

CSE 444: Database Internals

Lecture 10 Query Optimization (part 1)

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Reminders

- HW2 is due tonight
- 5th year master's reading is due tonight

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Know how to compute the cost of a plan

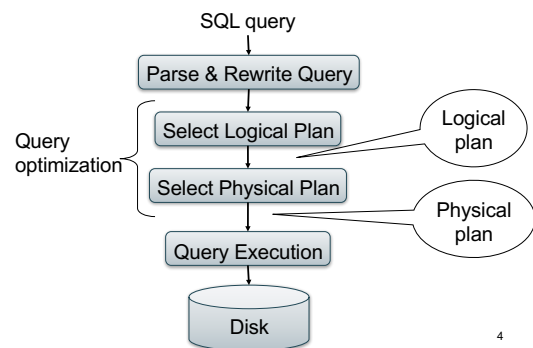
Next: Find a good plan automatically?

This is the role of the query optimizer

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Query Optimization Overview



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What We Already Know...

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

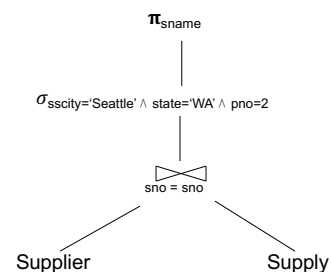
For each SQL query....

```
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...

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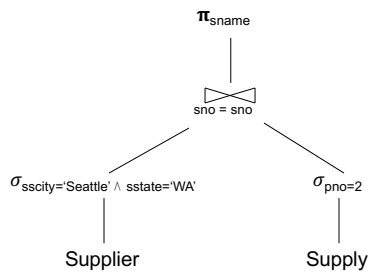
Example Query: Logical Plan 1



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Example Query: Logical Plan 2



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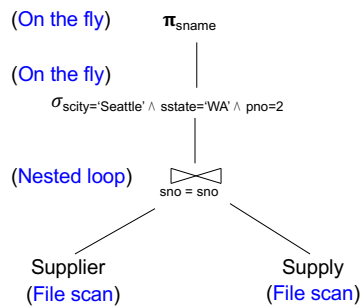
What We Also Know

- For each logical plan...
- There exist many physical plans

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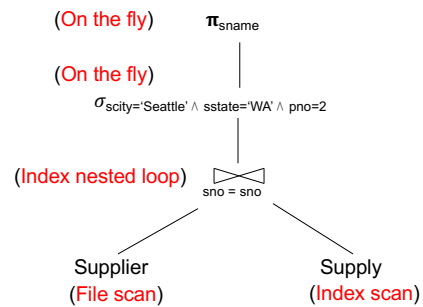
Example Query: Physical Plan 1



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Example Query: Physical Plan 2



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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

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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: histograms over base data
 - In SimpleDB as well... see lab 5.

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Outline

- Search space
- Algorithm for enumerating query plans

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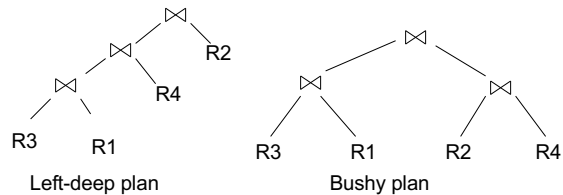
Relational Algebra Equivalences

- Selections
 - Commutative: $\sigma_{c_1}(\sigma_{c_2}(R))$ same as $\sigma_{c_2}(\sigma_{c_1}(R))$
 - Cascading: $\sigma_{c_1 \wedge c_2}(R)$ same as $\sigma_{c_2}(\sigma_{c_1}(R))$
- Projections
 - Cascading
- Joins
 - Commutative: $R \bowtie S$ same as $S \bowtie R$
 - Associative: $R \bowtie (S \bowtie T)$ same as $(R \bowtie S) \bowtie T$

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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

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Commutativity, Associativity, Distributivity

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

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

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

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Laws Involving Selection

$$\sigma_{C \text{ AND } C'}(R) = \sigma_C(\sigma_{C'}(R)) = \sigma_C(R) \cap \sigma_{C'}(R)$$

$$\sigma_{C \text{ OR } C'}(R) = \sigma_C(R) \cup \sigma_{C'}(R)$$

$$\sigma_C(R \bowtie S) = \sigma_C(R) \bowtie S$$

$$\sigma_C(R - S) = \sigma_C(R) - S$$

$$\sigma_C(R \cup S) = \sigma_C(R) \cup \sigma_C(S)$$

$$\sigma_C(R \bowtie S) = \sigma_C(R) \bowtie S$$

Assuming C on attributes of R

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Example: Simple Algebraic Laws

- Example: $R(A, B, C, D), S(E, F, G)$

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

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

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Example: Simple Algebraic Laws

- Example: $R(A, B, C, D), S(E, F, G)$
 $\sigma_{F=3}(R \bowtie_{D=E} S) = R \bowtie_{D=E} \sigma_{F=3}(S)$
 $\sigma_{A=5 \text{ AND } G=9}(R \bowtie_{D=E} S) = \sigma_{A=5}(R) \bowtie_{D=E} \sigma_{G=9}(S)$

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

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

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Laws involving grouping and aggregation

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

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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

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Laws Involving Constraints

Product(pid, pname, price, cid)
 Company(cid, cname, city, state)

Foreign key

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

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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

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Outline

- Search space
- **Algorithm for enumerating query plans**

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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*

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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*

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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

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Three Approaches to Search Space Enumeration

- Complete plans
- Bottom-up plans
- Top-down plans

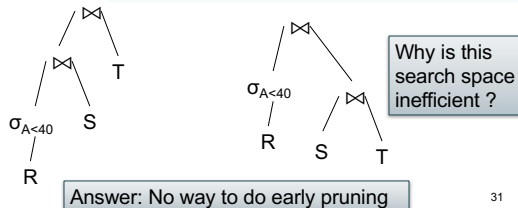
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Complete Plans

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
```



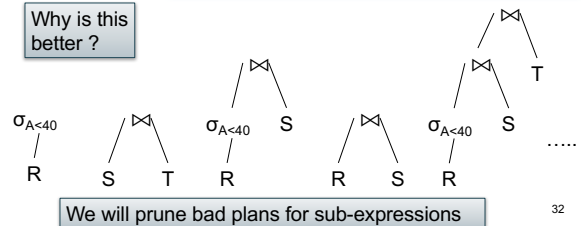
Answer: No way to do early pruning

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Bottom-up 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
```

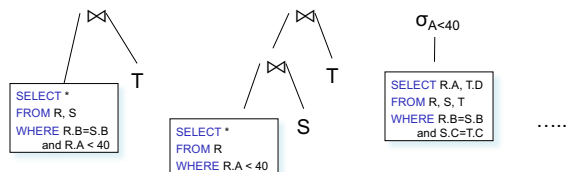


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Top-down Partial Plans

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
```



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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

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