$  Depositor )

customer_id	name	address
C3	Jolly	CB311

Customer -  $\pi_{\text{Customer's attributes}}$  (Customer  $\bowtie$  Depositor)  $\times \{ (null, ..., null) \}$ 

customer_id	name	address	account_id
C3	Jolly	CB311	null

Constructing the missed tuple by adding null value to extra attributes

The " $\times$  { (null, ..., null) }" part simply add back the remaining column values as null because there is no matched records in S (i.e., Depositor).



$$R \supset X S = (R \bowtie S) \cup (R - \pi_R(R \bowtie S)) \times \{(null, ..., null)\}$$

#### Customer

customer_id	name	address
C1	Kit	CB320
C2	Ben	CB326
C3	Jolly	CB311
C4	Yvonne	CB415

### **Depositor**

account_id	customer_id
A1	C1
A1	C2
A2	C2
A3	C4
A4	C4

#### **Customer** ⋈ **Depositor**

customer_id	name	address	account_id
C1	Kit	CB320	A1
C2	Ben	CB326	A1
C2	Ben	CB326	A2
C4	Yvonne	CB415	A3
C4	Yvonne	CB415	A4

Customer -  $\pi_{\text{Customer's attributes}}$  (Customer  $\bowtie$  Depositor )

customer_id	name	address
C3	Jolly	CB311

Customer -  $\pi_{\text{Customer's attributes}}$  (Customer  $\bowtie$  Depositor )  $\times$  { (null, ..., null) }

customer_id	name	address	account_id
C3	Jolly	CB311	null

(Customer  $\bowtie$  Depositor)  $\begin{cases} \begin{cases} \be$ 

customer_id	name	address	account_id
C1	Kit	CB320	A1
C2	Ben	CB326	A1
C2	Ben	CB326	A2
C3	Jolly	CB311	null
C4	Yvonne	CB415	A3
C4	Yvonne	CB415	A4

equivalent to: **Customer**  $\supset$  **Depositor** 

$$R \bowtie S = (R \bowtie S) \cup (S - \pi_S(R \bowtie S)) \times \{(null, ..., null)\}$$

#### Customer

customer_id	name	address
C1	Kit	CB320
C2	Ben	CB326
C3	Jolly	CB311
C4	Yvonne	CB415

### **Depositor**

customer_id
C1
C2
C2
C4
C5

### **Customer** ⋈ **Depositor**

customer_id	name	address	account_id
C1	Kit	CB320	A1
C2	Ben	CB326	A1
C2	Ben	CB326	A2
C4	Yvonne	CB415	A3

Let's illustrate why the right outer join is defined as

(R  $\bowtie$  S)  $\cup$  (S -  $\pi_S$ (R  $\bowtie$  S))  $\times$  { (null, ..., null) } through a step-by-step illustration.



$$R \bowtie S = (R \bowtie S) \cup (S - \pi_S(R \bowtie S)) \times \{(null, ..., null)\}$$

#### Customer

customer_id	name	address
C1	Kit	CB320
C2	Ben	CB326
C3	Jolly	CB311
C4	Yvonne	CB415

### **Depositor**

customer_id
C1
C2
C2
C4
C5

#### **Customer** ⋈ **Depositor**

customer_id	name	address	account_id
C1	Kit	CB320	A1
C2	Ben	CB326	A1
C2	Ben	CB326	A2
C4	Yvonne	CB415	A3

**Depositor** -  $\pi_{\text{Depositor's attributes}}$  (Customer  $\bowtie$  Depositor)

customer_id	account_id
C5	A4

### Finding missed tuples in the natural join

 $S - \pi_S(R \bowtie S)$  is to generate the tuples in S (i.e., Depositor) that are missed in the natural join. i.e., C5 in Depositor doesn't have any matched records in Customer, we use this part to recover C5's record.



$$R \bowtie S = (R \bowtie S) \cup (S - \pi_S(R \bowtie S)) \times \{(null, ..., null)\}$$

#### **Customer**

customer_id	name	address
C1	Kit	CB320
C2	Ben	CB326
C3	Jolly	CB311
C4	Yvonne	CB415

### **Depositor**

customer_id
C1
C2
C2
C4
C5

#### **Customer** ⋈ **Depositor**

customer_id	name	address	account_id
C1	Kit	CB320	A1
C2	Ben	CB326	A1
C2	Ben	CB326	A2
C4	Yvonne	CB415	A3

**Depositor** -  $\pi_{\text{Depositor's attributes}}$  (Customer  $\bowtie$  Depositor )

customer_id	account_id	
C5	A4	

Depositor -  $\pi_{\text{Depositor's attributes}}$  (Customer  $\bowtie$  Depositor)  $\times \{ (null, ..., null) \}$ 

customer_id	name	address	account_id
C5	null	null	A4

Constructing the missed tuple by adding *null* value to extra attributes

The " $\times$  { (null, ..., null) }" part simply add back the remaining column values as null because there is no matched records in R (i.e., Customer).



$$R \bowtie S = (R \bowtie S) \cup (S - \pi_S(R \bowtie S)) \times \{(null, ..., null)\}$$

#### **Customer**

customer_id	name	address
C1	Kit	CB320
C2	Ben	CB326
C3	Jolly	CB311
C4	Yvonne	CB415

### **Depositor**

account_id	customer_id
A1	C1
A1	C2
A2	C2
A3	C4
A4	C5

#### **Customer** ⋈ **Depositor**

customer_id	name	address	account_id
C1	Kit	CB320	A1
C2	Ben	CB326	A1
C2	Ben	CB326	A2
C4	Yvonne	CB415	А3

**Depositor** -  $\pi_{Depositor's \ attributes}$  (Customer  $\bowtie$  Depositor )

customer_id	account_id
C5	A4

**Depositor** -  $\pi_{\mathsf{Depositor's}}$  attributes (Customer  $\bowtie$  Depositor )  $\times$  { (null, ..., null) }

customer_id	name	address	account_id
C5	null	null	A4

(Customer  $\bowtie$  Depositor)  $\smile$  (Depositor -  $\pi_{\text{Depositor's attributes}}$  (Customer  $\bowtie$  Depositor))  $\times$  { (null, ..., null) }

customer_id	name	address	account_id
C1	Kit	CB320	A1
C2	Ben	CB326	A1
C2	Ben	CB326	A2
C4	Yvonne	CB415	A3
C5	null	null	A4

equivalent to: **Customer** Depositor

R	
Α	В
1	1
2	1
2	2
3	3
4	1
4	2
4	3
1 2 2 3 4 4 4	1 2 3 1

•	_			
		В		
		1		
		2		
	R	÷	ς	

C

R	÷	S	
	A		
	2		
	4		

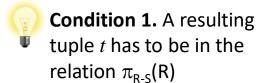
Notation:	R÷	S

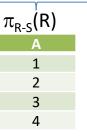






The attributes in relation S is a subset of the attributes in relation R.







