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)

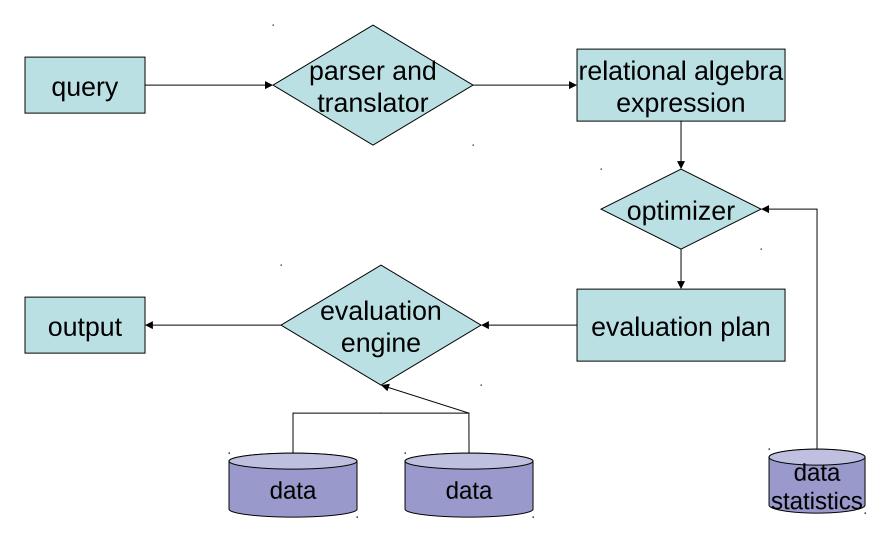
- SELECT * FROM student WHERE name=Paul
- Parse query and translate
 - check syntax, verify names, etc
 - translate into relational algebra (RDBMS)
 - create evaluation plans
- Find best plan (optimization)
- Execute plan

student	
<u>cid</u>	name
00112233	Paul
00112238	Rob
00112235	Matt

takes	
<u>cid</u>	courseid
00112233	312
00112233	395
00112235	312

course	
courseid	<u>coursename</u>
312	Advanced DBs
395	Machine Learning

Query Processing (2/2)



Relational Algebra (1/2)

- Query language
- Operations:
 - select: σ
 - project: π
 - union: ∪
 - difference: -
 - product: x
 - join: ⋈
 - Extended relational algebra operations:

Relational Algebra (2/2)

- SELECT * FROM student WHERE name=Paul
 - $-\sigma_{name=Paul}(student)$
- $\pi_{\text{name}}(\sigma_{\text{cid}<0.0112235}(\text{student}))$
- $\pi_{\text{name}}(\sigma_{\text{coursename=Advanced DBs}}(\text{(student }_{\text{cid}}\text{ takes}))$

student	
<u>cid</u>	name
00112233	Paul
00112238	Rob
00112235	Matt

takes	
<u>cid</u>	courseid
00112233	312
00112233	395
00112235	312

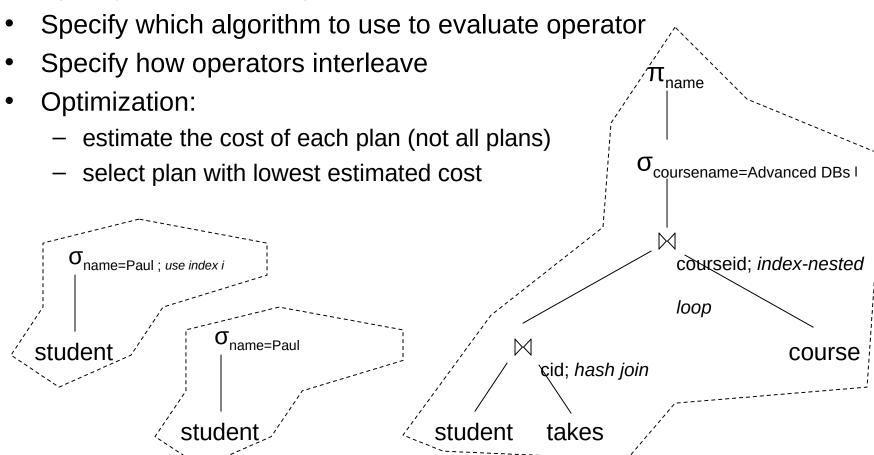
course		
courseid	<u>coursename</u>	
312	Advanced DBs	
395	Machine Learning	

Why Optimize?

