

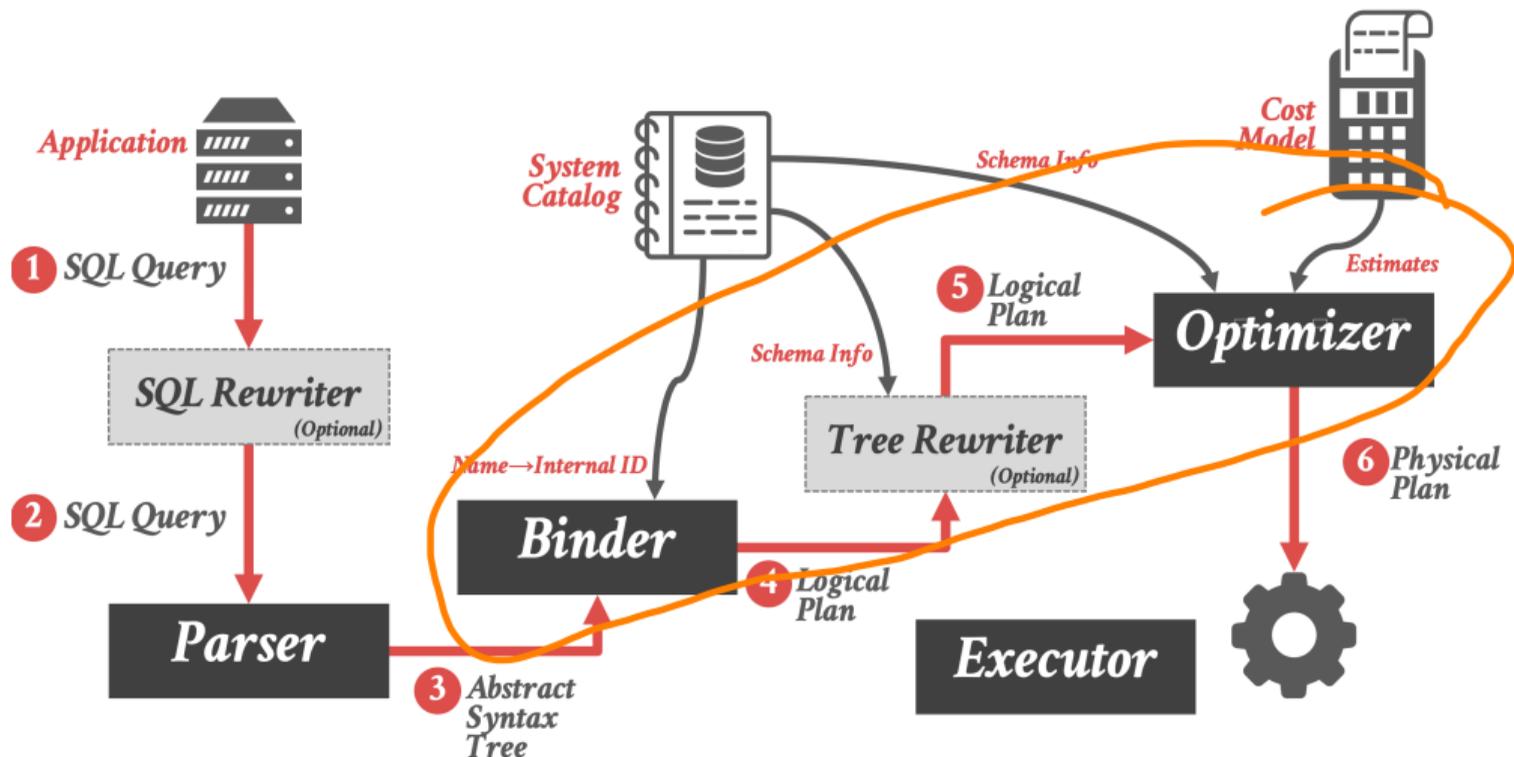
(April 7
Mid-term exam
April)