**Condition 2.** And if we combine t with each tuple  $s \in S$ , all the combined tuples have to be included in R.

For each tuple $t \in \pi_{R-S}(R)$	<b>2</b> " $\forall$ s $\in$ S, t $\cup$ s" p Generate tuples I all tuples s $\in$ S		"∀ s ∈ S, ( (t ∪ s) ∈ R)" part: Then we check if both tuples generated are in R.
$t_1 \in \pi_{R-S}(R)$ A 1	$\forall s \in S, (t_1 \cup s)$	A B 1 1 1 1 2	In this case, not both tuples are in R, so $t_1$ is <b>NOT</b> in result of R $\div$ S
$t_2 \in \pi_{R-S}(R)$	$\forall s \in S, (t_2 \cup s)$	A B 2 1 2 2	In this case, both tuples are in R, so $t_2$ is in result of R $\div$ S

R	_	(	S
А	В		В
1	1		1
2	1		2
2	2	_ '	
2 3	2	,	R÷S
			R÷S A
3			

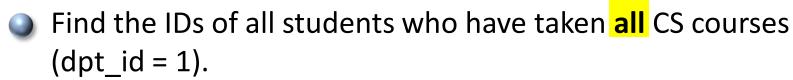
### **Observation**

Let's focus on the result of  $\mathbf{R} \div \mathbf{S}$  (say, the tuple A=2), it means that

- For the tuples with values in attribute A equals to 2 in relation R,
- Those tuples's values in attribute B covers ALL values in attribute B of S.



### Division is used to express queries with "all"



#### Student

student_id	name	dpt_id
1	Peter	1
2	Sharon	1
3	David	2
4	Joe	3

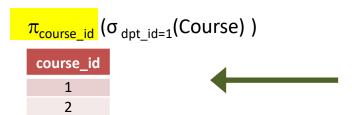
### **Takes**

student_id	course_id	Grade
1	1	Α
1	2	В
1	3	A+
2	3	B-
3	3	В
4	1	С
4	2	A-

#### **Course**

course_id	title	dpt_id	credit
1	Intro to DB	1	6
2	Programming I	1	6
3	Accounting	2	6

Step 1. All part (the relation S): What is the course ID of all CS courses (dpt\_id = 1)?



o <sub>dpt id=1</sub> (Course)			
course_id	title	dpt_id	credit
1	Intro to DB	1	6
2	Programming I	1	6



#### Student

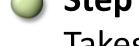
student_id	name	dpt_id
1	Peter	1
2	Sharon	1
3	David	2
4	Joe	3

### **Takes**

student_id	course_id	Grade
1	1	Α
1	2	В
1	3	A+
2	3	B-
3	3	В
4	1	С
4	2	A-

#### **Course**

course_id	title	dpt_id	credit
1	Intro to DB	1	6
2	Programming I	1	6
3	Accounting	2	6



Step 2. Dividend part (the relation R): What is the list of Takes tuples (i.e., all < student id, course id > pairs)?

$\pi_{\text{student}}$	id course	id	(Ta	kes)
<b>"</b> student	id,course	ıa	( . ∽	,

$\pi_{ ext{course id}}$	$(\sigma_{dpt id=1})$	(Course) )
-------------------------	-----------------------	------------

student_id	course_id
1	1
1	2
1	3
2	3
3	3
4	1
4	2

#### Student

student_id	name	dpt_id
1	Peter	1
2	Sharon	1
3	David	2
4	Joe	3

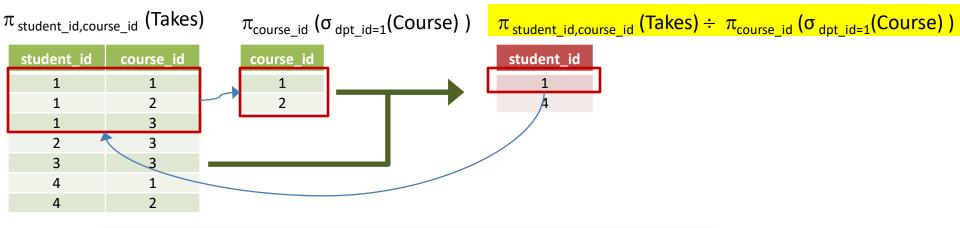
### **Takes**

student_id	course_id	Grade
1	1	Α
1	2	В
1	3	A+
2	3	B-
3	3	В
4	1	С
4	2	A-

### **Course**

course_id	title	dpt_id	credit
1	Intro to DB	1	6
2	Programming I	1	6
3	Accounting	2	6

Step 3. Division: Which student in  $\pi_{\text{student\_id,course\_id}}$  (Takes) takes all CS courses  $\pi_{\text{course\_id}}$  ( $\sigma_{\text{dpt\_id=1}}$ (course))?.



### Explanation: Let's focus on student\_id = 1 in the result

- It means that, for the tuples in Takes with student\_id=1,
- Those tuple's value in course\_id attribute covers **all** 1 and 2 (i.e., the CS courses).
- That is to say, student with student\_id 1 takes **ALL** CS courses.
- The same argument apples to student\_id 4.





### Division has a property: $R \times S \div S = R$

R	
Α	В
1	1
1	2
2	3

 $D \vee C$ 

С	D
1	2
2	2

В	С	D
1	1	2
1	2	2
2	1	2
2	2	2
3	1	2
3	2	2
	1 1 2 2 2 3	1 1 1 1 2 2 2 1 2 3 1

(11 / 5)		
Α	В	
1	1	= R
1	2	- <b>n</b>
2	3	

 $(R \times S) \div S$ 

An intuitive property of the division operator of the relational algebra is simply that it is the inverse of the cartesian product. For example, if you have two relations R and S, then, if U is a relation defined as the cartesian product of them:  $U = R \times S$ , the division is the operator such that  $U \div R = S$  and  $U \div S = R$ 

### Section 4

### Extended

# operators

- Aggregation function takes a collection of values and returns a single-valued result.
  - e.g., avg, min, max, sum, count, count-distinct
- Aggregate operation in relational algebra:
  - Grouping Divides the tuples into groups.
  - Aggregation Computes an aggregation function in each group to create a result tuple.

$$_{G_{1}, G_{2}, ..., G_{n}}$$
  $g_{F_{1}(A_{1}), F_{2}(A_{2}), ..., F_{n}(A_{n})}$  (E)

- E a relation (can be a result of relational algebra expression).
- G<sub>1</sub>, ..., G<sub>n</sub> attributes used to form groups.
  - Tuples with the same values in  $G_1$  to  $G_n$  are put into the same group.
  - G can be empty, which means that the whole relation is one group.
- $\bigcirc$   $\mathbf{F_i(A_i)}$  an aggregate function applied on an attribute.

#### **Account**

branch_id	account_id	balance
B1	A1	500
B2	A2	400
B2	A3	900
B1	A4	700

```
\rho_{Result(branch\_id,\ sum\_of\_balance)} ( _{branch\_id}\ g_{\ sum(balance)}\ (Account) )
```



branch_id	sum_of_balance
B1	1200
B2	1300

- Step 1. Let's group the tuples in Account according to their branch\_id.
- Step 2. Then aggregate the tuples in each group by summing their values in the balance attribute.
- Step 3. Since the resulting relation has no name after aggregation, we use renaming operator to give name to the relation and attributes.

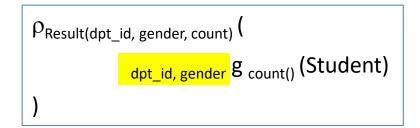


- Note that grouping can be done on multiple attributes.
- E.g., in this case, we group tuples in Student with the same values in both dpt\_id and gender attributes.
- i.e., We are finding the number of male / female students in each department.



