

CSE 444: Database Internals

Lecture 11 Query Optimization (part 2)

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1

Reminders

- Homework 2 due tonight by 11pm in drop box
- Lab 2 is due on Friday by 11pm
- HW 5 has been released and is due next week

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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

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3

The Three Parts of an Optimizer

- Cost estimation
 - Based on cardinality estimation
- Search space
 - Algebraic laws, restricted types of join trees
- Search algorithm
 - Will discuss next

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4

Search Algorithm

- 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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5

Dynamic Programming

Originally proposed in System R [1979]

- Only handles single block queries:

```
SELECT list
FROM R1, ..., Rn
WHERE cond1 AND cond2 AND ... AND condk
```

- Some heuristics for search space enumeration:
 - Selections down
 - Projections up
 - Avoid cartesian products

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6

Dynamic Programming

```
SELECT list
FROM R1, ..., Rn
WHERE cond1 AND cond2 AND ... AND cond
```

- For each subquery $Q \subseteq \{R_1, \dots, R_n\}$ compute the following:
 - $T(Q)$ = the estimated size of Q
 - $\text{Plan}(Q)$ = a best plan for Q
 - $\text{Cost}(Q)$ = the estimated cost of that plan

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7

Dynamic Programming

```
SELECT list
FROM R1, ..., Rn
WHERE cond1 AND cond2 AND ... AND cond
```

- Step 1:** For each $\{R_i\}$ do:
 - $T(\{R_i\}) = T(R_i)$
 - $\text{Plan}(\{R_i\})$ = access method for R_i
 - $\text{Cost}(\{R_i\})$ = cost of access method for R_i

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8

Dynamic Programming

```
SELECT list
FROM R1, ..., Rn
WHERE cond1 AND cond2 AND ... AND cond
```

- Step 2:** For each $Q \subseteq \{R_1, \dots, R_n\}$ of size k do:
 - $T(Q)$ = use estimator
 - Consider all partitions $Q = Q' \cup Q''$
compute $\text{cost}(\text{Plan}(Q') \bowtie \text{Plan}(Q''))$
 - $\text{Cost}(Q)$ = the smallest such cost
 - $\text{Plan}(Q)$ = the corresponding plan
- Note**
 - If we restrict to left-linear trees: Q'' = single relation
 - May want to avoid cartesian products

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9

Dynamic Programming

```
SELECT list
FROM R1, ..., Rn
WHERE cond1 AND cond2 AND ... AND cond
```

- Step 3:** Return $\text{Plan}(\{R_1, \dots, R_n\})$

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10

Example

```
SELECT *
FROM R, S, T, U
WHERE cond1 AND cond2 AND ...
```

- $R \bowtie S \bowtie T \bowtie U$
- Assumptions:

```
T(R) = 2000
T(S) = 5000
T(T) = 3000
T(U) = 1000
```

- Every join selectivity is 0.001

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11

```
T(R) = 2000
T(S) = 5000
T(T) = 3000
T(U) = 1000
```

Assume
 $B(\cdot) = T(\cdot)/10$
Join selectivity
is 0.001

Subquery	T	Plan	Cost
R	2000		
S	5000		
T	3000		
U	1000		
RS			
RT			
RU			
ST			
SU			
TU			
RST			
RSU			
RTU			
STU			
RSTU			

12

T(R) = 2000
T(S) = 5000
T(T) = 3000
T(U) = 1000

Assume
 $B(..) = T(..)/10$

Join selectivity
is 0.001

Subquery	T	Plan	Cost
R	2000		
S	5000		
T	3000		
U	1000		
RS	10000		
RT	6000		
RU	2000		
ST	15000		
SU	5000		
TU	3000		
RST	30000		
RSU	10000		
RTU	6000		
STU	15000		
RSTU	30000		

13

T(R) = 2000
T(S) = 5000
T(T) = 3000
T(U) = 1000

Assume
 $B(..) = T(..)/10$

Join selectivity
is 0.001

Subquery	T	Plan	Cost
R	2000	Clustered index scan R.A	200
S	5000		
T	3000		
U	1000		
RS	10000		
RT	6000		
RU	2000		
ST	15000		
SU	5000		
TU	3000		
RST	30000		
RSU	10000		
RTU	6000		
STU	15000		
RSTU	30000		