Lecture 19: Rule-Based Query Optimization

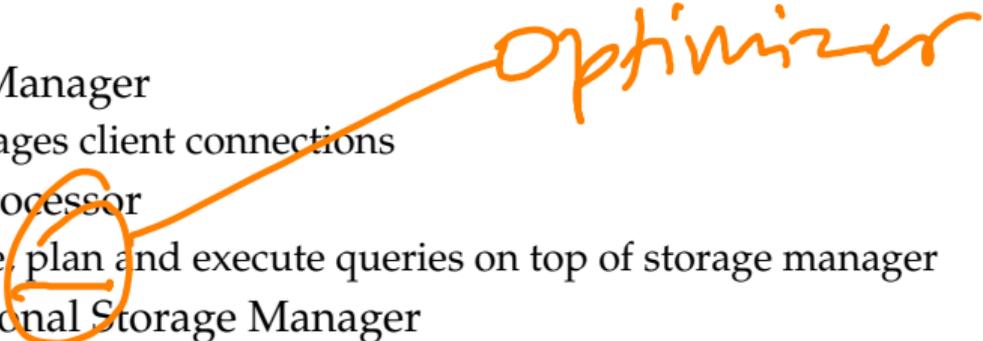
(April 5 -
Project updates /
office hours)

Recap

Anatomy of a Database System [Monologue]



Anatomy of a Database System [Monologue]

- Process Manager
 - ▶ Manages client connections
 - Query Processor
 - ▶ Parse, plan and execute queries on top of storage manager
 - Transactional Storage Manager
 - ▶ Knits together buffer management, concurrency control, logging and recovery
 - Shared Utilities
 - ▶ Manage hardware resources across threads
- optimizer*
- 

Anatomy of a Database System [Monologue]

- Process Manager
 - ▶ Connection Manager + Admission Control
- Query Processor
 - ▶ Query Parser
 - ▶ Query Optimizer (*a.k.a.*, Query Planner)
 - ▶ Query Executor
- Transactional Storage Manager
 - ▶ Lock Manager
 - ▶ Access Methods (*a.k.a.*, Indexes)
 - ▶ Buffer Pool Manager
 - ▶ Log Manager
- Shared Utilities
 - ▶ Memory, Disk, and Networking Manager

Today's Agenda

- Motivation
- Relational Algebra Equivalences
- Nested Queries

Motivation

Query Optimization

2. Low-level
 systems response
 ⇒ more users

- Remember that SQL is declarative.
 - User tells the DBMS what answer they want, not how to get the answer.
- There can be a big difference in performance based on plan is used:
 - Hours → Seconds

1. How to do it?
 Non-trivial

2. Railway
 1. Implementation flexibility
 ⇒ choice to change

IBM System R

PA Schinger

- First implementation of a query optimizer from the 1970s.
 - ▶ People argued that the DBMS could never choose a query plan better than what a human could write.
- Many concepts and design decisions from the System R optimizer are still used today.

Hinks

Commercial
System

Query Optimization

- Approach 1: Heuristics / Rules

- ▶ Rewrite the query to remove stupid / inefficient things.
- ▶ These techniques may need to examine catalog, but they do not need to examine data.

