Query processing and optimization

## Definitions

- Query processing
  - translation of query into low-level activities
  - evaluation of query
  - data extraction
- Query optimization
  - selecting the most efficient query evaluation

# Query Processing (1/2)

- SELECT \* FROM student WHERE name=Paul
- Parse query and translate
  - check syntax, verify names, etc
  - translate into relational algebra (RDBMS)
  - create evaluation plans
- Find best plan (optimization)
- Execute plan

student		tak		
<u>cid</u>	name	<u>cid</u>	<u>courseid</u>	courseic
00112233	Paul	00112233	312	312
00112238	Rob	00112233	395	395
00112235	Matt	00112235	312	

course

coursename

Advanced DBs

Machine Learning

#### Query Processing (2/2)



## Relational Algebra (1/2)

- Query language
- Operations:
  - select:  $\sigma$
  - project:  $\pi$
  - union:  $\cup$
  - difference: -
  - product: x
  - join: ⋈

# Relational Algebra (2/2)

- SELECT \* FROM student WHERE name=Paul
  - $\sigma_{name=Paul}(student)$
- $\pi_{name}(\sigma_{cid<00112235}(student))$
- $\pi_{name}(\sigma_{coursename=Advanced DBs}((student \Join_{cid} takes) \bowtie_{courseid} course))$

student		takes		course		
<u>cid</u>	name	<u>cid</u>	<u>courseid</u>	<u>courseid</u>	<u>coursename</u>	
00112233	Paul	00112233	312	312	Advanced DBs	
00112238	Rob	00112233	395	395	Machine Learning	
00112235	Matt	00112235	312			

# Why Optimize?

- Many alternative options to evaluate a query
  - $\pi_{name}(\sigma_{coursename=Advanced DBs}((student data takes) data course))$
  - $\pi_{name}$  ((student  $\aleph_{cid}$  takes)  $\aleph_{courseid} \sigma_{coursename=Advanced DBs}$  (course)))
- Several options to evaluate a single operation
  - $\sigma_{name=Paul}(student)$ 
    - scan file
    - use secondary index on student.name
- Multiple access paths
  - access path: how can records be accessed

#### **Evaluation plans**

- Specify which access path to follow
- Specify which algorithm to use to evaluate operator
- Specify how operators interleave
- Optimization:
  - estimate the cost of each plan (not all plans)
  - select plan with lowest estimated cost



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Query processing and optimization

name

 $\sigma_{coursename=Advanced DBs}$ 

# **Estimating Cost**

- What needs to be considered:
  - Disk I/Os
    - sequential
    - random
  - CPU time
  - Network communication
- What are we going to consider:
  - Disk I/Os
    - page reads/writes
  - Ignoring cost of writing final output

## **Operations and Costs**

## Operations and Costs (1/2)

- Operations:  $\sigma$ ,  $\pi$ ,  $\cup$ ,  $\cap$ , -, x,  $\bowtie$
- Costs:
  - N<sub>R</sub>: number of records in R
  - L<sub>R</sub>: size of record in R
  - F<sub>R</sub>: blocking factor
    - number of records in page
  - B<sub>R</sub>: number of pages to store relation R
  - V(A,R): number of distinct values of attribute A in R
  - SC(A,R): selection cardinality of A in R
    - A key: S(A,R)=1
    - A nonkey:  $S(A,R) = N_R / V(A,R)$
  - HT<sub>i</sub>: number of levels in index I
  - rounding up fractions and logarithms

### Operations and Costs (2/2)

- relation takes
  - 700 tuples
  - student cid 8 bytes
  - course id 4 bytes
  - 9 courses
  - 100 students
  - page size 512 bytes
  - output size (in pages) of query: which students take the Advanced DBs course?
    - N<sub>takes</sub> = 700
    - V(courseid, takes) = 9
    - SC(courseid,takes) = ceil( N<sub>takes</sub>/V(courseid, takes) ) = ceil(700/9) = 78
    - f = floor(512/8) = 64
    - B = ceil(78/64) = 2 pages