#### **Student**

student_id	name	dpt_id	gender
1	Peter	1	M
2	Sharon	1	F
3	David	2	М
4	Joe	2	M
5	Betty	1	F





#### Result

dpt_id	gender	count
1	М	1
1	F	2
2	M	2

### Section 5

# Algebraic properties

### Transformation of expression

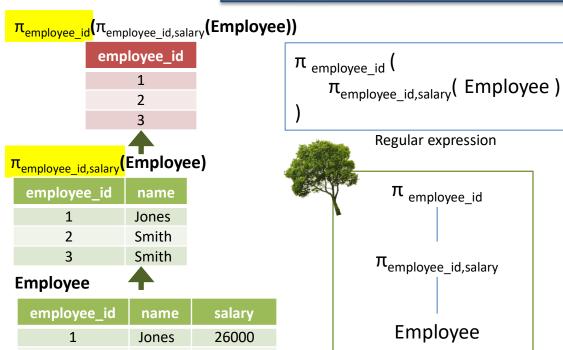
- A query can be expressed in several different ways, with different cost of evaluation.
- Two relational-algebra expressions are said to be equivalent if, on every legal database instance, the two expressions generate the same set of tuples.

In the following discussions, we treat a relation as a set of tuples, the order of the tuples is irrelevant.

Rule 1. Only the final operations in a sequence of projection operations are needed.

$$\pi_{L1} (\pi_{L2} (...(\pi_{Ln} (E))...)) = \pi_{L1} (E)$$

**Expression tree** 



Smith

Smith

28000 24000

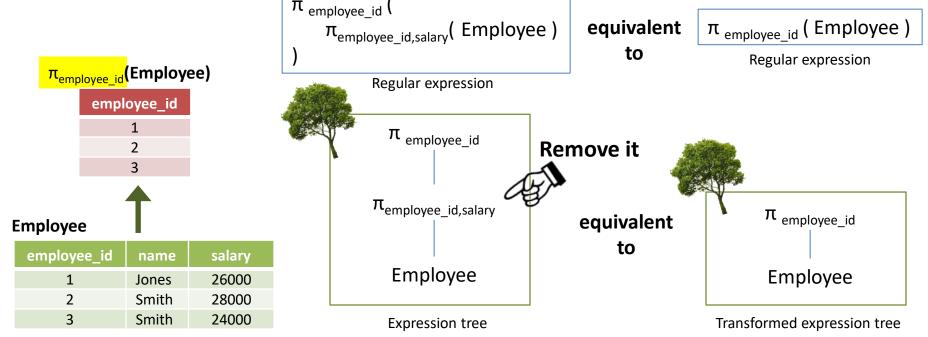
### **Expression tree:**

- Tells which operator is executed ahead of another.
- It allow transformation of the execution order by applying the equivalence rules. (Alter the tree)



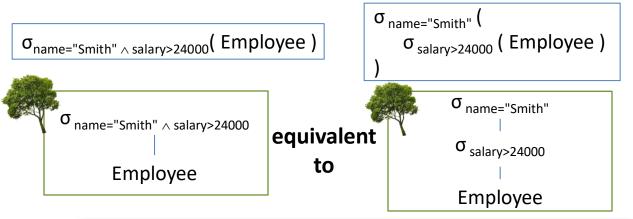
Rule 1. Only the final operations in a sequence of projection operations are needed.

$$\pi_{L1} (\pi_{L2} (...(\pi_{Ln} (E))...)) = \pi_{L1} (E)$$



Rule 2. Conjunctive selection operations can be deconstructed into a sequence of individual selections.

$$\sigma_{p1 \wedge p2}(E) = \sigma_{p1}(\sigma_{p2}(E))$$



employee id

You may wonder why breaking down the conjunctive selections is useful.

We will show that it is useful to reduce temporary result.

We can try to push each selection predicates down the tree (to perform selection as early as possible).



Z	Smith	28000
	1	
σ <sub>salary&gt;24000</sub> (Em	ployee)	
employee_id	name	salary
1	Jones	26000
2	Smith	28000
Employee	<b>↑</b>	
employee_id	name	salary
1	Jones	26000
2	Smith	28000
2	Coo.:+la	24000

 $\sigma_{\text{name}="\text{Smith}"} (\sigma_{\text{salarv}>24000} (\text{ Employee}))$ 

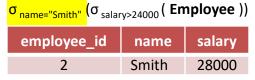
name

28000



Rule 3. Selection operations are commutative.

### $\sigma_{p1}(\sigma_{p2}(E)) = \sigma_{p2}(\sigma_{p1}(E))$



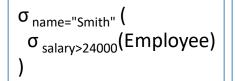


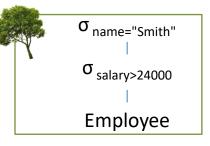
σ <sub>salary&gt;24000</sub>	(Employee)
Salai V/Z4UUL	, , ,

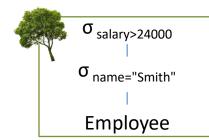
employee_id	name	salary
1	Jones	26000
2	Smith	28000
4	David	25000



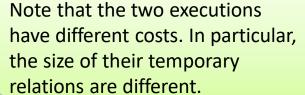
employee_id	name	salary
1	Jones	26000
2	Smith	28000
3	Smith	24000
4	David	25000
4		

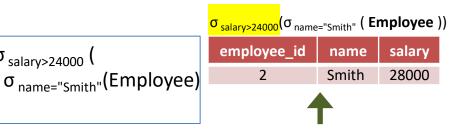


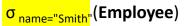




 $\sigma_{\text{salary}>24000}$  (







employee_id	name	salary
2	Smith	28000
3	Smith	24000



#### **Employes**

employee_id	name	salary
1	Jones	26000
2	Smith	28000
3	Smith	24000
Л	David	25000

$$r \bowtie s = \pi_{R \cup S}(\sigma_{r,A1 = s,A1 \land r,A2 = s,A2 \land ... \land r,An = s,An}(r \times s))$$

Rule 4. Natural join operations are associative.

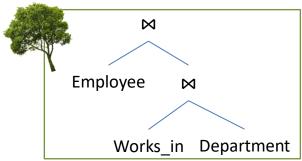
$$(E_1 \bowtie E_2) \bowtie E_3 = E_1 \bowtie (E_2 \bowtie E_3)$$

(Employee ⋈ Works\_in) ⋈ Department

Department

Employee Works\_in

equivalent to Employee ⋈ ( Works\_in ⋈ Department )

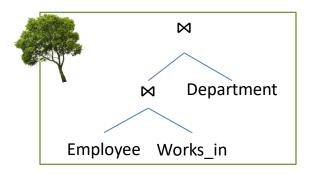


Although both expression trees return the **same** resulting relation, these two expression trees have different costs because the size of their temporary relations are different



### **Expression tree A.**

(Employee ⋈ Works in) ⋈ Department



### (Employee ⋈ Works\_in) ⋈ Department

employee_id	Employee. name	salary	department_id	Department. name
1	Jones	26000	1	Toys
2	Smith	28000	1	Toys
2	Smith	28000	2	Tools
3	Parker	35000	3	Food
4	Smith	24000	3	Food

### **Employee** ⋈ Works\_in

employee_id	name	salary	department_id
1	Jones	26000	1
2	Smith	28000	1
2	Smith	28000	2
3	Parker	35000	3
4	Smith	24000	3



Natural join evaluates 4\*5 = 20 combinations result temporary relation consists of 5 tuples and 4 columns.

