

Scope

In this lesson we look at assessing and improving query performance with the use of the EXPLAIN command and indexes.

SQL is a declarative language

SQL is a **declarative** language not an **imperative** language.

We specify what the result should be, not how it's done.

Declarative languages focus on what the outcome should be, rather than how to achieve it. Developers specify desired results and the system determines the steps required to achieve those results.

- Declarative languages include : HTML, css, prolog, ..
- Imperative languages are: python, javascript, Go, Java etc ... and bash

The relation in *Relational database*

Relational database come from **relation** theory.

- A and B are sets of objects: for instance $A = \{1,2,3\}$, $B = \{a,b,c\}$
- *tuples* are combination of elements of these objects : (1,a), (1,b), ..., (3, c)

In that context, a database **table** becomes a **subset** of all the possible combinations of the values of its columns.

For instance a table with columns A and B that can respectively take the values {1,2,3} and {a,b,c} is :

A	B
1	a
1	b
2	a
2	b
3	a
3	b

That's all for relation theory!

For more on that subject: chapter 2 and 3 of *PostgreSQL Query Optimization*2021_ (see the discord channel for the pdf)

Everything is a relation

One important concept from in relation theory, is that **an operation on a relation produces ... another relation!**

the whole table is already a subset of all possible column value combinations.

- Depth: A SELECT based on WHERE conditions, subsets the table
- Width: A SELECT on specific columns : subsets the attributes

A SELECT query produces a subset of the subset.

Depth x Width : (number of rows) x (number of attributes)

Which implies:

- a query produces a relation
- a column can have values that are relations

We saw that in the definition of the 1NF form:

A relation is in 1NF if and only if no attribute domain has **relations** as elements.

Understanding that sets of rows are in fact relations helps with the versatility of the SQL language.

For instance, it makes sense to use temporary relations defined from named queries using

```
WITH ... AS .
```

database optimizer, query planner

The **database optimizer** or **query planner** or **engine** chooses the most efficient method to get to the result as expressed by the query.

The database optimizer interprets SQL queries along this process:

- **compilation**: transforms the query into a high level logical plan
- **optimization**: transforms logical plan into execution plan by choosing the most efficient algorithms for the job
- **execution**: executes the algorithms

Woo! Algorithms

Here, think of **algorithms** as the actual **physical operations**: getting to the data which is stored in memory or in physical storage, scanning the data, looking up indexes, filtering etc.

In the context of the PostgreSQL query optimizer, an **algorithm** is a systematic method to analyze, plan, and execute database queries efficiently.

Choosing the right Algorithm

How does the query planner chooses the best algorithms to execute the query ?

Many factors can be involved

- the nature of the tables and the data
- the query itself
- the type of filter (WHERE) operator : `=` , `<` , `between` , `like '%string'`
etc
- the data types of the columns involved in the filters
- the number of rows of the table or the subqueries etc
- presence of `limit` , `order by` , `sort by`
- full text search is a topic by itself
- the performance of the machine

For each set of data (distribution, type) and indexes, etc ... there are specific rules and heuristics on which the query planner will base its choice of the most efficient algorithm to execute the query.

Understanding a query plan can be difficult.

Straight from the documentation : **Plan-reading is an *art* that requires some experience to master**,, emphasis here on the word *art*.

Planning and Cost reduction

Given a query, the role of the query optimizer is to find the most **efficient** way to execute it.

What is **efficient** ? Let's say for the moment that efficient means low query execution time.

The optimizer minimizes the **cost of execution** to lower execution time.:

- rewrites the query
- chooses the most efficient physical operations (i.e. algorithms) to execute the query

Cost function

However, the real execution time of a query also depends on the machine the server is running on. This machine may be running other processes that can consume available resources.

The query optimizer does not have access to the overall machine context. and the overall query time as experienced by the user is unknown to the optimizer. It also does not have access to external metrics such as monetary cost, user satisfaction, etc.

Instead the optimizer combines a measure of **CPU cycles** and **I/O accesses** to derive a **cost function**.

The **query optimizer** relies on resources that affect execution time: CPU cycles and I/O accesses which it combines into a **single cost function** that it tries to minimize.