- Many alternative options to evaluate a query
 - $-\pi_{\text{name}}(\sigma_{\text{coursename}=\text{Advanced DBs}}(\text{(student }_{\text{cid}} \text{ takes}) \text{ }_{\text{course}})$
 - π_{name} ((student $\overset{\bowtie}{\text{cid}}$ takes) $\overset{\bowtie}{\text{courseid}}$ $\sigma_{\text{coursename=Advanced DBs}}$ (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 (using index; which index?; etc.)

Evaluation plans

Specify which access path to follow



Estimating Cost

- What needs to be considered:
 - Disk I/Os
 - sequential (reading neighbouring pages is faster)
 - random
 - CPU time
 - Network communication
- What are we going to consider:
 - Disk I/Os
 - page (data block) reads/writes
 - Ignoring cost of writing final output

Operations and Costs (1/2)

- Operations: σ , π , \cup , \cap , -, x, \bowtie
- Costs:
 - N_R : number of records in R (other notation: T_R or T(R) -> tuple)
 - L_R: size of record in R (length of record)
 - bf_R : blocking factor (other notation: F_R)
 - number of records in a page (datablock)
 - B_R: number of pages to store relation R
 - V(R, A): number of distinct values of attribute A in R other notation: $I_A(R)$ (Image size)
 - SC(R,A): selection cardinality of A in R (number of matching rec.)
 - A key: SC(R, A)=1
 - A nonkey: $SC(R, A) = T_R / V(R, A)$ (uniform distribution assumption)
 - HT_i: number of levels in index I (-> height of tree)
 - rounding up fractions and logarithms

Operations and Costs (2/2)

relation takes

- 7000 tuples
- student cid 8 bytes
- course id 4 bytes
- 40 courses
- 1000 students
- page size 512 bytes
- output size (in pages) of query:

which students take the Advanced DBs course?

- $T_{takes} = 7000$
- V(courseid, takes) = 40
- SC(courseid, takes)=ceil(T_{takes}/V(courseid, takes))=ceil(7000/40)=175
- $bf_{takes} = floor(512/12) = 42$ $B_{takes} = 7000/42 = 167$ pages
- $bf_{output} = floor(512/8) = 64$ $B_{output} = 175/64 = 3 pages$

Cost of Selection σ (1/2)

- Linear search
 - read all pages, find records that match (assuming equality search)
 - average cost:
 - nonkey (multiple occurences): B_R, key: 0.5*B_R
- Binary search
 - on ordered field
 - average cost: $|\log_2 B_R| + m$
 - *m* additional pages to be read (first found then read the duplicates)
 - $m = \text{ceil}(SC(A,R)/bf_R) 1$
- Primary/Clustered Index (B+ tree)
 - average cost:
 - single record: HT_i + 1
 - multiple records: HT_i + ceil(SC(R,A)/bf_R)

Cost of Selection σ (2/2)

- Secondary Index (B+ tree)
 - 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 σ_{expr}

- conjunctive selections: $\sigma_{{ heta_1}^{\wedge}{ heta_2}...^{\wedge}{ heta_n}}$
 - perform simple selection using θ_i with the lowest evaluation cost
 - e.g. using an index corresponding to θ_i
 - apply remaining conditions θ on the resulting records
 - $\sigma_{cid>00112233^{\circ} courseid=312}(takes)$
 - cost: the cost of the simple selection on selected $\boldsymbol{\theta}$
 - multiple indices
 - select indices that correspond to θ_i s
 - scan indices and return RIDs (ROWID in Oracle)
 - answer: intersection of RIDs
 - cost: the sum of costs + record retrieval
- disjunctive selections: $O_{\theta_1 \circ \theta_2 \dots \circ \theta_n}$
 - multiple indices
 - union of RIDs
 - linear search

Projection and set operations

- SELECT DISTINCT cid FROM takes
 - π 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

(that's why we need temporary space on disk)

External Sort-Merge Algorithm (1/3)

