QUERY OPTIMIZATION

CS 564- Fall 2016

EXAMPLE QUERY

- EMP(<u>ssn</u>, ename, addr, sal, did)
 - 10000 tuples, 1000 pages
- DEPT(<u>did</u>, dname, floor, mgr)
 - 500 tuples, 50 pages

```
SELECT DISTINCT ename
```

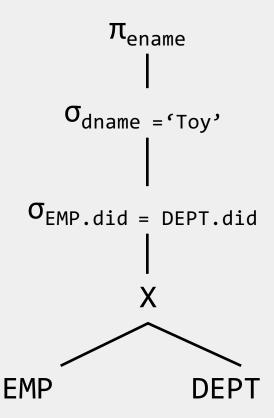
```
FROM Emp E, Dept D
```

WHERE E.did = D.did

AND D.dname = 'Toy';

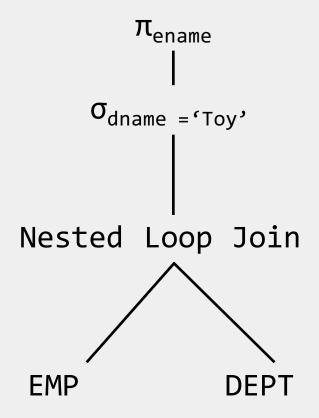
EVALUATION PLAN (1)

```
FROM Emp E, Dept D
WHERE E.did = D.did
AND D.dname = 'Toy';
```



EVALUATION PLAN (2)

```
FROM Emp E, Dept D
WHERE E.did = D.did
AND D.dname = 'Toy';
```



EVALUATION PLAN (3)

```
\pi_{\text{ename}}
SELECT DISTINCT ename
         Emp E, Dept D
FROM
                                              \sigma_{\text{dname}} = \tau_{\text{Toy}}
         E.did = D.did
WHERE
         D.dname = 'Toy';
AND
                                       Sort Merge Join
                            buffer size B= 50
                                        EMP
                                                          DEPT
```

EVALUATION PLAN (4)

```
\pi_{\text{ename}}
SELECT DISTINCT ename
                                                                    buffer size B= 50
         Emp E, Dept D
FROM
WHERE
         E.did = D.did
                                       Sort Merge Join
         D.dname = 'Toy';
AND
                                     \sigma_{\text{dname}} = \tau_{\text{Toy}}
                   index on dname
                                          DEPT
                                                                EMP
```

PIPELINED EVALUATION

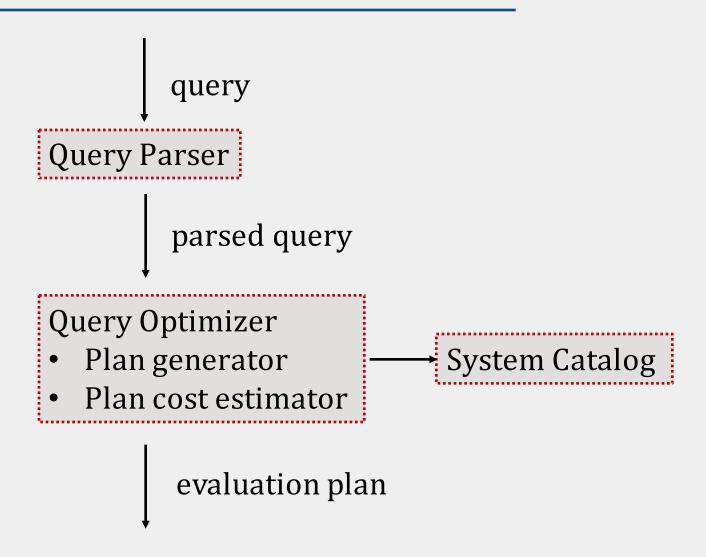
- Instead of materializing the temporary relation to disk, we can instead pipeline to the next operator in memory
- By using pipelining we benefit from:
 - no reading/writing to disk of the temporary relation
 - overlapping execution of operators
- Pipelining is not always possible!

QUERY OPTIMIZATION

The query optimizer

- 1. identifies candidate equivalent trees
- 2. for each tree it finds the best annotated version (using any available indexes): this is called a plan
- 3. chooses the best overall plan by estimating the cost of each plan

ARCHITECTURE OF AN OPTIMIZER



QUERY OPTIMIZATION

- query plan: annotated Relational Algebra tree
 - iteratorinterface: open() /getNext() /close()
 - can be pipelined or materialized
- The optimizer must solve two main issues:
 - What is the space of possible query plans?
 - How can we estimate the cost of each plan?
- Ideally: best plan!
- Practically: avoid worst plans + look at a subset of all plans

COST ESTIMATION

Estimating the cost of a query plan involves:

- estimating the cost of each operation in the plan
 - depends on input cardinalities
 - algorithm cost (we know this!)
- estimating the size of intermediate results
 - we need statistics about input relations
 - for selections and joins, we typically assume independence of predicates

COST ESTIMATION

- Statistics are stored in the system catalog:
 - number of tuples (cardinality)
 - size in pages
 - # distinct keys (when there is an index on the attribute)
 - range (for numeric values)
- The system catalog is updated periodically
- Commercial systems use additional statistics, which provide more accurate estimates:
 - histograms
 - wavelets

EVALUATION PLANS

- The space of possible query plans is typically huge and it is hard to navigate through
- The RA formalism provides us with mathematical rules that transform one RA expression to an equivalent one: for example
 - push selections down
 - reorder joins
- This way we can construct many equivalent alternative query plans

RA EQUIVALENCE (1)

• Commutativity of σ

$$\sigma_{P_1} (\sigma_{P_2}(R)) \equiv \sigma_{P_2}(\sigma_{P_1}(R))$$

• Cascading of σ

$$\sigma_{P_1 \wedge P_2 \wedge \cdots \wedge P_n}(R) \equiv \sigma_{P_1}(\sigma_{P_2}(\dots \sigma_{P_n}(R)))$$

• Cascading of π

$$\pi_{\alpha_1}(R) \equiv \pi_{\alpha_1}(\pi_{\alpha_2}(...\pi_{\alpha_n}(R)...))$$
 when $a_i \subseteq a_{i+1}$

We can evaluate selections in any order!

RA EQUIVALENCE (2)

Commutativity of join

$$R \bowtie S \equiv S \bowtie R$$

Associativity of join

$$(R \bowtie S) \bowtie T \equiv R \bowtie (S \bowtie T)$$

 We can reorder the computation of joins in any way (exponentially many orders)!

RA Equivalence (3)

• Selections + Projections

 $\sigma_{\rm P} \left(\pi_a(R) \right) \equiv \pi_a(\sigma_{\rm P}(R))$ (if the selection involves attributes that remain after projection)

• Selections + Joins

 $\sigma_{\rm P}(R\bowtie S)\equiv\sigma_{\rm P}(R)\bowtie S$ (if the selection involves attributes only in R)

We can push selections down the plan tree!

EVALUATION PLANS

Single relation plan (no joins):

- file scan
- index scan(s): clustered or non-clustered
 - more than one index may "match" predicates
- The optimizer chooses one with the least estimated cost
- We can also merge or pipeline selection and projection (and aggregate when there is no group by)

EVALUATION PLANS

Multiple relation plan

- joins can be evaluated in any order
- selections can be combined into the join operator
- selections and projections can be pushed down the plan tree using the RA equivalence transformations

Join Reordering

Consider the following join: $R \bowtie S \bowtie T \bowtie U$

Most DBMSs consider left-deep join plans

These allow for fully pipelined evaluation

