COL362/632 Introduction to Database Management Systems

Query Processing – Scans/Selections and Projections

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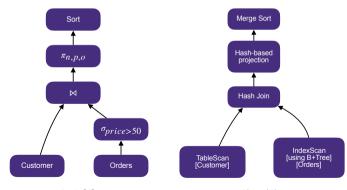
Outline

- Query Processing
- Selection and Scans
 - File Scan
 - Index Scan
- General Selections
- Selectivity Estimation
- Projections

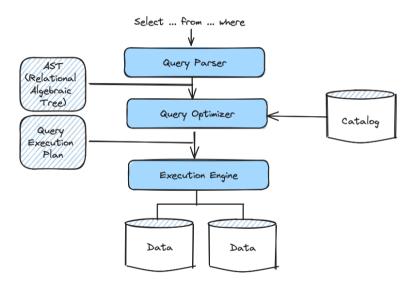
Query Processing (Overview)

- ► SQL : What?
- ▶ Physical plan: How?
- Query processing and optimization "compiles" a logical plan in to a physical plan

SELECT c.name, o.price, o.order_date FROM customer c, orders o WHERE c.customer_id = o.customer_id AND o.price > 50 ORDER BY o.order_date DESC;



Query Processing (Overview)



Cost of an Operation

Cost depends on

- CPU cost depends on
 - # tuples
 - # index entries
 - # operators or functions
- ► I/O cost depends on
 - # blocks transferred by disk (b)
 - # random I/Os (*S*)

We usually ignore CPU cost (for disk-based systems)

- ▶ cost of an operation = $b \times t_T + S \times t_S$
- \triangleright where, t_T is the time to transfer a block
- \triangleright and t_S is the block access time (recall: seek time plus rotational delay)

Cost of an Operation

Assumptions and heuristics

- cost of accessing a random page $=\frac{1}{10}^{th}$ of the actual cost (why?)
- \blacktriangleright B+Tree nodes are all in memory, and using a B+Tree only incurs I/O cost for leaf node

Optimizer's goal is to minimize the response time

- Hard to estimate response time depends on contents on the buffer, disks, data distribution
- ► Instead, minimize **resource consumption** (e.g., I/O cost)

Selections

- ▶ Access path: a "path" through which data can be located and accessed
- ▶ Scans are search algorithms that locate and retrieve "certain" records
 - 1. File scan
 - 2. Index scan

File Scan

Selection $(\sigma_{A=10})$

Linear search

$$ightharpoonup cost = t_S + b_r \times t_T$$

Linear search (if A is key)

• cost = $t_S + \frac{b_r}{2} \times t_T$ (average case)

Index Scan

Search algorithm that uses an index

- ▶ Use an index that is available on some predicate
- Cost depends on the index

Index Scan – B+Tree Index

Selection
$$(\sigma_{A=10})$$

Cost depends on height plus type of index

1. Clustering Index

Case: A is key

Case: A is not a key

Index Scan – B+Tree Index

Selection
$$(\sigma_{A=10})$$

Cost depends on height plus type of index

2. Non-clustering Index

Case: A is key

Case: A is not a key

► cost =
$$(h_i + n) \times (t_T + t_S)$$

 $n = \text{number of records fetched}$

Index Scan – B+Tree

Selection $(\sigma_{A>10} \text{ or } \sigma_{A\geq 10})$

1. Clustering Index

- ▶ Note: for $\sigma_{A < v}$ or $\sigma_{A < v}$, an index is not useful (Why?)