- Approach 2: Cost-based Search

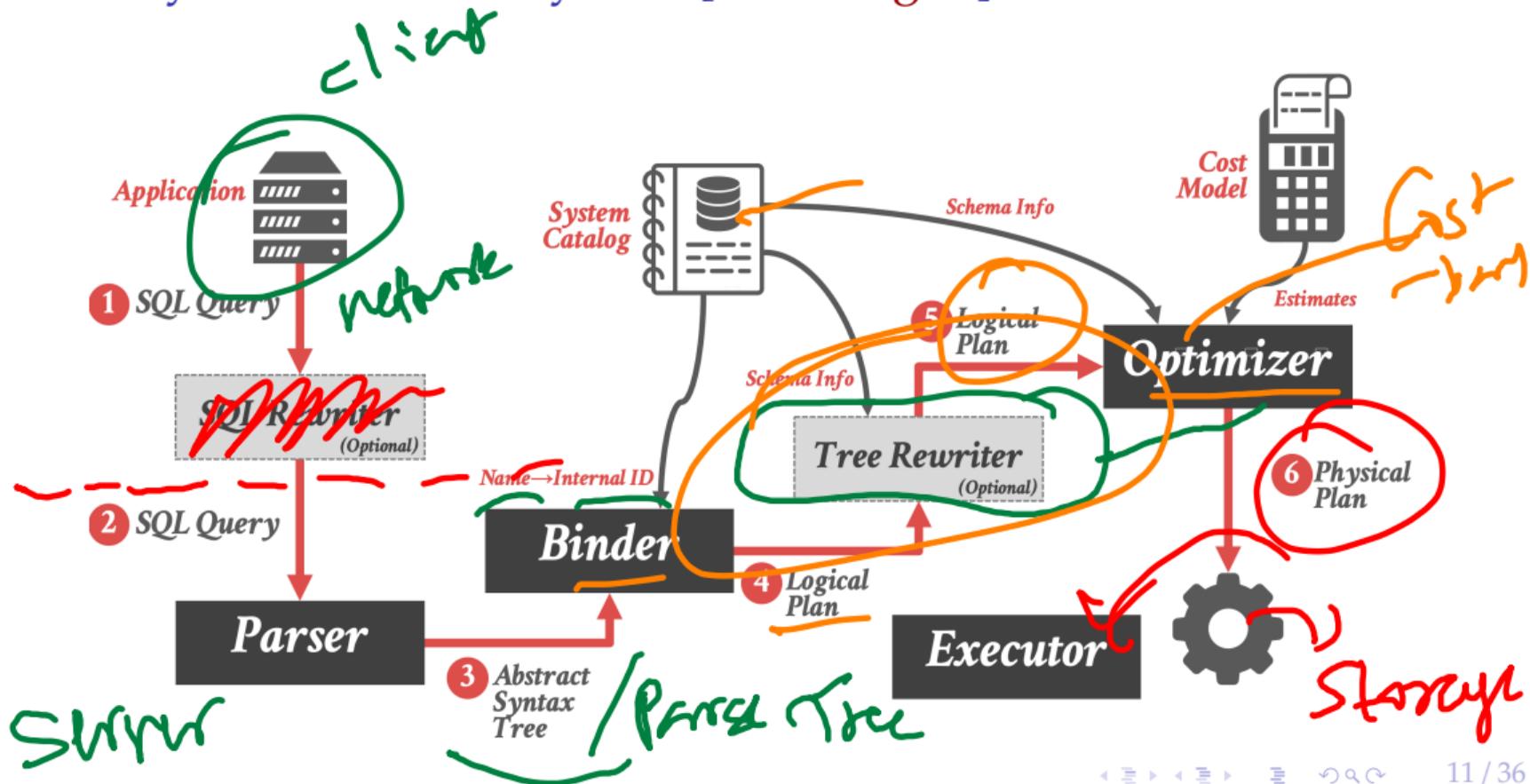
- ▶ Use a model to estimate the cost of executing a plan.
- ▶ Evaluate multiple equivalent plans for a query and pick the one with the lowest cost.

logical cost

data-dependent

meta-data

Anatomy of a Database System [Monologue]



Logical vs. Physical Plans

A ~~B~~ B
Plan Tree

- The optimizer generates a mapping of a logical algebra expression to the optimal equivalent physical algebra expression.
- Physical operators define a specific execution strategy using an access path.
 - ▶ They can depend on the physical format of the data that they process (i.e., sorting, compression).
 - ▶ Not always a 1:1 mapping from logical to physical.

index/scan
scan

Logical
Join
-
SMJ
-
HT
physical

Query Optimization is NP-Hard

Snowflake

- This is the hardest part of building a DBMS.
- If you are good at this, you will get paid well.
- People are starting to look at employing ML to improve the accuracy and efficacy of optimizers.

