

Database Administration

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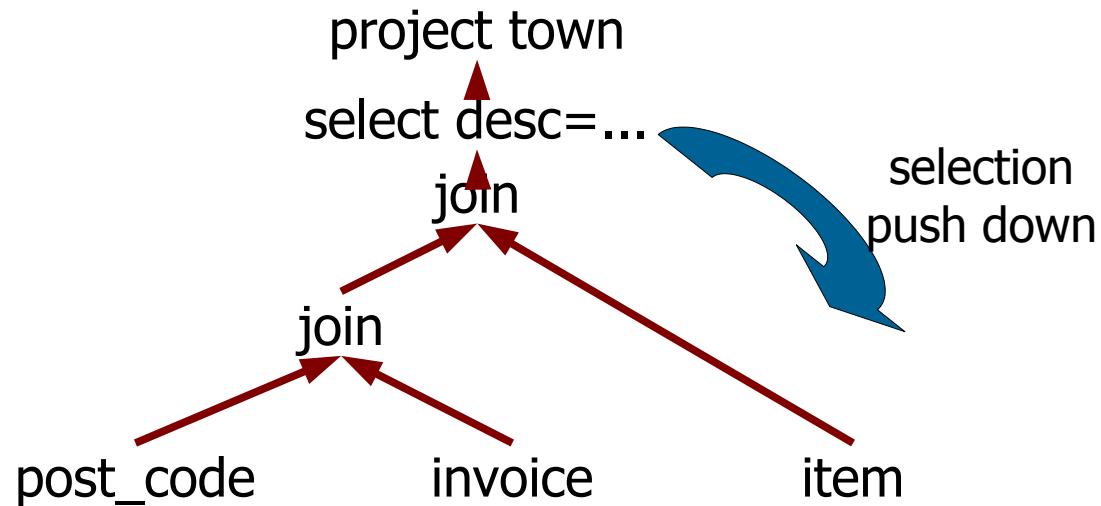


Roadmap

- What physical operators exist for each logical operation?
- How are physical operators implemented and composed?
- How is the best plan found?

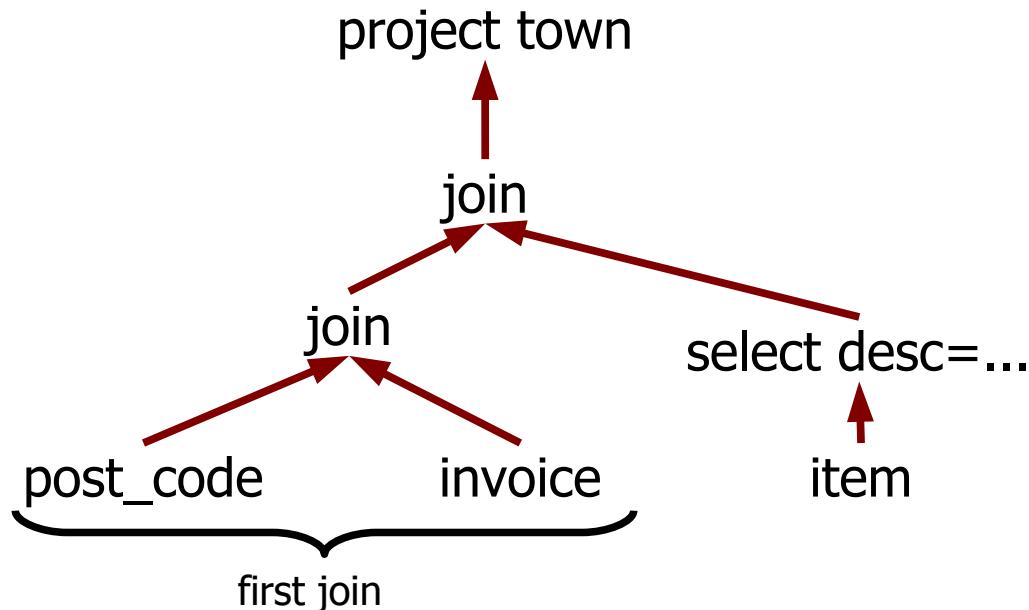
Optimization

```
select town from  
post_code natural join invoice  
natural join item where desc=...
```