Sort stage: create sorted runs

```
i=0;
repeat
    read M pages of relation R into memory (M: size of 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)

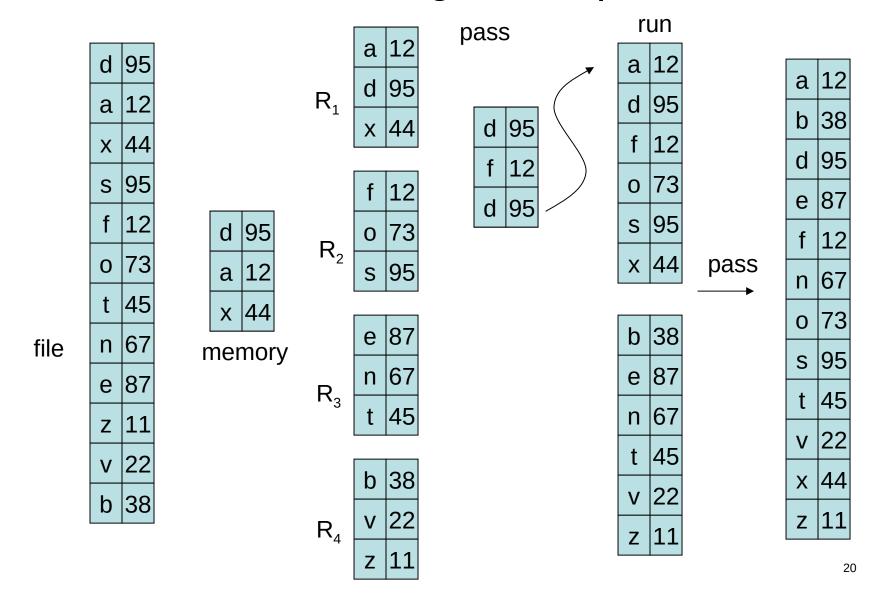
Merge stage: merge sorted runs

```
//assuming N < M (N <= M-1 we need 1 output buffer) allocate a page for each run file R_i // N pages allocated read a page P_i of each R_i repeat choose first record (in sort order) among N pages, say from page P_j write record to output and delete from page P_j if page is empty read next page P_j' from R_j 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



Sort-Merge cost

- B_R the number of pages of R
- Sort stage: 2 * B_R
 - 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: $\left|\log_{M-1}\left(\frac{B_R}{M}\right)\right|$
 - at each pass 2 * B_R pages are read/written
 - read/write relation (B_R + B_R)
 - apart from final write (B_R)

• Total cost:

$$-2 * B_R + 2 * B_R * \left| \log_{M-1} \left(\frac{B_R}{M} \right) \right| - B_R$$
 eg. $B_R = 1000000$, $M=100$

Projection

- $\pi_{A1,A2...}$ (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 (adjacent)
- cost
 - initial scan + sorting + final scan

Join

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

Nested loop join (1/2)

• R⋈S

```
\label{eq: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
 - $B_R + B_S$
 - worst case when memory holds one page of each relation
 - S scanned for each tuple in R
 - T_R * B_s + B_R

Block nested loop join (1/2)

```
\begin{array}{c} \text{for each page $X_{\text{R}}$ of $R$} \\ \text{for each page $X_{\text{S}}$ of $S$} \\ \text{for each tuple $t_{\text{R}}$ in $X_{\text{R}}$} \\ \text{for each $t_{\text{S}}$ in $X_{\text{S}}$} \\ \text{if $(t_{\text{R}}$ $t_{\text{S}}$ match) output $t_{\text{R}}$.$t_{\text{S}}$} \\ \text{end} \\ \text{end} \\ \text{end} \\ \text{end} \end{array}
```

Block nested loop join (2/2)

Costs:

- best case when smaller relation fits in memory
 - use it as inner relation
 - $B_R + B_S$
- worst case when memory holds one page of each relation
 - S scanned for each page in R
 - B_R * B_s + B_R

Block nested loop join (an improvement)