# Selection $\sigma$ (1/2)

- Linear search
  - read all pages, find records that match (assuming equality search)
  - average cost:
    - nonkey B<sub>R</sub>, key 0.5\*B<sub>R</sub>
- Binary search
  - on ordered field
  - average cost:  $|\log_2 B_R| + m$ 
    - *m* additional pages to be read
    - *m* = ceil( SC(A,R)/F<sub>R</sub> ) 1
- Primary/Clustered Index
  - average cost:
    - single record HT<sub>i</sub> + 1
    - multiple records HT<sub>i</sub> + ceil( SC(A,R)/F<sub>R</sub> )

## Selection $\sigma$ (2/2)

- Secondary Index
  - average cost:
    - key field  $HT_i + 1$
    - nonkey field
      - worst case  $HT_i + SC(A,R)$
      - linear search more desirable if many matching records

## Complex selection $\sigma_{expr}$

- conjunctive selections:  $\sigma_{\theta_1 \land \theta_2 \dots \land \theta_n}$ 
  - perform simple selection using  $\theta_i$  with the lowest evaluation cost
    - e.g. using an index corresponding to  $\theta_i$
    - apply remaining conditions  $\theta$  on the resulting records
    - $\sigma_{cid>00112233\land courseid=312}(takes)$
    - cost: the cost of the simple selection on selected  $\boldsymbol{\theta}$
  - multiple indices
    - select indices that correspond to  $\theta_i$ s
    - scan indices and return RIDs
    - answer: intersection of RIDs
    - cost: the sum of costs + record retrieval
- disjunctive selections:  $\sigma_{\theta_1 \lor \theta_2 \ldots \lor \theta_n}$ 
  - multiple indices
    - union of RIDs
  - linear search

### Projection and set operations

- SELECT DISTINCT cid FROM takes
  - $-\pi$  requires duplicate elimination
  - sorting
- set operations require duplicate elimination
  - $R \cap S$
  - $R \cup S$
  - sorting

# Sorting

- efficient evaluation for many operations
- required by query:
  - SELECT cid,name FROM student ORDER BY name
- implementations
  - internal sorting (if records fit in memory)
  - external sorting

## External Sort-Merge Algorithm (1/3)

• Sort stage: create sorted runs

```
i=0;
repeat
  read M pages of relation R into memory
  sort the M pages
  write them into file R<sub>i</sub>
  increment i
until no more pages
  N = i  // number of runs
```

## External Sort-Merge Algorithm (2/3)

• Merge stage: merge sorted *runs* 

```
//assuming N < M
allocate a page for each run file R<sub>i</sub> // N pages allocated
read a page P<sub>i</sub> of each R<sub>i</sub>
repeat
choose first record (in sort order) among N pages, say from page P<sub>j</sub>
write record to output and delete from page P<sub>j</sub>
if page is empty read next page P<sub>j</sub>' from R<sub>j</sub>
until all pages are empty
```

## External Sort-Merge Algorithm (3/3)

- Merge stage: merge sorted *runs*
- What if N > M ?
  - perform multiple passes
  - each *pass* merges M-1 runs until relation is processed
  - in next pass number of runs is reduced
  - final *pass* generated sorted output

#### Sort-Merge Example

12

12

73

11

21

95 d 12 а x |44 95 S f 12 73 0 t 45 file 67 n 87 е 11 Ζ 22 V 38 b



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## Sort-Merge cost

- B<sub>R</sub> the number of pages of R
- Sort stage: 2 \* B<sub>R</sub>
  - read/write relation
- Merge stage:
  - initially  $\left|\frac{B_R}{M}\right|$  runs to be merged
  - each pass M-1 runs sorted
  - thus, total number of passes:  $\log$

$$g_{M-1}\left(\frac{B_R}{M}\right)$$

- at each pass 2 \* B<sub>R</sub> pages are read
  - read/write relation
  - apart from final write
- Total cost:

$$- 2 * B_{R} + 2 * B_{R} * \left| \log_{M-1} \left( \frac{B_{R}}{M} \right) \right| - B_{R}$$