#### **Employee**

employee_id	name	salary
1	Jones	26000
2	Smith	28000
3	Parker	35000
4	Smith	24000

### Works in

employee_id	department_id
1	1
2	1
2	2
3	3
4	3

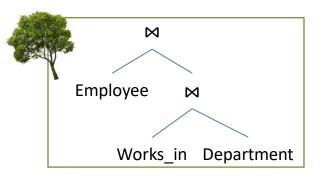
#### **Department**

department_id	name
1	Toys
2	Tools
3	Food



### **Expression tree B.**

Employee ⋈ ( Works\_in ⋈ Department )



### **Employee** ⋈ ( Works\_in ⋈ Department )

employee_id	Employee. name	salary	department_id	Department. name
1	Jones	26000	1	Toys
2	Smith	28000	1	Toys
2	Smith	28000	2	Tools
3	Parker	35000	3	Food
4	Smith	24000	3	Food

### Works\_in ⋈ Department



employee_id	department_id	name
1	1	Toys
2	1	Toys
2	2	Tools
3	3	Food
4	3	Food

Natural join evaluates 5\*3 = 15 combinations result temporary relation consists of 5 tuples and 3 columns.

#### **Employee**

employee_id	name	salary
1	Jones	26000
2	Smith	28000
3	Parker	35000
4	Smith	24000

#### Works in

employee_id	department_id		
1	1		
2	1		
2	2		
3	3		
4	3		

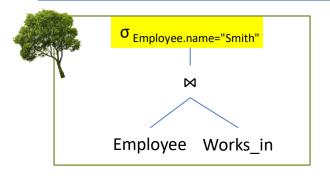
#### Department

department_id	name
1	Toys
2	Tools
3	Food

$$r \bowtie s = \pi_{R \cup S}(\sigma_{r,A1 = s,A1 \land r,A2 = s,A2 \land ... \land r,An = s,An}(r \times s))$$

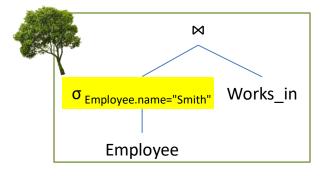
- Rule 5. The selection operation distributes over the natural join operation under the following two conditions
  - Rule 5a. It distributes when all the attributes in selection condition involve only the attributes of one of the expressions (say,  $E_1$ ) being joined.  $\sigma_p(E_1 \bowtie E_2) = (\sigma_p(E_1) \bowtie E_2)$

```
σ<sub>Employee.name="Smith"</sub>(
Employee ⋈ Works_in
)
```



 $(\sigma_{Employee.name="Smith"} (Employee) \bowtie Works_in)$ 

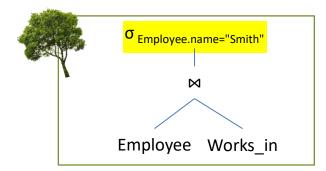
equivalent to





### **Expression tree A.**

σ<sub>Employee.name="Smith"</sub>(
Employee ⋈ Works\_in
)





employee_id	name	salary	department_id
2	Smith	28000	1
2	Smith	28000	2
4	Smith	24000	3



### **Employee** ⋈ Works\_in

employee_id	name	salary	department_id
1	Jones	26000	1
2	Smith	28000	1
2	Smith	28000	2
3	Parker	35000	3
4	Smith	24000	3



Natural join evaluates 4\*5 = 20 combinations result temporary relation consists of 5 tuples and 4 columns.

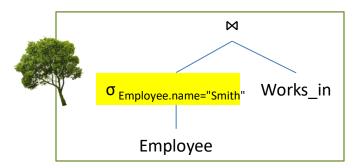
#### **Employee**

Works\_in

			_	
employee_id	name	salary	employee_id	department_id
1	Jones	26000	1	1
2	Smith	28000	2	1
3	Parker	35000	2	2
4	Smith	24000	3	3
			4	3

### **Expression tree B.**

 $\sigma_{\text{Employee.name="Smith"}} \text{ (Employee)} \bowtie \text{Works\_in}$ 





employee_id	name	salary	department_id
2	Smith	28000	1
2	Smith	28000	2
4	Smith	24000	3



Natural join evaluates 2\*5 = 10 combinations result temporary relation consists of 3 tuples and 4 columns.



employee_id	name	salary
2	Smith	28000
4	Smith	24000

#### **Employee**

name	salary
Jones	26000
Smith	28000
Parker	35000
Smith	24000
	Jones Smith Parker

#### Works in

employee_id	department_id	
1	1	
2	1	
2	2	
3	3	
4	3	

When comparing with the equivalence expression tree 1, we can see that if we push the selection predicates down the natural join (perform selection earlier than joining), then we will probably have a smaller temporary relation.

$$r \bowtie s = \pi_{R \cup S} (\sigma_{r,A1 = s,A1 \land r,A2 = s,A2 \land ... \land r,An = s,An} (r \times s))$$

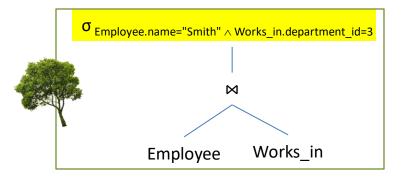
 $\bigcirc$  **Rule 5b.** The selection distributes when selection condition p1 involves only the attributes of  $E_1$  and p2 involves only the attributes of  $E_2$ .

$$\sigma_{p1 \wedge p2}(E_1 \bowtie E_2) = (\sigma_{p1}(E_1) \bowtie \sigma_{p2}(E_2))$$

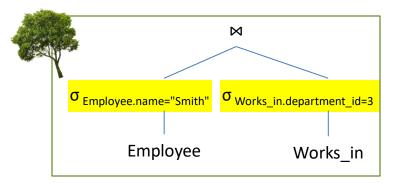
```
σ<sub>Employee.name="Smith" ∧ Works_in.department_id=3</sub>(
Employee ⋈ Works_in
)
```

```
( σ<sub>Employee.name="Smith"</sub> (Employee)

⋈
σ<sub>Works_in.department_id=3</sub> (Works_in))
```



### equivalent to

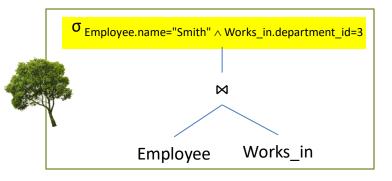


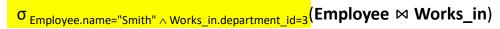


### **Expression tree A.**

σ<sub>Employee.name="Smith" ∧ Works\_in.department\_id=3</sub>(

Employee ⋈ Works\_in
)





employee_id	name	salary	department_id
4	Smith	24000	3



#### **Employee** ⋈ Works\_in

employee_id	name	salary	department_id
1	Jones	26000	1
2	Smith	28000	1
2	Smith	28000	2
3	Parker	35000	3
4	Smith	24000	3



Natural join evaluates 4\*5 = 20 combinations

#### **Employee**

Works\_in

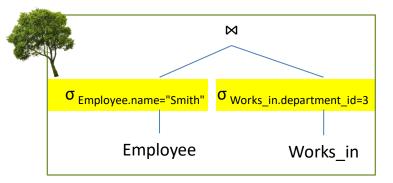
• •			_	
employee_id	name	salary	employee_id	department_id
1	Jones	26000	1	1
2	Smith	28000	2	1
3	Parker	35000	2	2
4	Smith	24000	3	3
			4	3

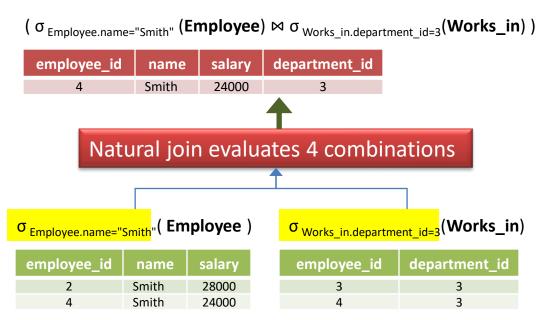


### **Expression tree B.**

( σ<sub>Employee.name="Smith"</sub> (Employee)