14

T(R) = 2000
T(S) = 5000
T(T) = 3000
T(U) = 1000

Assume
 $B(..) = T(..)/10$

Join selectivity
is 0.001

Subquery	T	Plan	Cost
R	2000	Clustered index scan R.A	200
S	5000	Table scan	500
T	3000		
U	1000		
RS	10000		
RT	6000		
RU	2000		
ST	15000		
SU	5000		
TU	3000		
RST	30000		
RSU	10000		
RTU	6000		
STU	15000		
RSTU	30000		

15

T(R) = 2000
T(S) = 5000
T(T) = 3000
T(U) = 1000

Assume
 $B(..) = T(..)/10$

Join selectivity
is 0.001

Subquery	T	Plan	Cost
R	2000	Clustered index scan R.A	200
S	5000	Table scan	500
T	3000	Table scan	300
U	1000	Unclustered index scan U.F	1000
RS	10000		
RT	6000		
RU	2000		
ST	15000		
SU	5000		
TU	3000		
RST	30000		
RSU	10000		
RTU	6000		
STU	15000		
RSTU	30000		

16

T(R) = 2000
T(S) = 5000
T(T) = 3000
T(U) = 1000

Assume
 $B(..) = T(..)/10$

Join selectivity
is 0.001

Subquery	T	Plan	Cost
R	2000	Clustered index scan R.A	200
S	5000	Table scan	500
T	3000	Table scan	300
U	1000	Unclustered index scan U.F	1000
RS	10000	R ~ S nested loop join	...
RT	6000		
RU	2000		
ST	15000		
SU	5000		
TU	3000		
RST	30000		
RSU	10000		
RTU	6000		
STU	15000		
RSTU	30000		

17

T(R) = 2000
T(S) = 5000
T(T) = 3000
T(U) = 1000

Assume
 $B(..) = T(..)/10$

Join selectivity
is 0.001

Subquery	T	Plan	Cost
R	2000	Clustered index scan R.A	200
S	5000	Table scan	500
T	3000	Table scan	300
U	1000	Unclustered index scan U.F	1000
RS	10000	R ~ S nested loop join	...
RT	6000	R ~ T index join	...
RU	2000		
ST	15000		
SU	5000		
TU	3000		
RST	30000		
RSU	10000		
RTU	6000		
STU	15000		
RSTU	30000		

18

T(R) = 2000
T(S) = 5000
T(T) = 3000
T(U) = 1000

Assume
 $B(.) = T(.) / 10$

Join selectivity
is 0.001

Subquery	T	Plan	Cost
R	2000	Clustered index scan R.A	200
S	5000	Table scan	500
T	3000	Table scan	300
U	1000	Unclustered index scan U.F	1000
RS	10000	R $\bowtie$ S nested loop join	...
RT	6000	R $\bowtie$ T index join	...
RU	2000	R $\bowtie$ U index join	...
ST	15000	S $\bowtie$ T hash join	...
SU	5000	...	...
TU	3000	...	...
RST	30000	...	...
RSU	10000	...	...
RTU	6000	...	...
STU	15000	...	...
RSTU	30000	...	...

19

T(R) = 2000
T(S) = 5000
T(T) = 3000
T(U) = 1000

Assume
 $B(.) = T(.) / 10$

Join selectivity
is 0.001

Subquery	T	Plan	Cost
R	2000	Clustered index scan R.A	200
S	5000	Table scan	500
T	3000	Table scan	300
U	1000	Unclustered index scan U.F	1000
RS	10000	R $\bowtie$ S nested loop join	...
RT	6000	R $\bowtie$ T index join	...
RU	2000	R $\bowtie$ U index join	...
ST	15000	S $\bowtie$ T hash join	...
SU	5000	...	...
TU	3000	...	...
RST	30000	(RT) $\bowtie$ S hash join	...
RSU	10000	(SU) $\bowtie$ R merge join	...
RTU	6000	...	...
STU	15000	...	...
RSTU	30000	...	...

