# Introduction to Data Management CSE 344 

## Lectures 8: Relational Algebra

## Announcements

- Homework 3 is posted
- Microsoft Azure Cloud services!
- Use the promotion code you received
- Due on 2/1
- Make sure you read the textbook!
- Very good coverage of RA


## Where We Are

- Data models
- SQL, SQL, SQL
- Declaring the schema for our data (CREATE TABLE)
- Inserting data one row at a time or in bulk (INSERT/.import)
- Querying the data (SELECT)
- Modifying the schema and updating the data (ALTER/UPDATE)
- Next step: More knowledge of how DBMSs work
- Relational algebra, query execution, and physical tuning
- Client-server architecture


## Query Evaluation Steps



## The WHAT and the HOW

- $\operatorname{SQL}=$ WHAT we want to get from the data
- Relational Algebra $=\mathrm{HOW}$ to get the data we want
- The passage from WHAT to HOW is called query optimization
- SQL $\rightarrow$ Logical Plan $\rightarrow$ Physical Plan
- Logical plan expressed using relational algebra


## Relational Algebra

## Turing Awards in Data Management



Charles Bachman, 1973 IDS and CODASYL

Ted Codd, 1981
Relational model


Jim Gray, 1998
Transaction processing

Michael Stonebraker, 2014 INGRES and Postgres

## Sets v.s. Bags

- Sets: \{a,b,c\}, \{a,d,e,f\}, \{\}, . . .
- Bags: $\{a, a, b, c\},\{b, b, b, b, b\}, \ldots$

Relational Algebra has two semantics:

- Set semantics = standard Relational Algebra
- Bag semantics = extended Relational Algebra

DB systems implement bag semantics (Why?)

## Relational Algebra Operators

- Union $\cup$, intersection $\cap$, difference
- Selection $\sigma$
- Projection $\pi$
- Cartesian product $X$, join $\bowtie$
- Rename $\rho$
- Duplicate elimination $\delta$
- Grouping and aggregation $\gamma$

Extended RA

- Sorting $\tau$

All operators take in 1 or more relations as inputs and return another relation

## Union and Difference

## R1 U R2 R1 - R2

What do they mean over bags ?

## What about Intersection?

- Derived operator using minus

$$
R 1 \text { R2 = R1 - (R1 - R2) }
$$

- Only makes sense if result is $\geq 0$
- Derived using join

$$
R 1 \cap R 2=R 1 \bowtie R 2
$$

- Only makes sense if R1 and R2 have the same schema


## Selection

- Returns all tuples which satisfy a condition

$$
\sigma_{\mathrm{c}}(\mathrm{R})
$$

- Examples

- The condition c can be =, <, <=, >, >=, <> combined with AND, OR, NOT


## Employee

| SSN | Name | Salary |
| :---: | :---: | :---: |
| 1234545 | John | 20000 |
| 5423341 | Smith | 60000 |
| 4352342 | Fred | 50000 |

$\sigma_{\text {salay }>40000}$ (Employee)

| SSN | Name | Salary |
| :---: | :---: | :---: |
| 5423341 | Smith | 60000 |
| 4352342 | Fred | 50000 |

## Projection

- Eliminates columns

$$
\Pi_{\mathrm{A} 1, \ldots, \mathrm{An}}(\mathrm{R})
$$

- Example: project social-security number and names:
$-\Pi_{\text {SSN, Name }}(E m p l o y e e) \rightarrow$ Answer(SSN, Name)

Different semantics over sets or bags! Why?

Employee

| SSN | Name | Salary |
| :---: | :---: | :---: |
| 1234545 | John | 20000 |
| 5423341 | John | 60000 |
| 4352342 | John | 20000 |

$\pi_{\text {Name,Salary }}$ (Employee)

| Name | Salary |
| :---: | :---: |
| John | 20000 |
| John | 60000 |
| John | 20000 |


| Name | Salary |
| :---: | :---: |
| John | 20000 |
| John | 60000 |

Bag semantics
Set semantics
Which is more efficient?