⋈
σ<sub>Works\_in.department\_id=3</sub> ( Works\_in ) )





#### **Employee**

name	salary
Jones	26000
Smith	28000
Parker	35000
Smith	24000
	Jones Smith Parker

#### Works in

_		
employee_id	department_id	
1	1	
2	1	
2	2	
3	3	
4	3	

When comparing with the equivalence expression 1, we can see that if we push the selection predicates down the natural join (perform selection earlier than joining), the natural join would consider fewer combinations.



$$r \bowtie s = \pi_{R \cup S}(\sigma_{r,A1 = s,A1 \land r,A2 = s,A2 \land ... \land r,An = s,An}(r \times s))$$

Rule 6. The projection operation can distribute over the natural join operation.

$$\pi_{L1 \cup L2} (E_1 \bowtie E_2) = \pi_{L1 \cup L2} ((\pi_{L1 \cup L3} (E_1)) \bowtie (\pi_{L2 \cup L3} (E_2)))$$

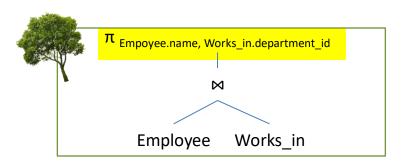
- Let L1 and L2 be some attributes from E1 and E2, respectively.
- $\bigcirc$  Let L3 be attributes that are involved in join condition, but are not in L1  $\cup$  L2.

 $\pi_{L1 \cup L2} (E_1 \bowtie E_2) = \pi_{L1 \cup L2} ((\pi_{L1 \cup L3} (E_1)) \bowtie (\pi_{L2 \cup L3} (E_2)))$ 

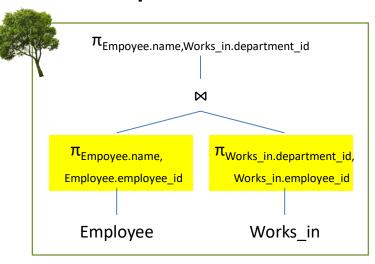
π<sub>Empoyee.name, Works\_in.department\_id</sub> (Employee ⋈ Works\_in)

- L1 = Empoyee.name
- L2 = Works\_in.department\_id
- L3 = employee\_id <</p>

The attribute used in natural join



### equivalent to

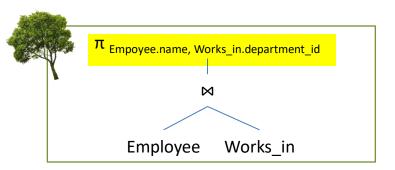




### **Expression tree A.**

π Empoyee.name, Works\_in.department\_id

(Employee ⋈ Works\_in)



π<sub>Empoyee.name, Works in.department id</sub> (Employee ⋈ Works\_in)

name	department_id			
Jones	1			
Smith	1			
Smith	2			
Parker	3			
Smith	3			

### Employee ⋈ Works\_in



employee_id	name	salary	department_id	since
1	Jones	26000	1	2012/1/1
2	Smith	28000	1	2011/3/2
2	Smith	28000	2	2014/2/1
3	Parker	35000	3	2013/2/2
4	Smith	24000	3	2013/2/8



Natural join evaluates 4\*5 = 20 combinations, result temporary relation consists of 5 columns.

### **Employee**

Works_ir	1
----------	---

employee_id	name	salary	employee_id	department_id	since
1	Jones	26000	1	1	2012/1/1
2	Smith	28000	2	1	2011/3/2
3	Parker	35000	2	2	2014/2/1
4	Smith	24000	3	3	2013/2/2
			4	3	2013/2/8



## **Expression tree B.**

 $\pi_{\text{ Empoyee.name, Works\_in.department\_id}}\text{(}\pi_{\text{Empoyee.name, Employee\_id}}\text{(}\text{Employee}\text{)}\bowtie$  $\pi_{Works\_in.department\_id, Works\_in.employee\_id}$  (Works\_in) )

name	department_id
Jones	1
Smith	1
Smith	2
Parker	3
Smith	3

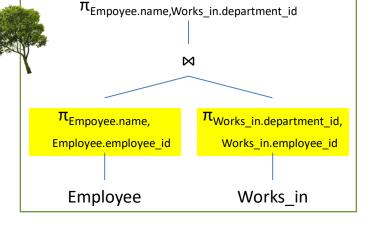
π Empoyee.name, Works in.department\_id (  $\pi_{\text{ Empoyee.name, Employee.employee\_id}} \text{ (Employee)}$ M  $\pi_{\text{ Works\_in.department\_id, Works\_in.employee\_id}}(\text{Works\_in})$ 

 $\pi_{\text{Empoyee.name, Employee.employee\_id}}\!($  Employee )  $\bowtie$ 

$\pi_{Works\_in.department}$	_id, Works_	_in.employee_id (Works_in)	

		–
employee_id	name	department_id
1	Jones	1
2	Smith	1
2	Smith	2
3	Parker	3
4	Smith	3

Natural join evaluates 4\*5 = 20 combinations. result temporary relation consists of 3 columns.



**Employee** 

### Works in

π<sub>Empoyee.name</sub>, Employee.employee\_id Employee ) π<sub>Works in.department id,</sub>

Works in.employee id (Works\_in)

1 /									
employee_id	name	salary	employee_id	department_id	since	employee_id	name	employee_id	department_id
1	Jones	26000	1	1	2012/1/1	1	Jones	1	1
2	Smith	28000	2	1	2011/3/2	2	Smith	2	1
3	Parker	35000	2	2	2014/2/1	3	Parker	2	2
4	Smith	24000	3	3	2013/2/2	4	Smith	3	3
37			4	3	2013/2/8			4	3
<i></i>									

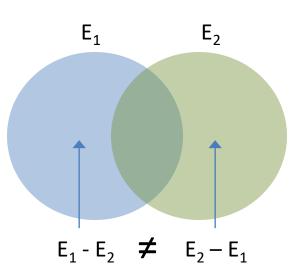
Rule 7. The set operations union and intersections are commutative.

$$E_1 \cup E_2 = E_2 \cup E_1$$

$$\mathsf{E}_1 \cap \mathsf{E}_2 = \mathsf{E}_2 \cap \mathsf{E}_1$$

The set different operation is NOT commutative

$$\mathsf{E}_1^{}-\mathsf{E}_2^{}\neq\mathsf{E}_2^{}-\mathsf{E}_1^{}$$

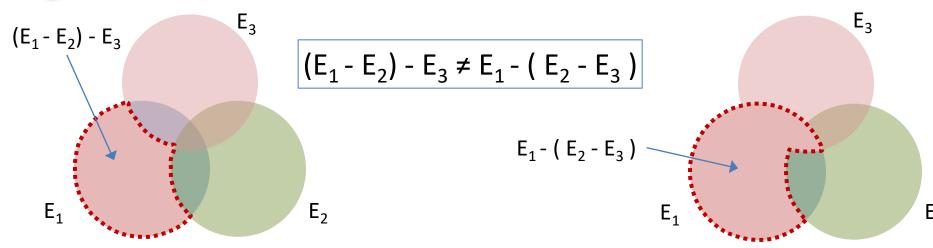


Rule 8. The set operations union and intersections are associative.

$$(E_1 \cup E_2) \cup E_3 = E_1 \cup (E_2 \cup E_3)$$

$$(\mathsf{E}_1 \cap \mathsf{E}_2) \cap \mathsf{E}_3 = \mathsf{E}_1 \cap (\mathsf{E}_2 \cap \mathsf{E}_3)$$

The set different operation is NOT associative.



