#### CSE 444: Database Internals

Lecture 10

Query Optimization (part 1)

### Reminders

Lab 2 Part 1 due tonight, 11:00pm

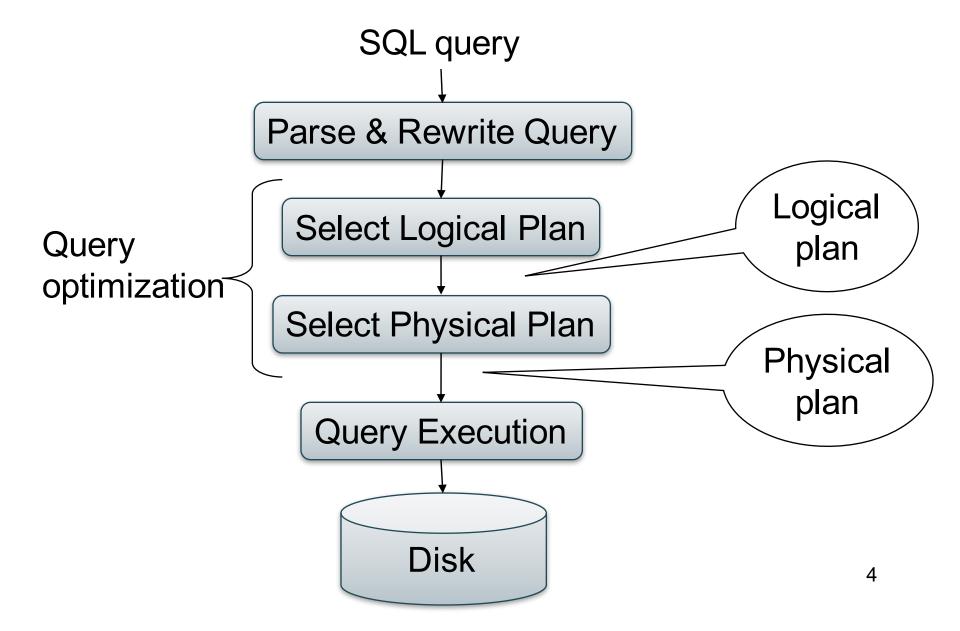
Homework 2 due on Monday, 11:00pm

### Know how to compute the cost of a plan

Next: Find a good plan automatically?

This is the role of the query optimizer

### **Query Optimization Overview**



### What We Already Know...

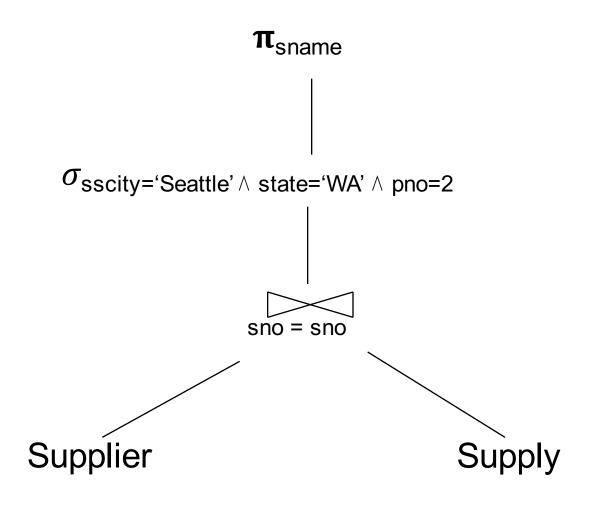
```
Supplier(sno,sname,scity,sstate)
Part(pno,pname,psize,pcolor)
Supply(sno,pno,price)
```

#### For each SQL query....

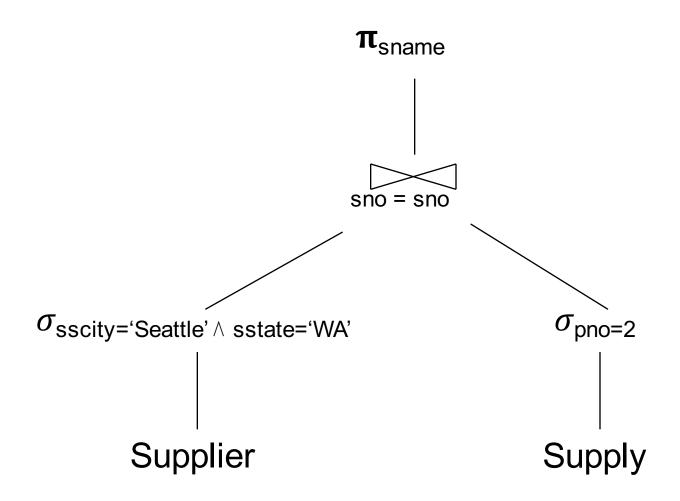
```
SELECT S.sname
FROM Supplier S, Supply U
WHERE S.scity='Seattle' AND S.sstate='WA'
AND S.sno = U.sno
AND U.pno = 2
```

There exist many logical query plan...