Depending on the machine that hosts the PostgreSQL server, PostgreSQL assigns different weights to these different factors to calculate the cost of a query

For instance:

```
SQL
SHOW cpu_operator_cost;
```

returns:

```
Bash
cpu_operator_cost
-----
0.0025
```

The `cpu_operator_cost` is defined as: *Sets the planner's estimate of the cost of processing each operator or function executed during a query. The default is 0.0025.*

The `cpu_tuple_cost` is defined as : *Sets the planner's estimate of the cost of processing each row during a query. The default is 0.01.*

There is a total of [13 cost related weights](#). All have default values.

Blocks, storage and I/O access

There's no magic. At the bottom of things the database data is stored in files.

Find where that data is with:

```
SQL
SHOW data_directory;
```

These are the folders you will find. The data is in `base` .

Folder	Description
<code>base</code>	Stores actual data for user-created databases.
<code>global</code>	Contains system-wide metadata and cluster control files.
<code>pg_wal</code>	Write-Ahead Logs for transaction logging and recovery.
<code>pg_multixact</code>	Manages MultiXact transactions for row locking.
<code>pg_subtrans</code>	Tracks sub-transactions for nested transactions.
<code>pg_commit_ts</code>	Stores commit timestamps of transactions.
<code>pg_tblspc</code>	Symbolic links to tablespaces for external data storage.
<code>pg_stat_tmp</code>	Temporary statistics data for monitoring performance.
<code>pg_snapshots</code>	Stores transaction snapshots for MVCC isolation.
<code>pg_twophase</code>	Manages two-phase commit transactions.
<code>pg_logical</code>	Handles logical replication and WAL decoding.
<code>pg_replslot</code>	Stores replication slots for data streaming replication.
<code>pg_serial</code>	Tracks serializable transactions.
<code>pg_stat</code>	Collects statistics about database activity.
<code>pg_xact</code>	Tracks transaction commit/abort statuses.
<code>pg_dynshmem</code>	Stores dynamic shared memory for process coordination.

Any file used for database objects is divided in **blocks** of the same length;

A block is the unit that is transferred between the hard drive and the main memory, and the number of I/O operations needed to execute any data access is equal to the number of blocks that are being read from or written to.

By default, PostgreSQL uses blocks containing 8192 bytes each.