## Composing RA Operators

## Patient

| no | name | zip | disease |
| :--- | :--- | :--- | :--- |
| 1 | p1 | 98125 | flu |
| 2 | p2 | 98125 | heart |
| 3 | p3 | 98120 | lung |
| 4 | p4 | 98120 | heart |

$\Pi_{\text {zip,disease }}$ (Patient)

| zip | disease |
| :--- | :--- |
| 98125 | flu |
| 98125 | heart |
| 98120 | lung |
| 98120 | heart |

$\sigma_{\text {disease='heart' }}$ (Patient)

| no | name | zip | disease |
| :--- | :--- | :--- | :--- |
| 2 | p2 | 98125 | heart |
| 4 | p4 | 98120 | heart |

$\Pi_{\text {zip,disease }}\left(\sigma_{\text {disease='heart' }}(\right.$ Patient $\left.)\right)$

| zip | disease |
| :--- | :--- |
| 98125 | heart |
| 98120 | heart |

## Cartesian Product

- Each tuple in R1 with each tuple in R2


## R1 X R2

- Rare in practice; mainly used to express joins


## Cross-Product Example

## Employee

| Name | SSN |
| :--- | :--- |
| John | 999999999 |
| Tony | 777777777 |

Dependent

| EmpSSN | DepName |
| :--- | :--- |
| 999999999 | Emily |
| 777777777 | Joe |

## Employee X Dependent

| Name | SSN | EmpSSN | DepName |
| :--- | :--- | :--- | :--- |
| John | 999999999 | 999999999 | Emily |
| John | 999999999 | 777777777 | Joe |
| Tony | 777777777 | 999999999 | Emily |
| Tony | 777777777 | 777777777 | Joe |

## Renaming

- Changes the schema, not the instance

$$
\rho_{\mathrm{B} 1, \ldots, \mathrm{Bn}}(\mathrm{R})
$$

- Example:
- Given Employee(Name, SSN)
$-\rho_{\mathrm{N}, \mathrm{s}}($ Employee) $\rightarrow$ Answer(N, S)
Not really used by systems, but needed on paper


## Natural Join

## R1 $\bowtie$ R2

- Meaning: $\mathrm{R} 1 \bowtie \mathrm{R} 2=\Pi_{A}\left(\sigma_{\theta}(\mathrm{R} 1 \times \mathrm{R} 2)\right)$
- Where:
- Selection $\sigma_{\theta}$ checks equality of all common attributes (i.e., attributes with same names)
- Projection $\Pi_{\mathrm{A}}$ eliminates duplicate common attributes


## Natural Join Example

R

| $\mathbf{A}$ | $\mathbf{B}$ |
| :---: | :---: |
| $X$ | $Y$ |
| $X$ | $Z$ |
| $Y$ | $Z$ |
| $Z$ | $V$ |

S | $\mathbf{B}$ | $\mathbf{C}$ |
| :---: | :---: |
| $Z$ | $U$ |
| $V$ | $W$ |
| $z$ | $V$ |

$\mathbf{R} \bowtie \mathbf{S}=$
$\Pi_{A B C}\left(\sigma_{\text {R.B }=\text { S } . B}(R \times S)\right)$

| A | B | C |
| :---: | :---: | :---: |
| $X$ | $Z$ | $U$ |
| $X$ | $Z$ | $V$ |
| $Y$ | $Z$ | $U$ |
| $Y$ | $Z$ | $V$ |
| $Z$ | $V$ | $W$ |

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## Natural Join Example 2

Anon Patient $P$

| age | zip | disease |
| :--- | :--- | :--- |
| 54 | 98125 | heart |
| 20 | 98120 | flu |

Voters V

| name | age | zip |
| :--- | :--- | :--- |
| p1 | 54 | 98125 |
| p2 | 20 | 98120 |

$P \bowtie V$
join predicate:

| age | zip | disease | name |
| :--- | :--- | :--- | :--- |
| 54 | 98125 | heart | p1 |
| 20 | 98120 | flu | p2 |

$$
\begin{aligned}
& P \cdot \operatorname{age}=V . \text { age } \\
& A N D \\
& P \cdot \operatorname{iip}=V . z i p
\end{aligned}
$$

## Natural Join

- Given schemas R(A, B, C, D), S(A, C, E), what is the schema of $R \bowtie S$ ?
- Given $R(A, B, C), S(D, E)$, what is $R \bowtie S$ ?
- Given $R(A, B), S(A, B)$, what is $R \bowtie S$ ?