# Example Query: Logical Plan 1



## Example Query: Logical Plan 2

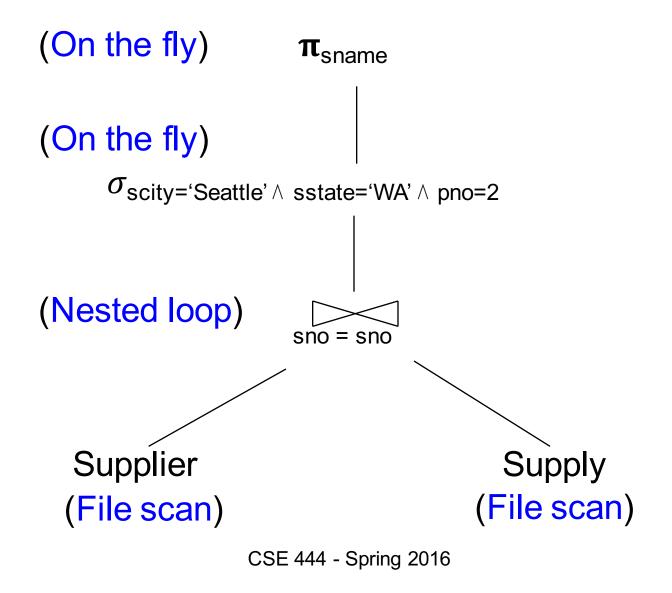


### What We Also Know

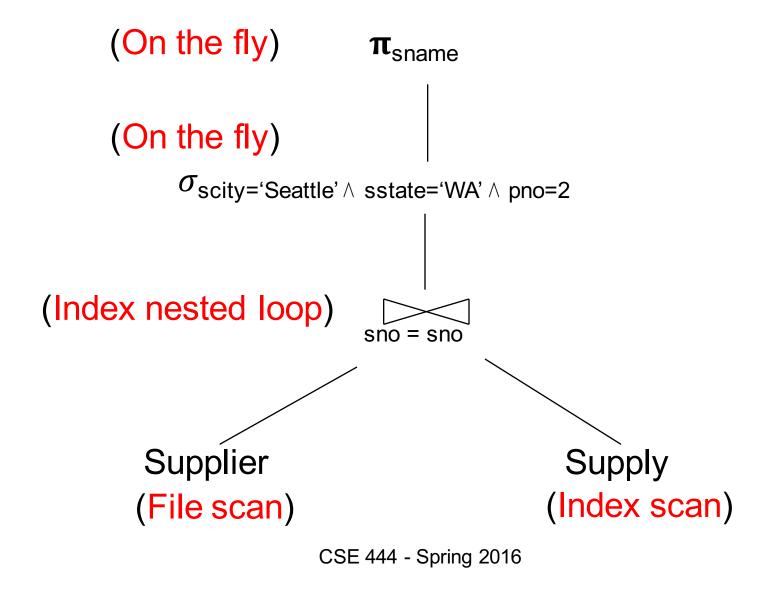
For each logical plan...

There exist many physical plans

### Example Query: Physical Plan 1



### Example Query: Physical Plan 2



### **Query Optimizer Overview**

- Input: A logical query plan
- Output: A good physical query plan
- Basic query optimization algorithm
  - Enumerate alternative plans (logical and physical)
  - Compute estimated cost of each plan
    - Compute number of I/Os
    - Optionally take into account other resources
  - Choose plan with lowest cost
  - This is called cost-based optimization

### Lessons

- No magic "best" plan: depends on the data
- In order to make the right choice
  - Need to have statistics over the data
  - The B's, the T's, the V's
  - Commonly (and in SimpleDB): histograms over base data

### Outline

Search space

Algorithm for enumerating query plans

### Relational Algebra Equivalences

#### Selections

- Commutative:  $\sigma_{c1}(\sigma_{c2}(R))$  same as  $\sigma_{c2}(\sigma_{c1}(R))$
- Cascading:  $\sigma_{c1 \land c2}(R)$  same as  $\sigma_{c2}(\sigma_{c1}(R))$