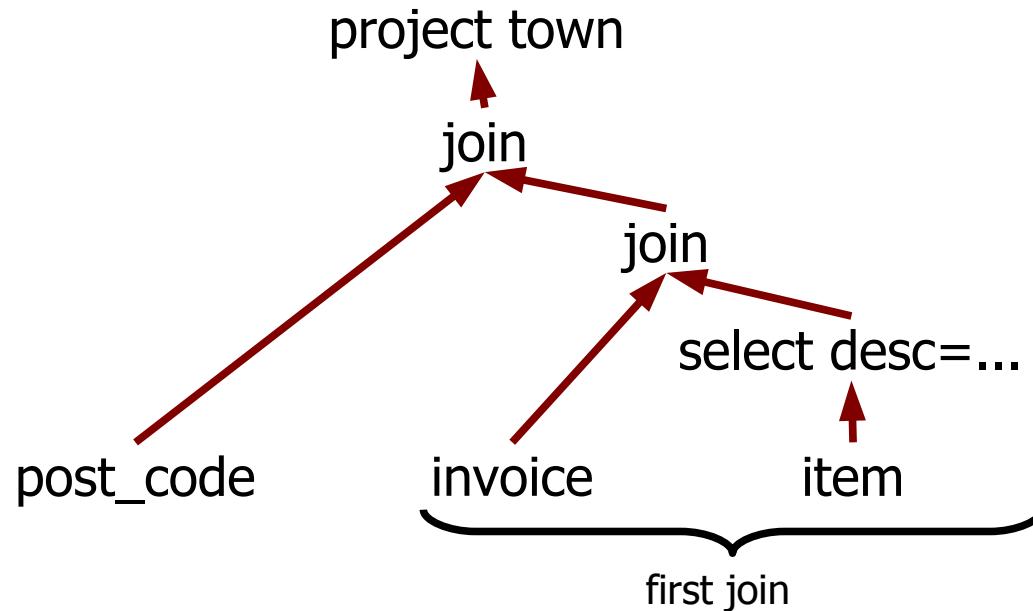
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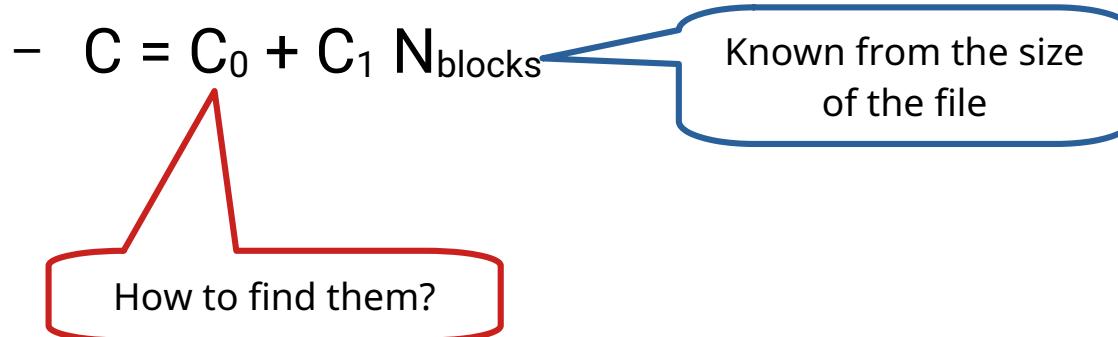
- How to estimate the cost of a plan?
- How to find alternative plans?

Cost estimation

- Tradeoff between:
 - Actually executing the query and measuring what resources it consumes and how long it takes to execute
 - vs
 - **Estimate that can be computed quickly**
 - vs
 - Considering that all queries have the same cost
- We don't want to know what is the actual cost
- We want to know which alternative costs less
 - Use a model that monotonously approximates real cost