Rule 9. The selection operation distributes over the union, intersection and set difference operations

$$\sigma_{p}(E_{1} \cup E_{2}) = \sigma_{p}(E_{1}) \cup \sigma_{p}(E_{2})$$

$$\sigma_{p}(E_{1} \cap E_{2}) = \sigma_{p}(E_{1}) \cap \sigma_{p}(E_{2})$$

$$\sigma_{p}(E_{1} - E_{2}) = \sigma_{p}(E_{1}) - \sigma_{p}(E_{2})$$

### Audio\_CD

ID	name	provider_id	stock	#tracks
CD1	One Heart	P1	55	14
CD2	Miracle	P2	4	14

### **DVD**

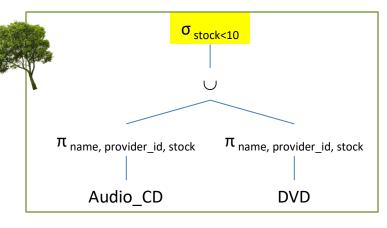
ID	name	provider_id	stock	length
DVD1	Prince of Persia	P2	3	110
DVD2	Iron man 3	P3	60	90
DVD3	Legend is born: Ip Man	P3	17	90

# $\frac{\sigma_{\text{stock}<10}}{\pi_{\text{name, provider\_id, stock}}} (\text{ Audio\_CD}) \cup \\ \pi_{\text{name, provider\_id, stock}} (\text{ DVD})$

### equivalent to

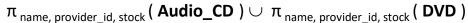
σ <sub>stock&lt;10</sub>	$_{0}$ ( $\pi$ $_{name, provider\_id, stock}$ ( $Audio\_CD$ )) $\cup$
σ <sub>stock&lt;10</sub>	$_{0}$ ( $\pi_{\text{name, provider_id, stock}}$ ( DVD ))

```
\sigma_{\text{stock} < 10} \text{ (} \\ \pi_{\text{name, provider\_id, stock}} \text{ ( Audio\_CD )} \cup \\ \pi_{\text{name, provider\_id, stock}} \text{ ( DVD )} \\ \text{)}
```



### $\sigma_{\text{stock}<10}(\pi_{\text{name, provider_id, stock}}(\text{Audio_CD}) \cup \pi_{\text{name, provider_id, stock}}(\text{DVD}))$

name	provider_id	stock
Miracle	P2	4
Prince of Persia	P2	3



name	provider_id	stock
One Heart	P1	55
Miracle	P2	4
Prince of Persia	P2	3
Iron man 3	P3	60
Legend is born: Ip Man	P3	17



### Audio\_CD

ID	name	provider_id	stock	#tracks
CD1	One Heart	P1	55	14
CD2	Miracle	P2	4	14

#### DVD

ID	name	provider_id	stock	length
DVD1	Prince of Persia	P2	3	110
DVD2	Iron man 3	P3	60	90
DVD3	Legend is born: Ip Man	Р3	17	90

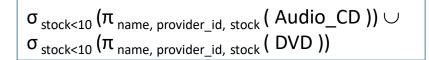
### $\pi_{\text{ name, provider\_id, stock}} \text{(Audio\_CD)}$

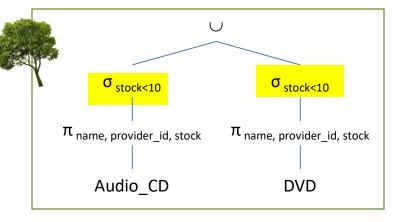
name	provider_id	stock
One Heart	P1	55
Miracle	P2	4

### $\pi_{name, provider\_id, stock}$ (DVD)

name	provider_id	stock
Prince of Persia	P2	3
Iron man 3	P3	60
Legend is born: Ip Man	Р3	17

 $\sigma_{\text{stock}<10}$  ( $\pi_{\text{name, provider_id, stock}}$  ( Audio\_CD ))  $\cup \sigma_{\text{stock}<10}$  ( $\pi_{\text{name, provider_id, stock}}$  ( DVD ))



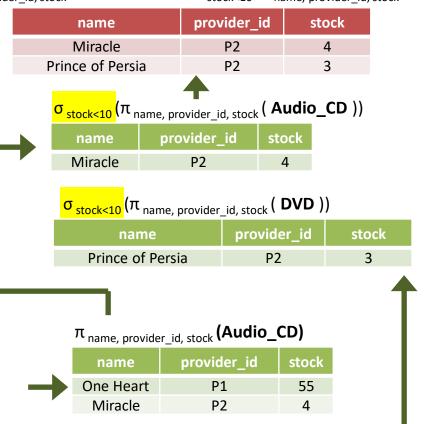


### Audio\_CD

ID	name	provider_id	stock	#tracks
CD1	One Heart	P1	55	14
CD2	Miracle	P2	4	14

### **DVD**

ID	name	provider_id	stock	length
DVD1	Prince of Persia	P2	3	110
DVD2	Iron man 3	P3	60	90
DVD3	Legend is born: Ip Man	Р3	17	90



 $\pi_{\text{ name, provider\_id, stock}} \text{(DVD)}$ 

name	provider_id	stock
Prince of Persia	P2	3
Iron man 3	Р3	60
Legend is born: Ip Man	P3	17

Rule 10. The projection operation distributes over the union operation

$$\pi_L(E_1 \cup E_2) = \pi_L(E_1) \cup \pi_L(E_2)$$

Projection does not distribute over intersection and set difference.

$$\pi_L (E_1 \cap E_2) \neq \pi_L (E_1) \cap \pi_L (E_2)$$

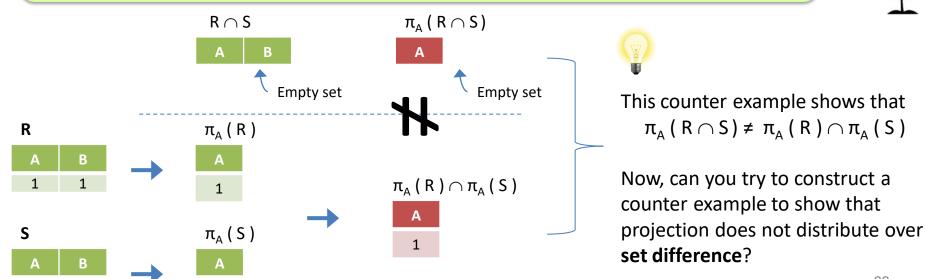
$$\pi_L (E_1 - E_2) \neq \pi_L (E_1) - \pi_L (E_2)$$



Why projection does not distribute over intersection and set difference?

To show that projection does not distribute over intersection, you only need to provide a counter example.