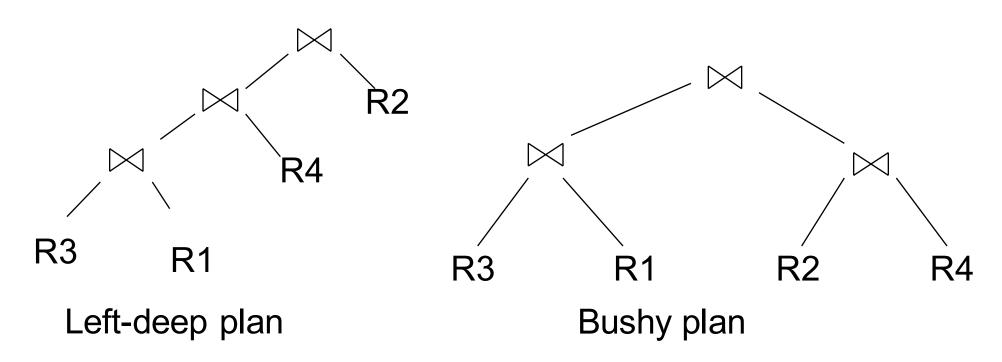
#### Projections

Cascading

#### Joins

- Commutative : R ⋈ S same as S ⋈ R
- Associative:  $R \bowtie (S \bowtie T)$  same as  $(R \bowtie S) \bowtie T$

# Left-Deep Plans, Bushy Plans, and Linear Plans



Linear plan: One input to each join is a relation from disk Can be either left or right input

# Commutativity, Associativity, Distributivity

$$R \cup S = S \cup R$$
,  $R \cup (S \cup T) = (R \cup S) \cup T$   
 $R \bowtie S = S \bowtie R$ ,  $R \bowtie (S \bowtie T) = (R \bowtie S) \bowtie T$ 

$$R\bowtie (S\cup T) = (R\bowtie S)\cup (R\bowtie T)$$

### Laws Involving Selection

$$\sigma_{CANDC'}(R) = \sigma_{C}(\sigma_{C'}(R)) = \sigma_{C}(R) \cap \sigma_{C'}(R)$$
 $\sigma_{CORC'}(R) = \sigma_{C}(R) \cup \sigma_{C'}(R)$ 
 $\sigma_{C}(R \bowtie S) = \sigma_{C}(R) \bowtie S$ 

$$\sigma_{c}(R - S) = \sigma_{c}(R) - S$$
  
 $\sigma_{c}(R \cup S) = \sigma_{c}(R) \cup \sigma_{c}(S)$   
 $\sigma_{c}(R \bowtie S) = \sigma_{c}(R) \bowtie S$ 

Assuming C on attributes of R

# Example: Simple Algebraic Laws

• Example: R(A, B, C, D), S(E, F, G)

$$\sigma_{F=3}(R\bowtie_{D=E}S)=$$

$$\sigma_{A=5 \text{ AND } G=9} (R \bowtie_{D=E} S) = G$$

# Example: Simple Algebraic Laws

• Example: R(A, B, C, D), S(E, F, G)

$$\sigma_{F=3}(R\bowtie_{D=E}S)=R\bowtie_{D=E}\sigma_{F=3}(S)$$

$$\sigma_{A=5 \text{ AND G}=9} (R \bowtie_{D=E} S) = \sigma_{A=5} (R) \bowtie_{D=E} \sigma_{G=9} (S)$$

## Laws Involving Projections

$$\Pi_{M}(R \bowtie S) = \Pi_{M}(\Pi_{P}(R) \bowtie \Pi_{Q}(S))$$

$$\Pi_{M}(\Pi_{N}(R)) = \Pi_{M}(R)$$
/\* note that  $M \subseteq N$  \*/

• Example R(A,B,C,D), S(E, F, G)  $\Pi_{A,B,G}(R\bowtie_{D=E}S) = \Pi_{?}(\Pi_{?}(R)\bowtie_{D=E}\Pi_{?}(S))$ 

### Laws Involving Projections

$$\Pi_{M}(R \bowtie S) = \Pi_{M}(\Pi_{P}(R) \bowtie \Pi_{Q}(S))$$

$$\Pi_{M}(\Pi_{N}(R)) = \Pi_{M}(R)$$
/\* note that  $M \subseteq N$  \*/

• Example R(A,B,C,D), S(E, F, G)

$$\Pi_{A,B,G}(R \bowtie_{D=E} S) = \Pi_{A,B,G} (\Pi_{A,B,D}(R) \bowtie_{D=E} \Pi_{E,G}(S))$$

# Laws involving grouping and aggregation

$$\gamma_{A, \text{ agg}(D)}(R(A,B) \bowtie_{B=C} S(C,D)) =$$
  
 $\gamma_{A, \text{ agg}(D)}(R(A,B) \bowtie_{B=C} (\gamma_{C, \text{ agg}(D)} S(C,D)))$ 

# Laws involving grouping and aggregation

$$\delta(\gamma_{A, \text{agg}(B)}(R)) = \gamma_{A, \text{agg}(B)}(R)$$

$$\gamma_{A, \text{ agg(B)}}(\delta(R)) = \gamma_{A, \text{ agg(B)}}(R)$$
if agg is "duplicate insensitive"

