

# Evaluating and optimizing SQL query with Postgres

## Explain Analyze

# An HTTP request journey



# Get the query and Read the Analyze

What you need:

- A way to get the raw SQL query
- A way to execute the analyze in production and retrieve the result

# Getting the raw sql

Django => `.query`

SqlAlchemy => `.str()`

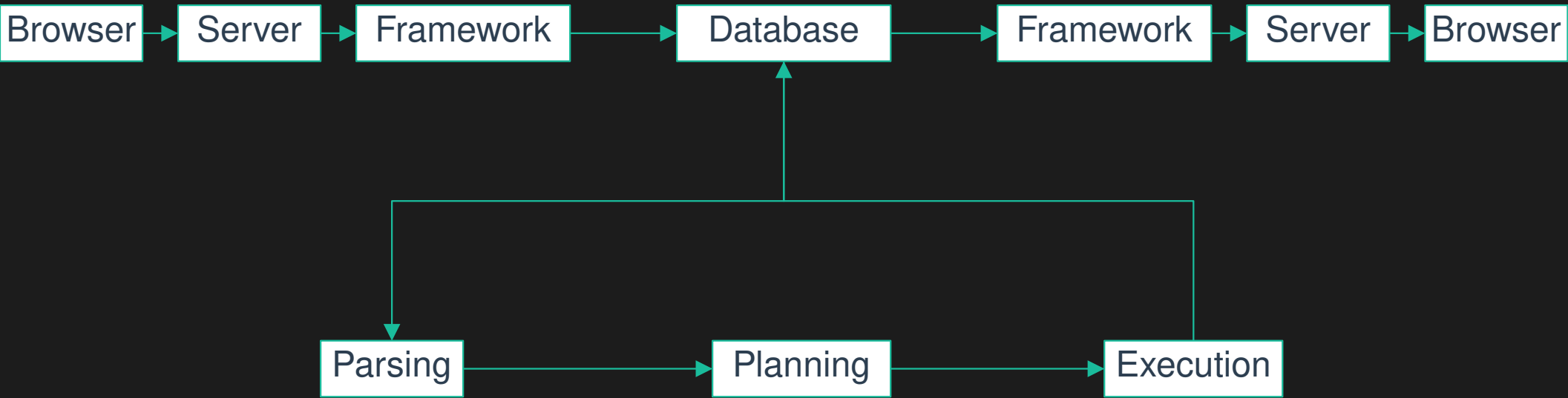
SqlModel =>

```
print(query.compile(compile_kwargs={'literal_bindings': True}))
```

- Query log
- `Import pdb;`  
`pdb.set_trace()`

# Live example

# An HTTP request journey



# Table stats

```
afpy_db=# explain analyze select * from activities;
```

## QUERY PLAN

```
-----  
Seq Scan on activities  (cost=0.00..114787.33 rows=6261033 width=56) (actual time=0.247..163.826 rows=6261121.00 loops=1)  
  Buffers: shared hit=282 read=51895  
Planning Time: 0.051 ms  
Execution Time: 289.072 ms  
(4 rows)
```

```
afpy_db=# select relpages, reltuples from pg_class where relname='activities';
```

```
relpages | reltuples  
-----+-----  
    52177 | 6.261033e+06  
(1 row)
```

```
afpy_db=# select 52177 * 1.0 + 6261033 * 0.01;  
?column?
```

```
-----  
    114787.33  
(1 row)
```

# pg\_stats



# Ways to optimize for a Query

- 1) Optimize the raw query
- 2) Add an index or remove one
- 3) Upgrade your database spec: work\_mem and shared buffers

# Live example

From sequential scan to index only scan

# Conclusion    How to Automate the Process?

- HypoPG
- PG analyze



**Questions**