## Projection

- π<sub>A1,A2...</sub> (R)
- remove unwanted attributes
  - scan and drop attributes
- remove duplicate records
  - sort resulting records using all attributes as sort order
  - scan sorted result, eliminate duplicates (adjucent)
- cost
  - initial scan + sorting + final scan

#### Join

- $\pi_{name}(\sigma_{coursename=Advanced DBs}((student \overset{\bowtie}{_{cid}} takes) \overset{\bowtie}{_{courseid}} course))$
- implementations
  - nested loop join
  - block-nested loop join
  - indexed nested loop join
  - sort-merge join
  - hash join

# Nested loop join (1/2)

• R ⋈ S

```
for each tuple t_R of R
for each t_S of S
if (t_R t_S match) output t_R t_S
end
end
```

- Works for any join condition
- S inner relation
- R outer relation

# Nested loop join (2/2)

- Costs:
  - best case when smaller relation fits in memory
    - use it as inner relation
    - B<sub>R</sub>+B<sub>S</sub>
  - worst case when memory holds one page of each relation
    - S scanned for each tuple in R
    - N<sub>R</sub> \* B<sub>s</sub> + B<sub>R</sub>

#### Block nested loop join (1/2)

for each page  $X_R$  of R for each page  $X_S$  of S for each tuple  $t_R$  in  $X_R$ for each  $t_S$  in  $X_S$ if ( $t_R t_S$  match) output  $t_R t_S$ end end end end

## Block nested loop join (2/2)

- Costs:
  - best case when smaller relation fits in memory
    - use it as inner relation
    - B<sub>R</sub>+B<sub>S</sub>
  - worst case when memory holds one page of each relation
    - S scanned for each page in R
    - B<sub>R</sub> \* B<sub>s</sub> + B<sub>R</sub>

### Indexed nested loop join

- R⊠S
- Index on inner relation (S)
- for each tuple in outer relation (R) *probe* index of inner relation
- Costs:
  - B<sub>R</sub> + N<sub>R</sub> \* c
    - c the cost of index-based selection of inner relation
  - relation with fewer records as outer relation

## Sort-merge join

- R⊠S
- Relations sorted on the join attribute
- Merge sorted relations
  - pointers to first record in each relation
  - read in a group of records of S with the same values in the join attribute
  - read records of R and process
- Relations in sorted order to be read once
- Cost:
  - cost of sorting +  $B_S$  +  $B_R$



# Hash join

- R ⋈ S
- use  $h_1$  on joining attribute to map records to partitions that fit in memory
  - records of R are partitioned into  $R_0 \dots R_{n-1}$
  - records of S are partitioned into  $S_0...S_{n-1}$
- join records in corresponding partitions
  - using a hash-based indexed block nested loop join
- Cost:  $2^*(B_R + B_S) + (B_R + B_S)$



#### Exercise: joins

- R⊠S
- N<sub>R</sub>=2<sup>15</sup>
- B<sub>R</sub> = 100
- N<sub>S</sub>=2<sup>6</sup>
- B<sub>S</sub> = 30
- $B^+$  index on S
  - order 4
  - full nodes
- nested loop join: best case worst case
- block nested loop join: best case worst case
- indexed nested loop join

## Evaluation

- evaluate multiple operations in a plan
- materialization
- pipelining



## Materialization

- create and read temporary relations
- create implies writing to disk
  - more page writes



# Pipelining (1/2)

- creating a pipeline of operations
- reduces number of read-write operations
- implementations
  - demand-driven data pull
  - producer-driven data push



# Pipelining (2/2)

- can pipelining always be used?
- any algorithm?
- cost of  $R \bowtie S$ 
  - materialization and hash join:  $B_R + 3(B_R + B_S)$
  - pipelining and indexed nested loop join:  $N_R * HT_i$



# **Query Optimization**

## Choosing evaluation plans

- cost based optimization
- enumeration of plans
  - $R \bowtie S \bowtie T$ , 12 possible orders
- cost estimation of each plan
- overall cost
  - cannot optimize operation independently