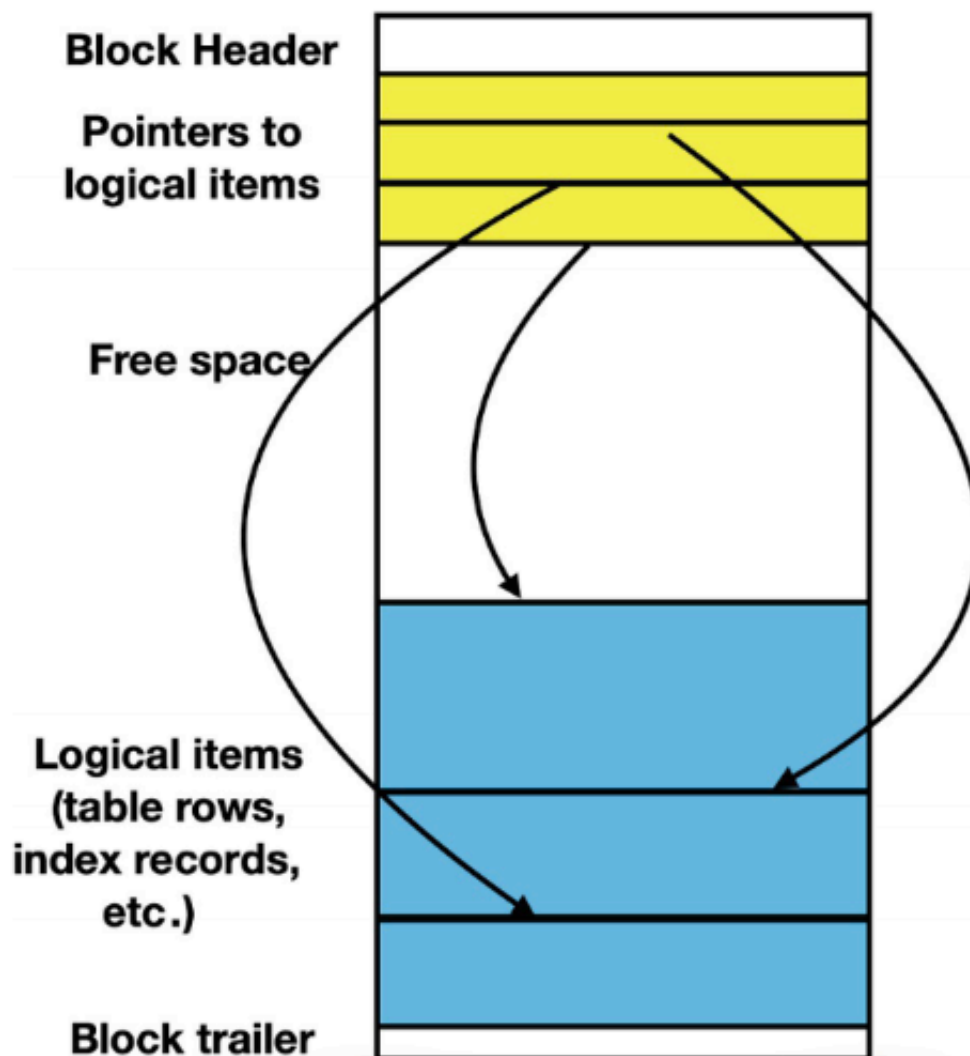
Looking up one of the subfolders of `base` :

```

ll /usr/local/var/postgresql@16/base/1
total 15304
drwx----- 301 alexis  admin   9.4K  6 Sep 09:05 .
drwx-----  16 alexis  admin   512B 23 Sep 11:55 ..
-rw-----   1 alexis  admin   8.0K 22 Mar 2024 112
-rw-----   1 alexis  admin   8.0K 22 Mar 2024 113
-rw-----   1 alexis  admin  120K 22 Mar 2024 1247
-rw-----   1 alexis  admin   24K 22 Mar 2024 1247_fsm
-rw-----   1 alexis  admin   8.0K 22 Mar 2024 1247_vm
-rw-----   1 alexis  admin  456K 22 Mar 2024 1249

```

Several small items can reside in the same block; larger items may spread among several blocks.



EXPLAIN <SQL>

You can use the `EXPLAIN` command in front of a SQL query to see what query plan the

planner creates for that query.

Postgresql optimization engine is excellent but sometimes needs troubleshooting.

Important : The optimization engine will return different plans depending on the machine it runs on so the same query may result in different execution plans on different hardware.

Example

Let's look at the query plan for a query on the WorldHits database: the

`versatility_score` query from S04.04.

Just add EXPLAIN in front of the query:

```
SQL
EXPLAIN WITH artist_stats AS (
  SELECT
    artist,
    COUNT(*) AS track_count,
    ROUND(CAST(STDDEV(energy) AS numeric), 2) AS energy_std,
    ROUND(CAST(STDDEV(danceability) AS numeric), 2) AS danceability,
    ROUND(CAST(STDDEV(acousticness) AS numeric), 2) AS acousticness,
    ROUND(CAST(STDDEV(instrumentalness) AS numeric), 2) AS instrumentalness
  FROM tracks
  GROUP BY artist
  HAVING COUNT(*) >= 5
),
artist_versatility AS (
  SELECT
    artist,
    track_count,
    ROUND(CAST(energy_std + danceability_std + acousticness_std + instrumentalness_std AS numeric), 2) AS versatility_score
  FROM artist_stats
)
SELECT
  artist,
  track_count,
  versatility_score
FROM artist_versatility
ORDER BY versatility_score DESC
LIMIT 10;
```

We get the execution plan:

QUERY PLAN

```

-----
Limit  (cost=27.77..27.80 rows=10 width=54) (actual time=0.498..0.501 r
-> Sort  (cost=27.77..27.83 rows=22 width=54) (actual time=0.497..0.4
    Sort Key: (round((((artist_stats.energy_std + artist_stats.dance
    Sort Method: top-N heapsort  Memory: 26kB
    -> Subquery Scan on artist_stats  (cost=25.15..27.30 rows=22 w
        -> HashAggregate  (cost=25.15..27.08 rows=22 width=150) (
            Group Key: tracks.artist
            Filter: (count(*) >= 5)
            Batches: 1  Memory Usage: 48kB
            Rows Removed by Filter: 29
        -> Seq Scan on tracks  (cost=0.00..20.26 rows=326 w

```

Lots of information!