— size of database
application

— = \$\$\$

— PostgreSQL
Admin (DBAs)
Cloud-native
DBMS

Relational Algebra Equivalences

Relational Algebra Equivalences

- Two relational algebra expressions are equivalent if they generate the same set of tuples.
- The DBMS can identify better query plans without a cost model.
- This is often called query rewriting.

H inputs, $E_1(i) = E_2(i)$

Predicate Pushdown



```

SELECT s.name, e.cid
FROM student AS s, enrolled AS e
WHERE s.sid = e.sid
AND e.grade = 'A'

```

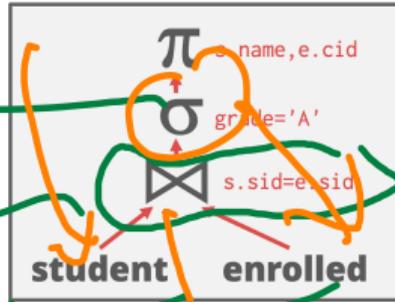
$\pi_{\text{name, cid}}(\sigma_{\text{grade}='A'}(\text{student} \bowtie \text{enrolled}))$



Predicate Pushdown

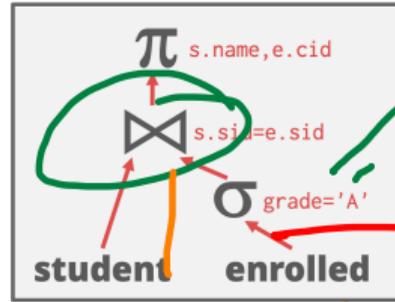
```
SELECT s.name, e.cid
FROM student AS s, enrolled AS e
WHERE s.sid = e.sid
AND e.grade = 'A'
```

$f(s.city)$



Time

100 X 500



100 X 5

Predicate Pushdown

```
SELECT s.name, e.cid
  FROM student AS s, enrolled AS e
 WHERE s.sid = e.sid
    AND e.grade = 'A'
```


$$\pi_{\text{name, cid}}(\sigma_{\text{grade='A'}}(\text{student} \bowtie \text{enrolled}))$$
$$=$$
$$\pi_{\text{name, cid}}(\text{student} \bowtie (\sigma_{\text{grade='A'}}(\text{enrolled})))$$

Relational Algebra Equivalences

• Selections:

- ▶ Perform filters as early as possible.
- ▶ Reorder predicates so that the DBMS applies the most selective one first.
- ▶ Break a complex predicate, and push down $\sigma_{p_1 \wedge p_2 \wedge \dots \wedge p_n}(R) = \sigma_{p_1}(\sigma_{p_2}(\dots \sigma_{p_n}(R)))$

• Simplify a complex predicate

- ▶ $(X=Y \text{ AND } Y=3) \rightarrow X=3 \text{ AND } Y=3$

transitive
closure

$$\begin{aligned}
 & X=Y \ \& \ Y=3 \\
 \rightarrow & X=3 \ \& \ Y=3 \\
 & \underline{X-1=Y} \ \& \ Y=3
 \end{aligned}$$

- @.grade
- @.age

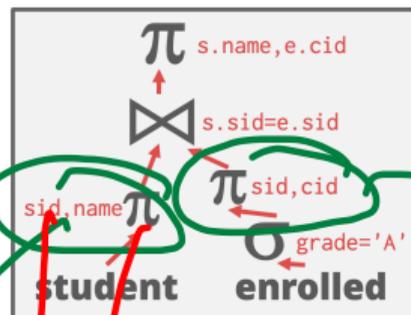
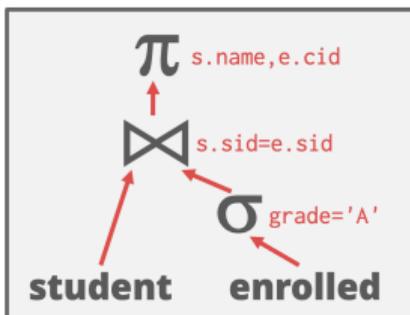
Relational Algebra Equivalences