## Cost estimation

- operation (σ, π, ⋈ ...)
- implementation
- size of inputs
- size of outputs
- sorting



## Size Estimation (1/2)

- $\sigma_{A=v}(R)$ - SC(A,R)
- $\sigma_{A \leq v}(R)$  $- N_R * \frac{v - \min(A, R)}{\max(A, R) - \min(A, R)}$ •  $\sigma_{\theta_1 \land \theta_2 \land \dots \land \theta_n}(R)$ 
  - - multiplying probabilities
    - $N_R * [(s_1/N_R) * (s_2/N_R) * ... (s_n/N_R)]$
- $\sigma_{\theta_1 \vee \theta_2 \vee \ldots \vee \theta_n}(R)$ 
  - probability that a record satisfy none of  $\theta$ :  $[(1-s_1/N_p)*(1-s_2/N_p)*...*(1-s_p/N_p)]$
  - $N_{R} * (1 [(1 s_{1}/N_{R}) * (1 s_{2}/N_{R}) * ... * (1 s_{n}/N_{R})])$

## Size Estimation (2/2)

- R x S
  - $-N_R * N_S$
- R⋈S
  - $R \cap S = \emptyset$ :  $N_R^* N_S$
  - $\,R \cap S$  key for R: maximum output size is  $N_s$
  - $\,R \cap S$  foreign key for R:  $N_S$
  - $R \cap S = \{A\}$ , neither key of R nor S
    - $N_R^*N_S / V(A,S)$
    - N<sub>S</sub>\*N<sub>R</sub> / V(A,R)

## Expression Equivalence

• conjunctive selection decomposition

 $- \sigma_{\theta_1 \land \theta_2}(R) = \sigma_{\theta_1}(\sigma_{\theta_2}(R))$ 

commutativity of selection

 $- \sigma_{\theta_1}(\sigma_{\theta_2}(R)) = \sigma_{\theta_2}(\sigma_{\theta_1}(R))$ 

• combining selection with join and product

 $- \sigma_{\theta_1}(\mathsf{R} \times \mathsf{S}) = \mathsf{R} \Join_{\theta_1} \mathsf{S}$ 

- commutativity of joins
  - $\mathbb{R} \bowtie_{\theta_1} \mathbb{S} = \mathbb{S} \bowtie_{\theta_1} \mathbb{R}$
- distribution of selection over join
  - $\sigma_{\theta_1 \land \theta_2}(\mathsf{R} \bowtie \mathsf{S}) = \sigma_{\theta_1}(\mathsf{R}) \bowtie \sigma_{\theta_2}(\mathsf{S})$
- distribution of projection over join

−  $π_{A1,A2}(R \bowtie S) = π_{A1}(R) \bowtie π_{A2}(S)$ 

• associativity of joins:  $R \bowtie (S \bowtie T) = (R \bowtie S) \bowtie T$ 

## Cost Optimizer (1/2)

- transforms expressions
  - equivalent expressions
  - heuristics, *rules of thumb* 
    - perform selections early
    - perform projections early
    - replace products followed by selection  $\sigma$  (R x S) with joins R $\bowtie$ S
    - start with joins, selections with smallest result
      - create left-deep join trees



### Cost Optimizer (2/2)



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#### **Cost Evaluation Exercise**

- $\pi_{name}(\sigma_{coursename=Advanced DBs}((student \bowtie_{cid} takes) \bowtie_{courseid} course))$
- R = student  $\bowtie_{cid}$  takes
- S = course
- $N_s = 10$  records
- assume that on average there are 50 students taking each course
- blocking factor: 2 records/page
- what is the cost of  $\sigma_{\text{coursename}=\text{Advanced DBs}}$  (R  $\bowtie_{\text{courseid}}$  S)
- what is the cost of  $R \bowtie \sigma_{\text{coursename}=\text{Advanced DBs}} S$
- assume relations can fit in memory

## Summary

- Estimating the cost of a single operation
- Estimating the cost of a query plan
- Optimization
  - choose the most efficient plan