```
Memory size: M
for each M-1 page size chunk M_R of R
    for each page X<sub>S</sub> of S
           for each tuple t_R in M_R
                      for each t_s in X_s
                                 if (t<sub>R</sub> t<sub>S</sub> match) output t<sub>R</sub>.t<sub>S</sub>
                      end
           end
    end
end
```

Block nested loop join (an improvement)

Costs:

- best case when smaller relation fits in memory
 - use it as inner relation
 - B_R+B_S
- general case
 - S scanned for each M-1 size chunk in R
 - $(B_R / (M-1)) * B_s + B_R$

Indexed nested loop join

- R ⋈ S
- Index on inner relation (S)
- for each tuple in outer relation (R) probe index of inner relation
- Costs:
 - $-B_R + T_R * C$
 - c the cost of index-based selection of inner relation

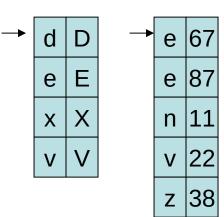
$$c \approx T(s)/V(s,A)$$

(if A is the join column and index is kept in memory)

relation with fewer records as outer relation

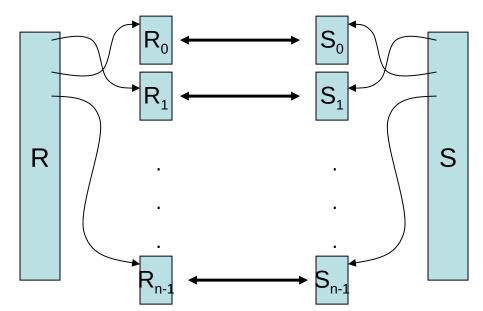
Sort-merge join

- 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



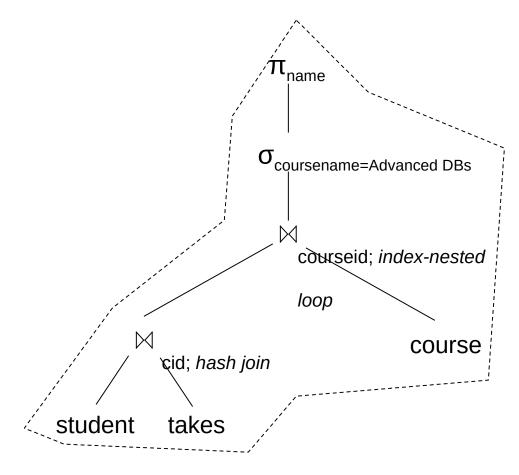
Hash join

- R ⋈ S
- use h_1 on joining attribute to map records to partitions that fit in memory
 - records of R are partitioned into $R_0...R_{n-1}$
 - records of S are partitioned into S₀... S_{n-1}
- join records in corresponding partitions
 - using a hash-based indexed block nested loop join
- Cost: $2*(B_R+B_S) + (B_R+B_S)$



Evaluation

- evaluate multiple operations in a plan
- materialization
- pipelining

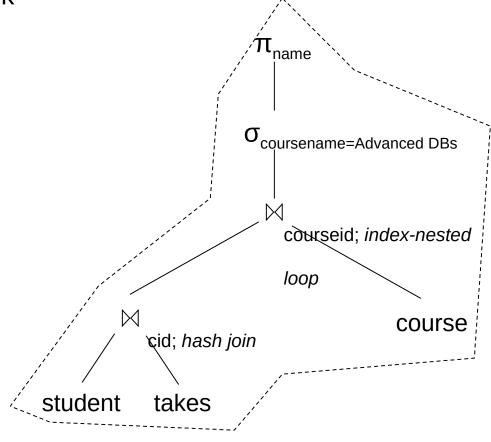


Materialization

create and read temporary relations

create implies writing to disk

more page writes