20

T(R) = 2000
T(S) = 5000
T(T) = 3000
T(U) = 1000

Assume
 $B(.) = T(.) / 10$

Join selectivity
is 0.001

Subquery	T	Plan	Cost
R	2000	Clustered index scan R.A	200
S	5000	Table scan	500
T	3000	Table scan	300
U	1000	Unclustered index scan U.F	1000
RS	10000	R $\bowtie$ S nested loop join	...
RT	6000	R $\bowtie$ T index join	...
RU	2000	R $\bowtie$ U index join	...
ST	15000	S $\bowtie$ T hash join	...
SU	5000	...	...
TU	3000	...	...
RST	30000	(RT) $\bowtie$ S hash join	...
RSU	10000	(SU) $\bowtie$ R merge join	...
RTU	6000	...	...
STU	15000	...	...
RSTU	30000	(RT) $\bowtie$ (SU) hash join	...

21

Discussion

- Need to consider both $R \bowtie S$ and $S \bowtie R$
 - Because the cost may be different!
- When computing the cheapest plan for $(Q) \bowtie R$, we may consider new access methods for R, e.g. an index look-up that makes sense only in the context of the join

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22

Discussion

Given a query with n relations R1, ..., Rn

- How many entries do we have in the dynamic programming table?
- For each entry, how many alternative plans do we need to inspect?

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23

Discussion

Given a query with n relations R1, ..., Rn

- How many entries do we have in the dynamic programming table?
 - A: $2^n - 1$
- For each entry, how many alternative plans do we need to inspect?
 - A: for each entry with k tables, examine $2^k - 2$ plans

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24

Reducing the Search Space

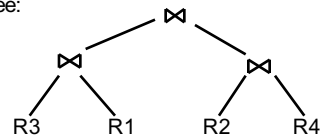
- Left-linear trees
- No cartesian products

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25

Join Trees

- $R1 \bowtie R2 \bowtie \dots \bowtie Rn$
- Join tree:



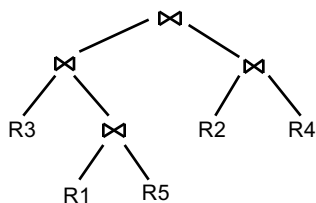
- A plan = a join tree
- A partial plan = a subtree of a join tree

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26

Types of Join Trees

- Bushy:

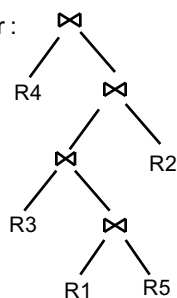


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27

Types of Join Trees

- Linear:

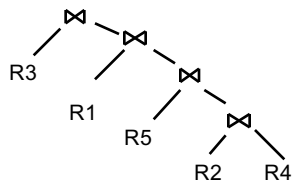


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28

Types of Join Trees

- Right deep:

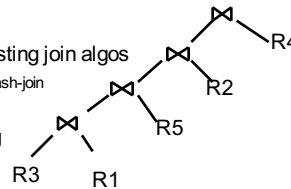


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29

Types of Join Trees

- Left deep:
 - Work well with existing join algos
 - Nested-loop and hash-join
 - Facilitate pipelining
 - Dynamic programming can be used with all trees



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30

Selinger Algorithm

Selinger enumeration algorithm considers

- Different logical and physical plans *at the same time*
- Cost of a plan is IO + CPU
- Concept of *interesting order* during plan enumeration
 - Same order as that requested by ORDER BY or GROUP BY
 - Attributes that appear in eq-join predicates

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35

More about the Selinger Algorithm

- Step 1: Enumerate all access paths for a single relation
 - File scan or index scan
 - Keep the cheapest for each *interesting order*
- Step 2: Consider all ways to join two relations
 - Use result from step 1 as the outer relation
 - Consider every other possible relation as inner relation
 - Estimate cost when using sort-merge or nested-loop join
 - Keep the cheapest for each *interesting order*
- Steps 3 and later: Repeat for three relations, etc.

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36

Example on the Board

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37