Example: Simple sequential scan

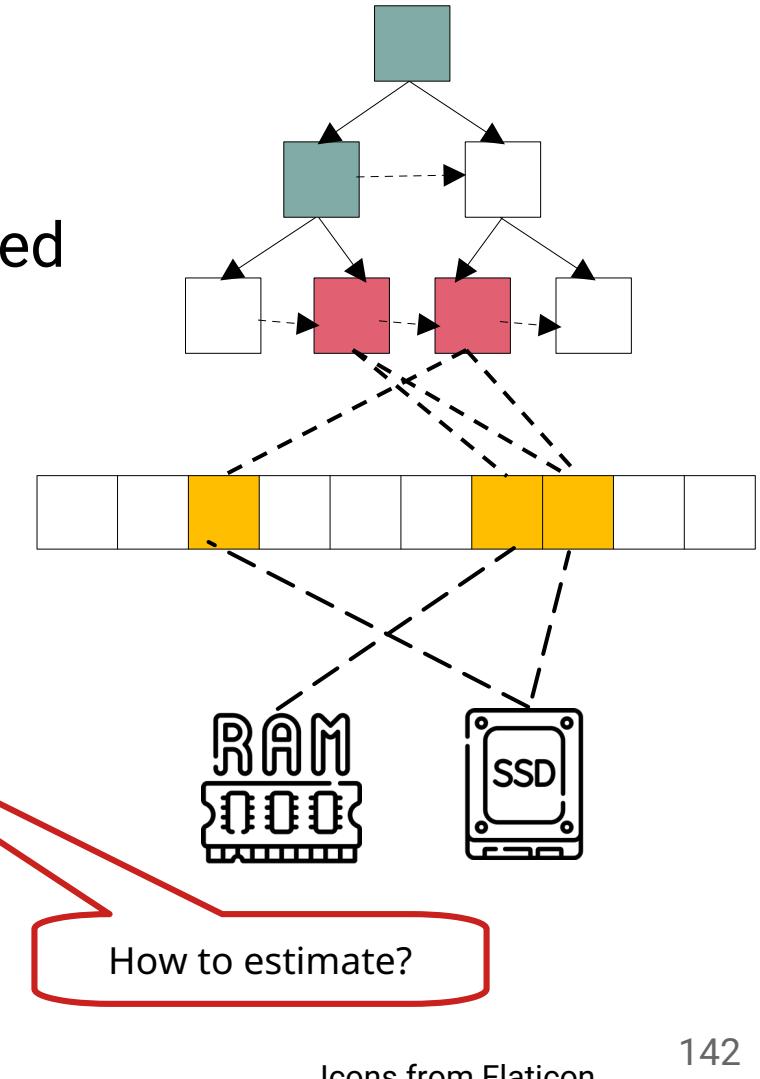
- Assumption of main cost factor:
 - Number of disk block I/O operations
- Cost model as a simple equation:



Example: Index scan

- Cost factors:
 - Inner tree blocks $\sim \log$ size of table
 - Tree leaf blocks \sim number of selected rows
 - Table blocks:
 - if clustered $\sim \frac{\text{selected rows}}{\text{rows per block}}$
 - if somewhat clustered... ?
 - if not clustered $\sim \text{selected rows}$
- Average cost of block depends on effectiveness of caching

How to estimate?



Example: Hash Join

- CPU cost:
 - Building hash table from inner table ~ N_r input rows
 - Checking each row in the outer table ~ N_i input rows
 - Creating the resulting N_m rows
 - Very high cost if not enough memory to hold N_i rows in the hash table
-
- N_r
- N_i
- N_m
- How to estimate?

Coefficients

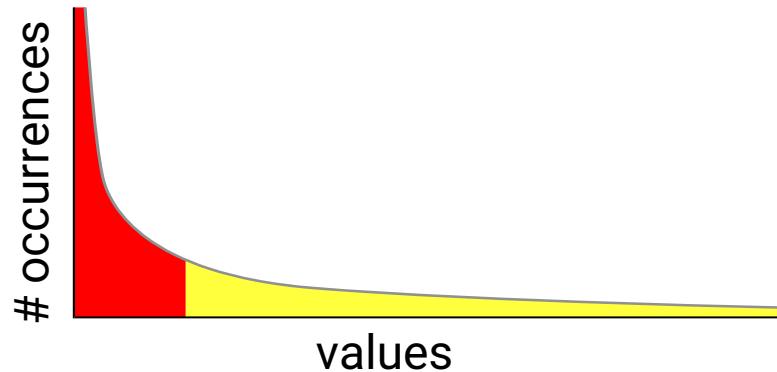
- Cost coefficients depend on the actual hardware, software, and even workloads
- Can be estimated by profiling simple workloads
- Can be tuned by the DBA

Cardinality - Selection

- Assumption of uniform distribution
- Know #distinct
 - Expected copies of each tuple:
 - #rows / #distinct