- Projections:
 - ▶ Perform them early to create smaller tuples and reduce intermediate results (if duplicates are eliminated)
 - ▶ Project out all attributes except the ones requested or required (*e.g.*, joining keys)
- This is not important for a column store

Projection Pushdown

```

SELECT s.name, e.cid
FROM student AS s, enrolled AS e
WHERE s.sid = e.sid
AND e.grade = 'A'
  
```

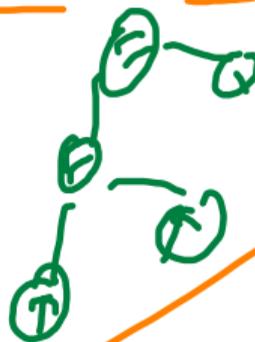


Impossible / Unnecessary Predicates

```
CREATE TABLE A (
  id INT PRIMARY KEY,
  val INT NOT NULL );
```

```
SELECT * FROM A WHERE 1 = 0; --- No Results
SELECT * FROM A WHERE 1 = 1;
```

```
SELECT * FROM A;
```



$$\frac{x - x}{0} = 1$$

Semantics
Expr
Tree

Bryan
Andy

Syntax



CITD

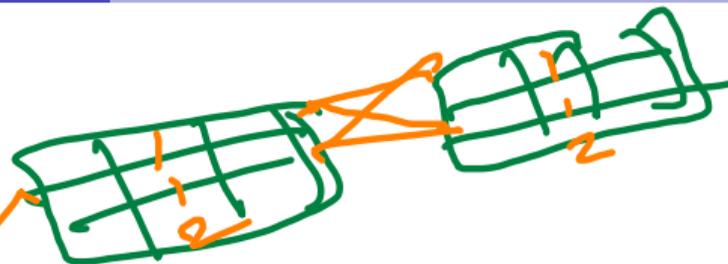
Join Elimination

```
SELECT A1.* FROM A AS A1 JOIN A AS A2
ON A1.id = A2.id;
```

```
SELECT * FROM A;
```

$2 \times 2 = 4$

id \neq primary key



Ignoring Projections

```
SELECT * FROM A AS A1
WHERE EXISTS(SELECT val FROM A AS A2
             WHERE A1.id = A2.id);
```

```
SELECT * FROM A;
```

SQLAlchemy

Object
Relational
Mapper

ORM

Merging Predicates

DML **EVNRIAS** Rule Engine

SELECT * FROM A WHERE val BETWEEN 1 AND 100 OR val BETWEEN 50 AND 150;

SELECT * FROM A WHERE val BETWEEN 1 AND 150;