2. Non-clustering Index

- ▶ Only use when *n* is small
 - (HW: compare with Linear search. What would happen if n is large?)
- ▶ Often used with Bitmaps and sorting (HW: Read about Bitmap Index Scan in PostgreSql)

General Selections

- ▶ So far we only looked at σ_p where $p \equiv A \theta v$
- Predicates often involve
 - 1. conjunctions, i.e., $\sigma_{p_1 \wedge p_2 \wedge \cdots \wedge p_n}$ e.g., R.A = 10 AND R.B < 20
 - 2. disjunctions, i.e., $\sigma_{p_1 \vee p_2 \vee \cdots \vee p_n}$ e.g., R.A > 10 OR R.B < 20
- General selections require Index Matching
 - an index matches a selection predicate if the index can evaluate it
 - e.g, R(A,B,C) and a hash index on composite key (A,B)

 $\sigma_{A=5 \land B=10}$ matches the index

 $\sigma_{A=5}$ does not matches the index

Index Matching – Hash Index

Selection $(\sigma_{p_1 \wedge p_2 \wedge \cdots \wedge p_n})$

A hash index on composite key (A,B,..) matches the selection condition if

▶ all attributes in the index search key appear in the predicate with $\theta \in \{=\}$

Example R(A,B,C)

selection condition	hash index on (A,B)	hash index on (A)
A = 5 and B = 10	yes	yes
A = 5	no	yes
C = 5	no	no
A > 5 and $B < 10$	no	no
A = 5 and B = 10	yes	yes
A = 5 and $B = 10$ and $C < 1$	yes	yes

Index Matching – B+Tree

Selection $(\sigma_{p_1 \wedge p_2 \wedge \cdots \wedge p_n})$

A B+Tree index on composite key (A,B,..) matches the selection condition if

- ▶ the attributes in the predicate form a prefix of the search key of B+Tree
- ▶ $\theta \in \{=, <, >, \le, \ge\}$

Example R(A,B,C)

selection condition	B+Tree index on (A,B)	B+Tree index on (B,C)
A = 5 and B = 10	yes	yes
A = 5	yes	no
B = 5	no	yes
A > 5 and $B < 10$	yes	yes
A = 5 and B = 10	yes	yes
A = 5 and $B = 10$ and $C < 1$	yes	yes

Index Matching

When more than one index can match

Example

- ▶ B+Tree index on (A,B), and a hash index on (C)
- ▶ Selection condition: A = 5 and B = 3 and C = 10

Which index to use?

[Option 1] Use B+Tree index, retrieve tuples with (A=5 and B=3), then check C=10 for every retrieved tuple

[Option 2] Use hash index, retrieve tuples with (C=10), then check A=5 and B=3 for every retrieved tuple

[Option 3] Use B+Tree index, retrieve tuples with (A=5 and B=3), use hash index and retrieve tuples with C=10, intersect the record ids, only then fetch the tuples from disk

Disjunctive selection by union of identifiers

Example

- ► Hash index on (A) and B+Tree index on (B)
- ▶ Selection condition: A = 5 or B > 10
- ▶ [Option 1] Do a linear scan
- ▶ [Option 2] Use both indexes, fetch the records ids, and do a union, only then retrieve the tuples

Example

- ► Hash index on (A) + hash index on (B)
- ▶ Selection condition: A = 5 or B > 10
- ▶ [Option 1] Only option is to do a linear scan

Selectively Estimation

- Selectivity = fraction of tuples that need to be retrieved
- ► Goal should to chose the **most selective path** (Why?)
- ▶ But, estimating selectivity of an access path is a hard (active research area!)

Selectivity Estimation

Example

- hash index on (A, B)
- ▶ selection condition: A=3 and B=6

Selectivity $pprox rac{1}{\#\textit{keys}}$

- Number of keys can be determined from the index
- Only holds under uniform distribution assumption
- selection condition: A=3 and B=6 and C=10
- ► If no knowledge of #keys
 - selectivity = product of selectivity of primary conjuncts
 - or, use $\frac{1}{10}$ as default
 - holds only under independence assumption

Selectivity Estimation

Example

▶ Selection condition: 25 < A < 50

selectivity
$$\approx \frac{\text{interval}}{\text{max} - \text{min}}$$

Simple Projections

```
select R.A, R.B, from ... where ...
```

► Scan the file for each tuple output R.A, R.B

Projection w/ Duplicate Elimination

```
select distinct R.A, R.B from ... where ...
```

- first do simple projections
- remove duplicates
 - Sort-based duplicate elimination
 - Optimization: eliminate duplicates when merging runs
 - Hash-based duplicate elimination