# Section 6

# Example of query optimization

Find the names of all instructors in the CS department (dpt\_id = 1) who have taught a course in 2<sup>nd</sup> semester, together with the course title of all the courses that the instructors teach.

```
SELECT I.name, C.title
FROM Instructor I, Teaches T, Course C
WHERE I.dpt_id = 1 AND
T.sem=2 AND
I.instructor_id = T.instructor_id AND
T.course_id = C.course_id;
```

```
\begin{array}{c} \pi_{\text{I.name,C.title}} \, ( \\ \sigma_{\text{I.dpt\_id} = 1 \ \land \ \text{T.sem} = 2} ( \\ \rho_{\text{I}} ( \text{Instructor}) \bowtie ( \rho_{\text{T}} ( \text{Teaches}) \bowtie \rho_{\text{C}} ( \text{Course}) ) \\ ) \\ ) \end{array}
```

Relational algebra

Course

SQL

Instructor					
instructor_id	name	dpt_id			
1	Kit	1			
2	Ben	1			
3	Michael	2			
4	William	3			

reacties					
instructor_id	course_id	sem			
1	1	1			
1	2	2			
2	4	1			
3	3	2			

course_id	title	credit
1	Intro to DB	6
2	Programming I	6
3	Accounting	6
4	Algorithms	6

```
\begin{array}{c} \pi_{l.name,C.title} (\\ \sigma_{l.dpt\_id=1 \ \land \ T.sem=2} (\\ \rho_{l} (\ Instructor) \bowtie (\rho_{T} (\ Teaches) \bowtie \rho_{C} (\ Course)) \end{array} \\ ) \\ ) \\ to \\ \begin{array}{c} \pi_{l.name,C.title} (\\ \sigma_{l.dpt\_id=1 \ \land \ T.sem=2} (\\ \rho_{l} (\ Instructor) \bowtie \\ (\rho_{T} (\ Teaches) \bowtie \pi_{C.course\_id,\ C.title} (\rho_{C} (\ Course))) \\ ) \\ ) \\ \end{array}
```

# BEFORE π<sub>I.name,C.title</sub> σ<sub>I.dpt\_id=1 ∧ T.sem=2</sub> Ν Γ C

### Rule 6

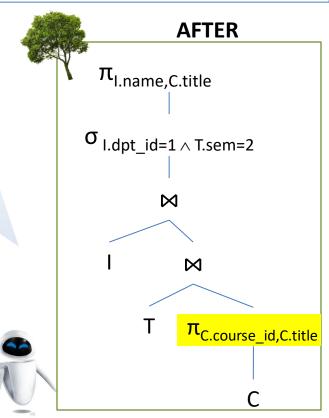
```
\pi_{L1 \cup L2}(E_1 \bowtie E_2) = 

\pi_{L1 \cup L2}((\pi_{L1 \cup L3}(E_1)) \bowtie (\pi_{L2 \cup L3}(E_2)))
```

Let's try to push the projection  $\pi_{\text{C.title}}$  downward and apply it ahead of the natural joins.

Since this natural join requires C.course\_id = T.course\_id, therefore, we have to add the joining attribute C.course\_id to make the projection

π<sub>C.course\_id,C.title</sub>

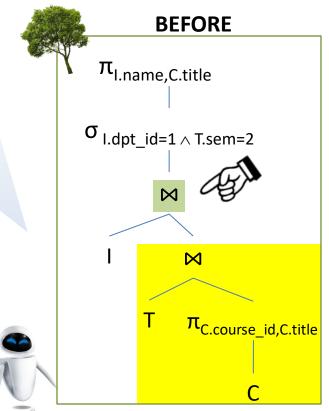


```
\begin{array}{c} \pi_{l.name,C.title}(\\ \sigma_{l.dpt\_id=1 \ \land \ T.sem=2}(\\ (\rho_{l}(\ Instructor) \bowtie \rho_{T}(\ Teaches)\ )\\ \bowtie \pi_{C.course\_id,\ C.title}(\ \rho_{c}(\ Course))\\ )\\ )\\ \end{array} equivalent to \begin{array}{c} \pi_{l.name,C.title}(\\ \sigma_{l.dpt\_id=1 \ \land \ T.sem=2}(\\ \rho_{l}(\ Instructor) \bowtie \\ (\rho_{T}(\ Teaches) \bowtie \pi_{C.course\_id,\ C.title}(\rho_{C}(\ Course)))\\ )\\ )\\ )\\ \end{array}
```

# AFTER π<sub>l.name,C.title</sub> σ<sub>l.dpt\_id=1 \ T.sem=2</sub> μ π<sub>C.course\_id,C.title</sub> Γ Γ

### $(E_1 \bowtie E_2) \bowtie E_3 = E_1 \bowtie (E_2 \bowtie E_3)$

- Now we would like to push the selection down to reduce the size of the temporary result of the natural join.
- As the selection involves relations I and T only, we would like to rearrange the natural joins to make I and T under one natural join.
- Since natural joins are associative, we can make such rearrangement.

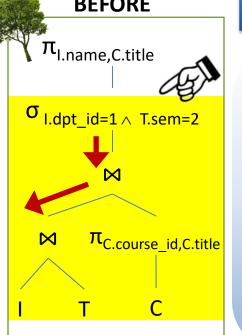


```
\pi_{l.name.C.title} (
     \sigma_{\text{I.dpt id}=1 \land \text{T.sem}=2}
             (\rho_{I}(Instructor) \bowtie \rho_{T}(Ieaches))
                \bowtie \pi_{\text{C.course id, C.title}} (\rho_{\text{C}} (Course))
```

```
equivalent
    to
```

```
\pi_{l,name,C,title} (
       \sigma_{\text{I.dpt id}=1} \wedge \tau_{\text{I.sem}=2}
                    \rho_{I}(Instructor) \bowtie \rho_{T}(Teaches)
       \bowtie \ \pi_{\text{ C.course\_id, C.title}} \text{( } \rho_{\text{C}}\text{( Course))}
```

### **BEFORE**

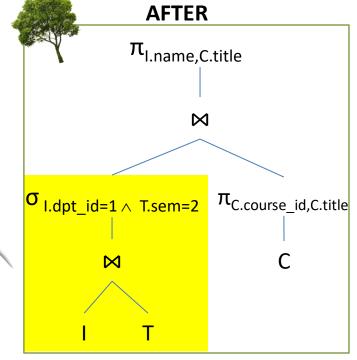


### Rule 5a

### $\sigma_{p1}(E_1 \bowtie E_2) = (\sigma_{p1}(E_1) \bowtie E_2)$

Now we can push the

selection down one level. According to Rule 5a, we can distribute both selection predicates to the L.H.S. of the selection as the R.H.S. does not contain any attribute in the selection predicate.



```
\pi_{l,name,C,title}
       (\sigma_{I.dpt_{id}=1}(\rho_I(Instructor))
             \sigma_{\text{T.sem}=2} ( \rho_{\text{T}}( Teaches) )
       \bowtie \ \pi_{\text{ C.course\_id, C.title}} \text{( } \rho_{\text{C}}\text{( Course) )}
```

```
equivalent
    to
```

```
\pi_{l,name,C,title} (
       \sigma_{\text{I.dpt id}=1} \wedge \tau_{\text{I.sem}=2}
                  \rho_{I}( Instructor) \bowtie \rho_{T}( Teaches)
       \bowtie \pi _{\text{C.course\_id, C.title}} ( \rho_{\text{C}}( Course))
```

# **AFTER** $\pi_{\text{I.name,C.title}}$ M M $\pi_{\text{C.course id,C.title}}$ $\sigma_{\text{I.dpt\_id=1}} \sigma_{\text{T.sem=2}}$

### Rule 5b

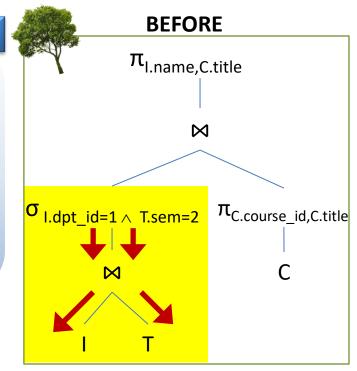
 $\overline{\left(\sigma_{p1}, \rho_{p2}\right)} = \overline{\left(\sigma_{p1}, \overline{E_1}\right)} \bowtie \overline{\sigma_{p1}}$ 

Now we can further push the selection one more level down by applying rule 5b.