Which of the following are "duplicate insensitive"? sum, count, avg, min, max

## Laws Involving Constraints

Foreign key

Product(<u>pid</u>, pname, price, cid) Company(<u>cid</u>, cname, city, state)

 $\Pi_{\text{pid, price}}(\text{Product} \bowtie_{\text{cid=cid}} \text{Company}) = \Pi_{\text{pid, price}}(\text{Product})$ 

### Search Space Challenges

- Search space is huge!
  - Many possible equivalent trees
  - Many implementations for each operator
  - Many access paths for each relation
    - File scan or index + matching selection condition
- Cannot consider ALL plans
  - Heuristics: only partial plans with "low" cost

### Outline

Search space

Algorithm for enumerating query plans

### **Key Decisions**

#### Logical plan

- What logical plans do we consider (left-deep, bushy?); Search Space
- Which algebraic laws do we apply, and in which context(s)?; Optimization rules
- In what order do we explore the search space?;
   Optimization algorithm

### **Key Decisions**

#### Physical plan

What physical operators to use?

What access paths to use (file scan or index)?

Pipeline or materialize intermediate results?

These decisions also affect the search space

### Two Types of Optimizers

- Heuristic-based optimizers:
  - Apply greedily rules that always improve plan
    - Typically: push selections down
  - Very limited: no longer used today
- Cost-based optimizers:
  - Use a cost model to estimate the cost of each plan
  - Select the "cheapest" plan
  - We focus on cost-based optimizers

# Three Approaches to Search Space Enumeration

Complete plans

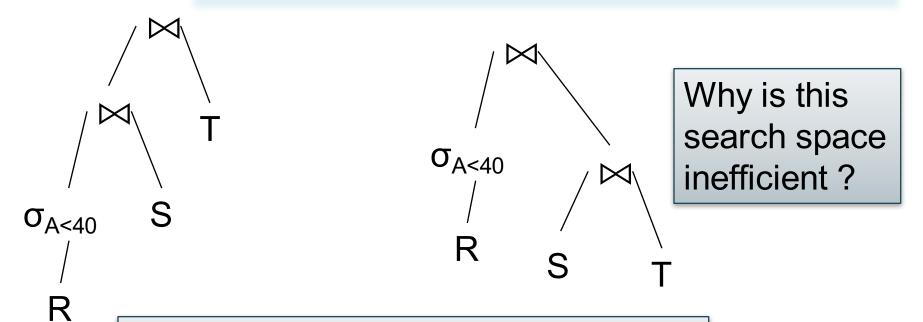
Bottom-up plans

Top-down plans

### Complete Plans

R(A,B)
S(B,C)
T(C,D)

SELECT \*
FROM R, S, T
WHERE R.B=S.B and S.C=T.C and R.A<40



Answer: No way to do early pruning

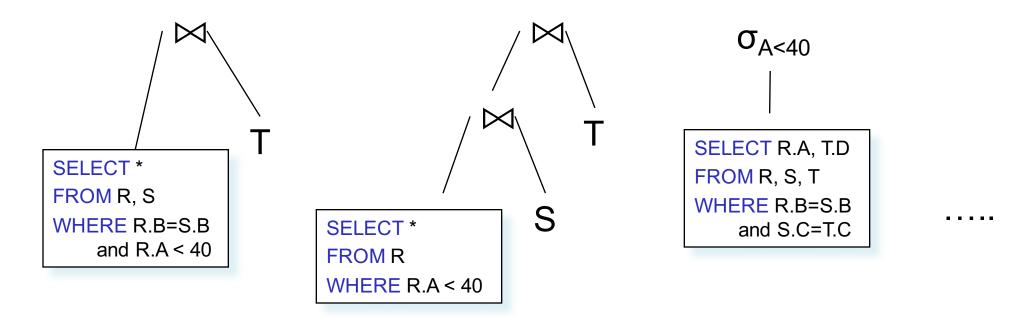
### **Bottom-up Partial Plans**

R(A,B)**SELECT\*** S(B,C)FROM R, S, T T(C,D)WHERE R.B=S.B and S.C=T.C and R.A<40 Why is this better?  $\sigma_{A<40}$  $\sigma_{A<40}$  $\sigma_{A<40}$ 32 We will prune bad plans for sub-expressions

### Top-down Partial Plans

R(A,B)S(B,C)T(C,D)

SELECT \*
FROM R, S, T
WHERE R.B=S.B and S.C=T.C and R.A<40



# Two Types of Plan Enumeration Algorithms

- Dynamic programming (in class)
  - Based on System R (aka Selinger) style optimizer[1979]
  - Limited to joins: join reordering algorithm
  - Bottom-up
- Rule-based algorithm (will not discuss)
  - Database of rules (=algebraic laws)
  - Usually: dynamic programming
  - Usually: top-down