Nodes

A **node** in a query plan represents a **single operation** or step in the execution of a query. A node is a line.

Each node describes a specific action that Postgres will execute, such as

- scanning a table,
- applying a filter,
- joining tables,
- or sorting results.

Nodes are typically arranged in a tree structure, with **child nodes** feeding results to their **parent nodes**, ultimately leading to the final result of the query.

Sequential Scan

Let's look at the query plan for a simple select query.

```
explain select * from trees;
```

QUERY PLAN

```

-----
Seq Scan on trees  (cost=0.00..4577.39 rows=211339 width=50)
(1 row)

```

The operation is a **Sequential Scan** on the `trees` table.

Sequential Scan: the algorithm is simply reading all rows from the table in order, from beginning to end. It's the most straightforward way to retrieve all data when no filtering is required.

The query plan shows:

- The **estimated cost** ranges from 0.00 (start-up cost) to 4577.39 (total cost).
- It expects to return **211,339 rows**.
- The average **width** of each row is estimated to be 50 bytes.

Note: The **width** is the estimated average size of each row in bytes.

Since we're selecting all columns and all rows, Postgres has no choice but to read the entire table sequentially.

A Sequential Scan is the most efficient method for retrieving all data from a table.

Costs of a Sequential Scan

The estimated cost for a Sequential Scan is computed as :

```
(number of pages read from disk * seq_page_cost) + (rows scanned * cpu_
```

By default, `seq_page_cost = 1.0` and `cpu_tuple_cost = 0.01`,

see <https://www.postgresql.org/docs/current/runtime-config-query.html#GUC-SEQ-PAGE-COST> and <https://www.postgresql.org/docs/current/runtime-config-query.html#GUC-CPU-TUPLE-COST>

We can lookup the number of pages for a given table with

```
SELECT relname, relkind, reltuples, relpages
FROM pg_class
WHERE relname = 'trees';
```

which returns

relname	relkind	reltuples	relpages
trees	r	211339	2464

The trees table occupies 2464 pages.

So the estimated cost is $(2464 * 1.0) + (211339 * 0.01) = 4577.39$ which is the exact number we got in the query plan.

The `pg_class` table

The `pg_class` table in PostgreSQL is a system catalog that contains metadata about tables, indexes, sequences, views, and other relation-like objects.

Column	Description
<code>relname</code>	Name of the relation (table, index, etc.).
<code>reltuples</code>	Estimated number of rows in the table or index.
<code>relpages</code>	Number of disk pages that the table or index uses (approximate).
<code>relkind</code>	Type of relation: 'r' (table), 'i' (index), 'S' (sequence), 'v' (view), etc.
<code>relhasindex</code>	<code>true</code> if the table has any indexes.
<code>...</code>	other columns

`pg_class` tracks:

- Tables and Views
- Indexes
- Sequences
- Materialized Views
- Partitions

EXPLAIN ANALYZE

EXPLAIN ANALYZE complements the EXPLAIN command by also sampling the data and calculating statistics that will be used to optimize the plan.

While EXPLAIN uses default statistical estimates about the data, EXPLAIN ANALYZE uses real stats.

Furthermore, EXPLAIN ANALYZE

- actually runs the query (but does not display the output)
- calculates the real time it takes to run
- returns the true number of rows

And

- **I/O timings:** time spent on I/O operations.
- **Worker information:** Details about workers used in parallel queries.
- **Memory usage:** memory used during query execution.
- **Buffers information:** details about shared and local buffer usage, including hits and reads.