According to Rule 5b, we can distribute

- $\sigma_{I,dpt id=1}$  to I
- $\sigma_{T.sem=2}$  to T





# Illustration (original tree)

```
\begin{array}{c} \pi_{l.name,C.title} (\\ \sigma_{l.dpt\_id = 1 \ \land \ T.sem = 2} (\\ \rho_{l} (\ Instructor) \bowtie \\ (\ \rho_{T} (\ Teaches) \bowtie \rho_{C} (\ Course) \ ) \\ ) \end{array}
```



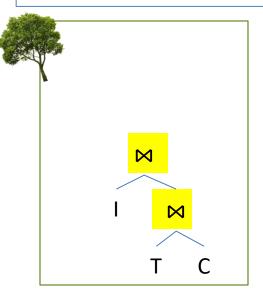
instructor_id	name	dpt_id	course_id	sem	title	credit
1	Kit	1	1	1	Intro to DB	6
1	Kit	1	2	2	Programming I	6
2	Ben	1	4	1	Algorithms	6
3	Michael	2	3	2	Accounting	6



### $\rho_{\mathsf{T}}(\mathsf{Teaches}) \bowtie \rho_{\mathsf{C}}(\mathsf{Course})$

instructor_id	name	dpt_id	course_id	sem
1	Kit	1	1	1
1	Kit	1	2	2
2	Ben	1	4	1
3	Michael	2	3	2

Natural join evaluates 4\*4 = 16 combinations.



### Instructor

instructor_id	name	dpt_id
1	Kit	1
2	Ben	1
3	Michael	2
4	William	3

### **Teaches**

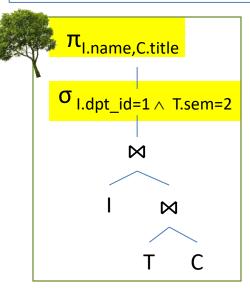
instructor_id	course_id	sem
1	1	1
1	2	2
2	4	1
3	3	2

### Course

course_id	title	credit
1	Intro to DB	6
2	Programming I	6
3	Accounting	6
4	Algorithms	6

# Illustration (original tree)

```
\begin{array}{c} \pi_{l.name,C.title} (\\ \sigma_{l.dpt\_id = 1 \ \land \ T.sem = 2} (\\ \rho_l (\ \textbf{Instructor}) \bowtie \\ (\ \rho_T (\ \textbf{Teaches}) \bowtie \rho_C (\ \textbf{Course}) \ ) \\ ) \end{array}
```



### $\rho_{I}$ ( Instructor) $\bowtie$ ( $\rho_{T}$ ( Teaches) $\bowtie \rho_{C}$ ( Course) )

instructor_id	name	dpt_id	course_id	sem	title	credit
1	Kit	1	1	1	Intro to DB	6
1	Kit	1	2	2	Programming I	6
2	Ben	1	4	1	Algorithms	6
3	Michael	2	3	2	Accounting	6



### $\sigma_{\text{I.dpt\_id} = 1 \land \text{T.sem} = 2}$ ( $\rho_{\text{I}}$ ( Instructor) $\bowtie$ ( $\rho_{\text{T}}$ ( Teaches) $\bowtie$ $\rho_{\text{C}}$ ( Course) ))

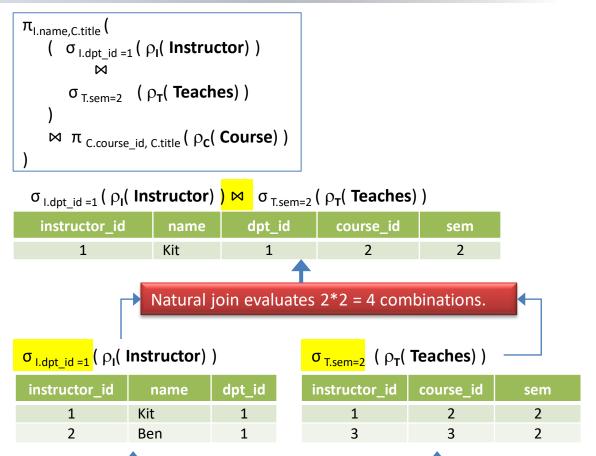
instructor_id	name	dpt_id	course_id	sem	title	credit
1	Kit	1	2	2	Programming I	6

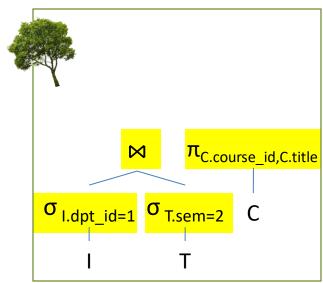


 $\pi_{\text{I.name,C.title}}$  (  $\sigma_{\text{I.dpt\_id}=1 \ \land \ \text{T.sem=2}}$  (  $\rho_{\text{I}}$  ( Instructor)  $\bowtie$  (  $\rho_{\text{T}}$  ( Teaches)  $\bowtie \ \rho_{\text{C}}$  ( Course) )))

name	title
Kit	Programming I

# Illustration (transformed tree)





C.course_id, C.title ( Pc( Course) )				
course_id	title			
1	Intro to DB			
2	Programming I			
3	Accounting			
4	Algorithms			

( o ( Course) )

Course		
course_id	title	credit
1	Intro to DB	6
2	Programming I	6
3	Accounting	6
4	Algorithms	6

instructor id

course id

sem

**Teaches** 

dpt id

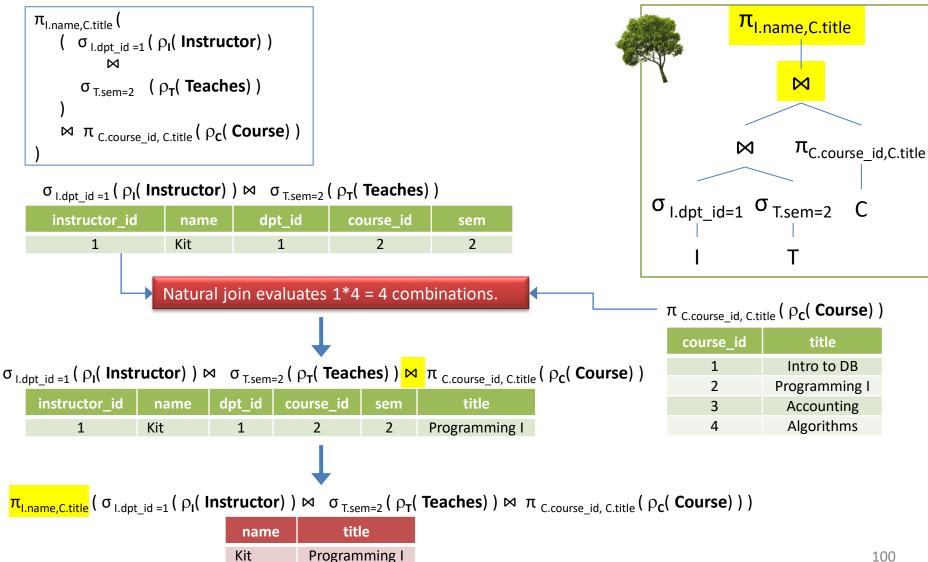
name

William

Instructor

instructor id

# Illustration (transformed tree)



# Summary

- Relational algebra (RA) defines a set of algebraic operations on tables, and output tables as result.
  - 6 fundamental operations
  - Additional operations does not extend the power of the fundamental operators, but they simplify the expression.
  - Extended operations add expressive power.
- Relational algebra (RA) is the basics of query optimization.

# Lecture 6

# END

### **COMP3278**

# Introduction to Database Management Systems

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