Cardinality - Selection

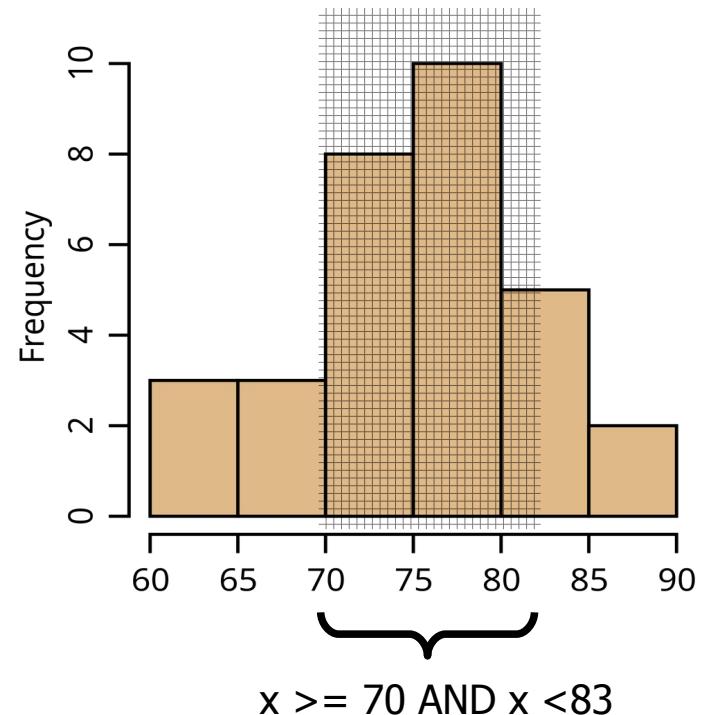
- Real data are usually not uniformly distributed:
 - 80/20 rule
 - Power law



- Know most popular and #occurrences
- Compute estimated #occurrences of others as uniformly distributed

Cardinality - Selection

- Data often have complex multimodal distributions
- Know histogram:
 - % of occurrences in each interval or
 - interval for fixed % of occurrences
- Compute #occurrences as uniformly distributed within each interval



Cardinality - Conjunction

- Filter conditions are often the conjunction of conditions on different columns
- Statistics on multiple columns are expensive:
 - Multidimensional
 - Many possible combinations
- Assume independently distributed values in different columns:
 - selectivity = #selected / #total rows
 - $\text{selectivity}(a \wedge b) = \text{selectivity}(a) * \text{selectivity}(b)$
- Idem for disjunction

Cardinality - Join

- Cardinality for cross-product of A and B:
 - $\# \text{rows from A} \times \# \text{rows from B}$
- Cardinality for join, first attempt:
 - Align buckets of histograms for A and B
 - Estimate the cardinality of each matching bucket as a cross-product:
 - Not good, as there are non-matching values
 - e.g. A has even numbers, B has odd numbers → no match!

Cardinality - Join

- Assume that one relation A has all values (containment)
- Cardinality for join, second attempt:
 - Align buckets of histograms for A and B
 - Estimate the cardinality of each matching bucket:
 - Each of #rows in B matches ($\# \text{rows in A} / \# \text{distinct in A}$)
 - Because A has all the values
 - Estimate as $\# \text{rows in B} \times \# \text{rows in A} / \# \text{distinct in A}$
 - Generalize for the case where B has all values:
 - $\# \text{rows in A} \times \# \text{rows in B} / \max(\# \text{distinct in A}, \# \text{distinct in B})$

Summary

- Tradeoff between complexity (data and computation) and accuracy
- Many other techniques:
 - More statistics
 - Heuristics
 - Hinting
 - Sampling of query
 - Machine learning
 - ...

Cost model in PostgreSQL

- Computes two costs for each (sub-)plan:
 - (cost of first row .. cost of last row)
 - Interesting cases:
 - (0 .. big value) → row at a time operators
 - (big value ... big value) → full relation operators
- Coefficients can be tuned in postgresql.conf
 - See “Planner Cost Constants”



Statistics in PostgreSQL

- Automatically keeps:
 - # distinct / # records
 - most popular
 - histogram
 - correlation
 - measures sorting / clustering of a column
- Computation of statistics:
 - Using sampling on large tables
 - Explicitly with ANALYZE
 - Implicitly as part of the “autovaccum” process



Statistics in PostgreSQL

- Multivariate statistics relate multiple columns:
 - Avoid the assumption of independence
- Statistics on multiple columns or arbitrary expressions
- Available statistics:
 - `nndistinct` → distinct combinations of columns / expressions
 - `mcv` → most common combinations
 - `dependencies` → relation between columns / expressions
- Useful for join estimation when containment assumption does not hold