EXPLAIN ANALYZE on updates

Since EXPLAIN ANALYZE actually runs the query, you have to be cautious when using it on UPDATE, INSERT or DELETE queries 💣💣💣

- use EXPLAIN first to check for missing indexes and get a rough idea about possible optimizations then after you've optimized the tables and your query, run EXPLAIN ANALYZE to get an accurate timing of the query execution

Rollback

To analyze a data-modifying query without changing the tables, wrap the EXPLAIN ANALYZE statement between `BEGIN` and `ROLLBACK`. That way the query will be rolled back after its execution.

For instance:

```
BEGIN;

EXPLAIN ANALYZE UPDATE trees SET height = height / 10 WHERE height > 100;

ROLLBACK;
```

EXPLAIN a WHERE

Let's add a `where` clause on the height attribute, to our simple `select * from trees;` and see what the optimizer makes of it

```
EXPLAIN select * from trees where height < 9;
```

returns:

QUERY PLAN

```
-----
Seq Scan on trees (cost=0.00..5105.74 rows=111270 width=50)
  Filter: (height < 9)
(2 rows)
```MMi

```

Compare that to the original query plan without the filter:

```
```sql
```

QUERY PLAN

```
-----
Seq Scan on trees (cost=0.00..4577.39 rows=211339 width=50)

```

So the cost has gone up (4577.39 -> 5105.74) while the number of rows has gone down (211339 -> 111270) !

It's because we're asking the Seq Scan node to check the condition for each row it scans, and outputs only the ones that pass the condition.

The estimate of output rows has been reduced because of the WHERE clause.

However, the scan still has to visit all 211339 rows, so the cost hasn't decreased; in fact it has gone up a bit (by `211339 * cpu_operator_cost` you can check) to reflect the extra CPU time spent checking the WHERE condition.

```
treesdb_v04=# SHOW cpu_operator_cost;
cpu_operator_cost
-----
0.0025
(1 row)

```

So now the cost is calculated by

```
(number of pages read from disk * seq_page_cost) + \
(rows scanned * cpu_tuple_cost) + \
(rows scanned * cpu_operator_cost)

```

The impact of indexing

The trees table for the primary key `id`

```
\d+ trees
Indexes:
    "trees_pkey" PRIMARY KEY, btree (id)
Access method: heap
```

SQL

Let's see how this index `trees_pkey` is put to use by the query planner when we filter on `id` instead of `height`

```
EXPLAIN select * from trees where id < 1000
```

SQL

We get

```
QUERY PLAN
-----
Bitmap Heap Scan on trees (cost=189.37..2777.48 rows=9929 width=50)
  Recheck Cond: (id < 10000)
    -> Bitmap Index Scan on trees_pkey (cost=0.00..186.89 rows=9929 width=0)
          Index Cond: (id < 10000)
(4 rows)
```

SQL

The query planner executed a two step plan where the **child** node: **Bitmap Index Scan** feeds the rows into a **parent** node: **Bitmap Heap Scan**.

`HEAP` meaning the entire table.

The overall costs has gone down since the parent scan only has to check the filtering condition on 10k rows

BITMAP

A **bitmap** is a compact data structure used to represent a set of row identifiers.

- **Structure:** It is essentially an array of bits (0s and 1s).
- **Representation:** Each bit corresponds to a row in the table. A '1' indicates the row matches the query condition, while a '0' means it doesn't.
- **Efficiency:** Bitmaps are memory-efficient for representing large sets of rows, especially when dealing with millions of records.
- **Operations:** Bitmaps allow for quick set operations (AND, OR, NOT) when combining multiple conditions.
- **Purpose:** In a **Bitmap Heap Scan**, the bitmap helps the database quickly identify which rows need to be fetched from the table, avoiding unnecessary I/O.

Difference between Bitmap Heap Scan and Seq Scan

- Sequential Scan (Seq Scan):
 - Reads all rows in the table sequentially, one after another.
 - Doesn't use any index.
 - Efficient for reading a large portion of the table.
- Bitmap Heap Scan: Uses a two-step process:
 - a) First, it creates a bitmap in memory where each bit represents a table block.
 - b) Then, it actually fetches the rows from the table based on this bitmap.
 - Often used when an index scan returns a moderate number of rows.