Pipelining (1/2)

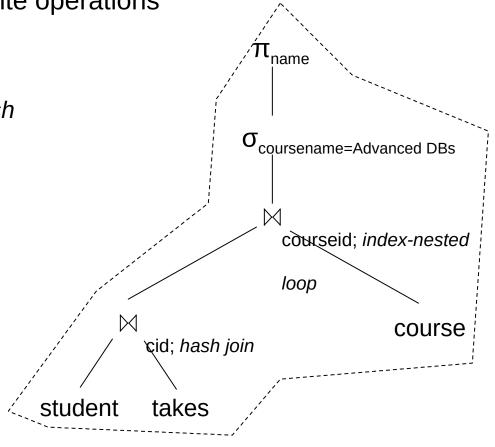
creating a pipeline of operations

reduces number of read-write operations

implementations

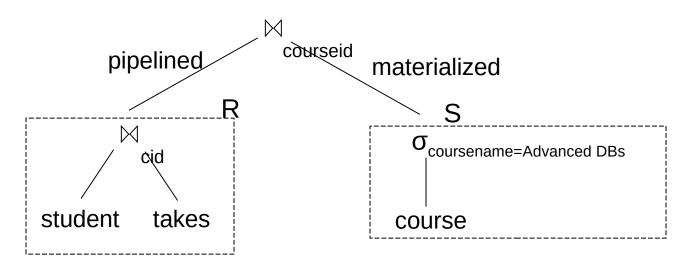
demand-driven - data pull

producer-driven - data push



Pipelining (2/2)

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

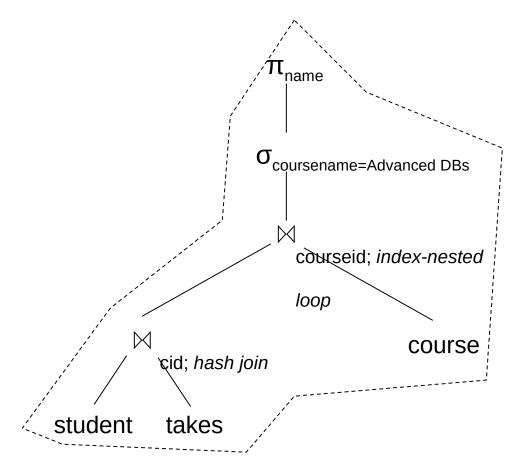


Choosing evaluation plans

- cost based optimization
- enumeration of plans
 - $R \bowtie S \bowtie T$, 12 possible orders (3! * 2) ($R \bowtie S$) $\bowtie T$, $R \bowtie (S \bowtie T)$
- cost estimation of each plan
- overall cost
 - cannot optimize operation independently

Cost estimation

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



Expression Equivalence

conjunctive selection decomposition

$$- \sigma_{\theta_1 \hat{\theta}_2}(R) = \sigma_{\theta_1}(\sigma_{\theta_2}(R))$$

commutativity of selection

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

combining selection with join and product

$$- \sigma_{\theta 1}(R \times S) = \dots R \bowtie_{\theta 1} S = \dots$$

commutativity of joins

$$- R^{\bowtie}_{\theta 1} S = S^{\bowtie}_{\theta 1} R$$

distribution of selection over join

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

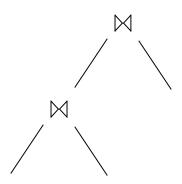
distribution of projection over join

$$- \pi_{A1,A2}(R \bowtie S) = \pi_{A1}(R) \bowtie \pi_{A2}(S)$$

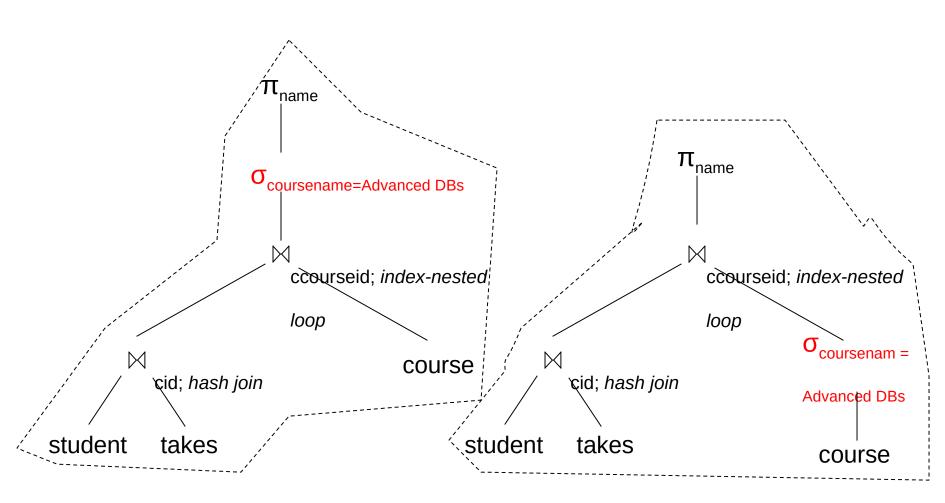
• associativity of joins: $R^{\bowtie}(S^{\bowtie}T) = (R^{\bowtie}S)^{\bowtie}T$

Cost Optimizer (1/2)

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



Cost Optimizer (2/2)



Summary

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