AnonPatient (age, zip, disease)
Voters (name, age, zip)

## Theta Join

- A join that involves a predicate

$$
R 1 \bowtie_{\theta} R 2=\sigma_{\theta}(R 1 \times R 2)
$$

- Here $\theta$ can be any condition
- No projection in this case!
- For our voters/patients example:
$\mathrm{P} \bowtie_{\text {P.zip }}$ 于 V .zip and P.age $>=\mathrm{V}$.age -1 and P.age $<=\mathrm{V}$.age +1 V


## Equijoin

- A theta join where $\theta$ is an equality predicate
- Projection drops all redundant attributes

$$
\text { R1 } \bowtie_{\theta} R 2=\pi_{A}\left(\sigma_{\theta}(R 1 \times R 2)\right)
$$

- By far the most used variant of join in practice
- What is the relationship with natural join?


## Equijoin Example

AnonPatient $P$

| age | zip | disease |
| :--- | :--- | :--- |
| 54 | 98125 | heart |
| 20 | 98120 | flu |

Voters V

| name | age | zip |
| :--- | :--- | :--- |
| p1 | 54 | 98125 |
| p2 | 20 | 98120 |

$\mathrm{P} \bowtie_{\text {P.age }} \mathrm{Y}_{\mathrm{Y} . \text { age }} \mathrm{V}$

| age | P.zip | disease | name | V.zip |
| :--- | :--- | :--- | :--- | :--- |
| 54 | 98125 | heart | p1 | 98125 |
| 20 | 98120 | flu | p2 | 98120 |

## Join Summary

- Theta-join: $R \bowtie_{\theta} S=\sigma_{\theta}(R \times S)$
- Join of $R$ and $S$ with a join condition $\theta$
- Cross-product followed by selection $\theta$
- Equijoin: $R \bowtie_{\theta} S=\pi_{A}\left(\sigma_{\theta}(R \times S)\right)$
- Join condition $\theta$ consists only of equalities
- Projection $\pi_{A}$ drops all redundant attributes
- Natural join: $R \bowtie S=\pi_{A}\left(\sigma_{\theta}(R x S)\right)$
- Equality on all fields with same name in $R$ and in $S$
- Projection $\pi_{A}$ drops all redundant attributes


## So Which Join Is It ?

When we write $R \bowtie S$ we usually mean an equijoin, but we often omit the equality predicate when it is clear from the context

## More Joins

- Outer join
- Include tuples with no matches in the output
- Use NULL values for missing attributes
- Does not eliminate duplicate columns
- Variants
- Left outer join
- Right outer join
- Full outer join


## Outer Join Example

## AnonPatient $P$

| age | zip | disease |
| :--- | :--- | :--- |
| 54 | 98125 | heart |
| 20 | 98120 | flu |
| 33 | 98120 | lung |

AnnonJob J

| job | age | zip |
| :--- | :--- | :--- |
| lawyer | 54 | 98125 |
| cashier | 20 | 98120 |


| $P$ 二 $\triangle J$ | P.age | P.zip | disease | job | J.age | J.zip |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | 54 | 98125 | heart | lawyer | 54 | 98125 |
|  | 20 | 98120 | flu | cashier | 20 | 98120 |
| $\cdots$ RoJ | 33 | 98120 | lung | null | 33 | 98120 |
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## Some Examples

```
Supplier(sno,sname,scity,sstate)
Part(pno,pname,psize,pcolor)
Supply(sno,pno,qty,price)
```

Name of supplier of parts with size greater than 10
$\Pi_{\text {sname }}\left(\right.$ Supplier $\bowtie$ Supply $\bowtie\left(\sigma_{\text {psize>10 }}(\right.$ Part $\left.)\right)$
Name of supplier of red parts or parts with size greater than 10 $\Pi_{\text {sname }}\left(\right.$ Supplier $\bowtie$ Supply $\bowtie\left(\sigma_{\text {psize }>10}(\right.$ Part $) \cup \sigma_{\text {pcolor='red' }}($ Part $\left.\left.)\right)\right)$

Can be represented as trees as well (as seen from last class)