The key difference lies in how these methods access the data:

- Reading sequentially in Seq Scan:

The database reads all data pages in order, one after another. This is efficient because it follows the natural order of data on disk.

- Fetching rows separately in Bitmap Heap Scan:

After creating the bitmap, the database jumps around the table to fetch only the specific rows that match the condition.

This involves more random I/O operations, which are generally slower than sequential reads.

In the end, the cost is lower with a bitmap scan because the planner has to deal with a much smaller number of rows

but if the filtering is less stringent, the planner will still use a Seq Scan

For instance:

```
SQL
treesdb_v04=# EXPLAIN select * from trees where id < 150000;
               QUERY PLAN
-----
Seq Scan on trees (cost=0.00..5105.74 rows=149967 width=50)
  Filter: (id < 150000)
(2 rows)
```

Because in that case when the filtering is too wide, the advantage of using a bitmap scan is too small.

Note: We could probably plot a curve by increasing the id threshold and seeing when the cutoff from Seq Scan to Bitmap Scan occurs

Explore some more

EXPLAIN the following queries

Index Scan

```
select * from trees where id = 808;
```

SQL

```
select * from trees where id < 10000 and stage_id = 1;
```

SQL

In that query, why does the cost is not lower (343 rows in the parent node)

The added condition `stage_id = 1;` reduces the output row count estimate, but not the cost because we still have to visit the same set of rows. Notice that the `stage_id = 1` clause cannot be applied as an index condition, since this index is only on the `id` column. Instead it is applied as a filter on the rows retrieved by the index. Thus the cost has actually gone up slightly to reflect this extra checking.

equality on a column with index

```
treesdb_v04=# explain select * from trees where id = 808;
```

SQL

QUERY PLAN

```
-----  
Index Scan using trees_pkey on trees (cost=0.42..8.44 rows=1 width=50)  
Index Cond: (id = 808)
```

The table rows are fetched in index order, which makes them even more expensive to read, but there are so few that the extra cost of sorting the row locations is not worth it. You'll most often see this plan type for queries that fetch just a single row.

It's also often used for queries that have an ORDER BY condition that matches the index order, because then no extra sorting step is needed to satisfy the ORDER BY. In this example, adding ORDER BY `id` would use the same plan because the index already implicitly provides the requested ordering.

much scans many index

We already have three types of scans:

- Seq Scan
- Bitmap Heap Scan
- Index Scan

Let's recap

1. Index Scan:

- Used when: Fetching a small number of rows based on an index condition.
- How it works:
 - Directly uses the index to find the location of rows matching the condition.
 - Then fetches those specific rows from the table.
- Efficiency: Very efficient for retrieving a small number of rows, especially with highly selective conditions.
- In your example: It's using the primary key index (trees_pkey) to quickly find the row with id = 808.

1. Sequential Scan (Seq Scan):

- Used when: Reading a large portion of the table or when no suitable index is available.
- How it works:
 - Reads all rows in the table sequentially, one after another.
 - Checks each row against the condition (if any).
- Efficiency: Efficient for reading large portions of a table, but can be slow for selective queries on large tables.

1. Bitmap Heap Scan:

- Used when: Fetching a moderate number of rows based on an index condition.
- How it works:
 - Two-step process: a) First, creates a bitmap in memory where each bit represents a table block that contains matching rows. b) Then, fetches the actual rows from the table based on this bitmap.

- Efficiency: More efficient than Index Scan when retrieving a larger number of rows, but still selective enough not to warrant a full Sequential Scan.

Key differences:

1. Data access pattern:

- Index Scan: Might jump around the table to fetch specific rows.
- Seq Scan: Reads the entire table sequentially.
- Bitmap Heap Scan: First creates a map of relevant blocks, then fetches those blocks (which can be more organized than Index Scan but less sequential than Seq Scan).

2. Use of indexes:

- Index Scan: Directly uses the index.
- Seq Scan: Doesn't use any index.
- Bitmap Heap Scan: Uses an index to create the bitmap, then accesses the table.

3. Suitability for different data volumes:

- Index Scan: Best for very selective queries (few rows).
- Seq Scan: Best for reading large portions of the table.
- Bitmap Heap Scan: Good middle ground for moderate selectivity.