p2 **Theorem P002 Rule**

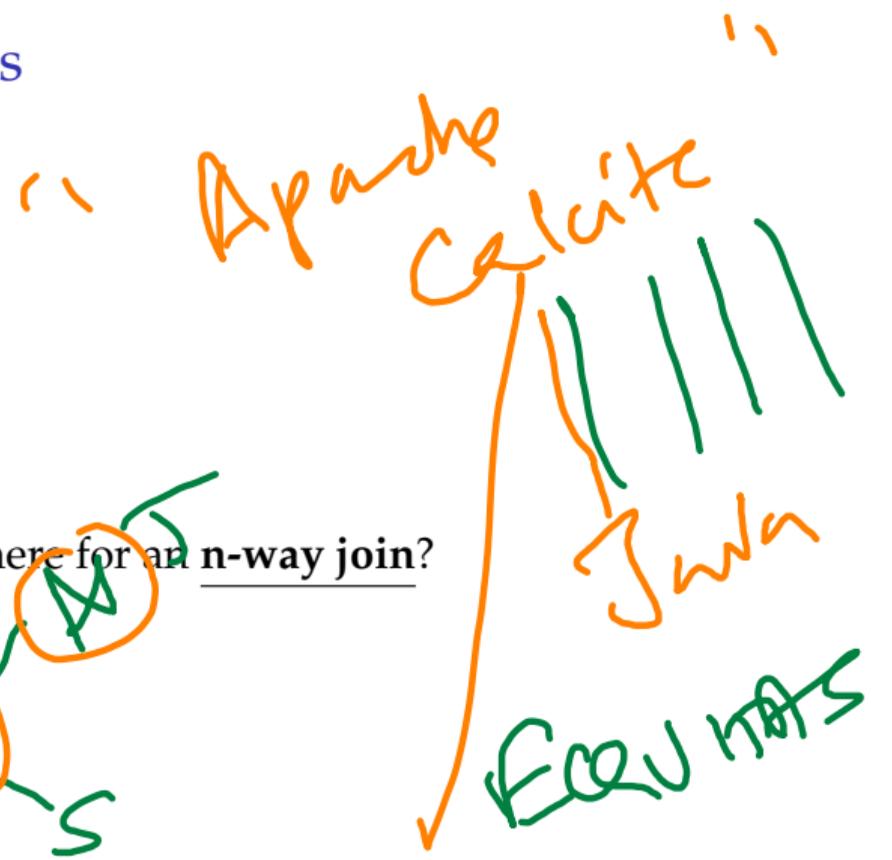
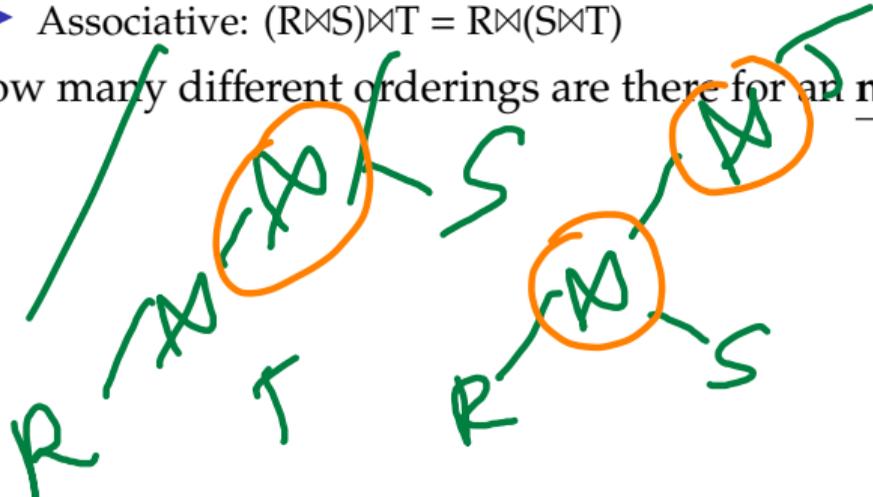


Relational Algebra Equivalences

- Joins:

- ▶ Commutative: $R \bowtie S = S \bowtie R$
- ▶ Associative: $(R \bowtie S) \bowtie T = R \bowtie (S \bowtie T)$

- How many different orderings are there for an n-way join?



Relational Algebra Equivalences

- How many different orderings are there for an n-way join?
- Catalan number: $\approx 4_n$
 - ▶ Exhaustive enumeration will be too slow.
 - ▶ We'll later look at how an optimizer prunes the search space.

Optimization time

Exhaustive time

Nested Sub-Queries

Nested Sub-Queries

- The DBMS treats nested sub-queries in the WHERE clause as functions that take parameters and return a single value or set of values.
- Two Approaches:
 - ▶ Rewrite to de-correlate and/or flatten them
 - ▶ Decompose nested query and store result to temporary table

Nested Sub-Queries: Rewrite

```
SELECT name FROM sailors AS S
WHERE EXISTS (
  SELECT * FROM reserves AS R
  WHERE S.sid = R.sid
  AND R.day = '2018-10-15'
)
```

outer

Wishywash
COW BOKE

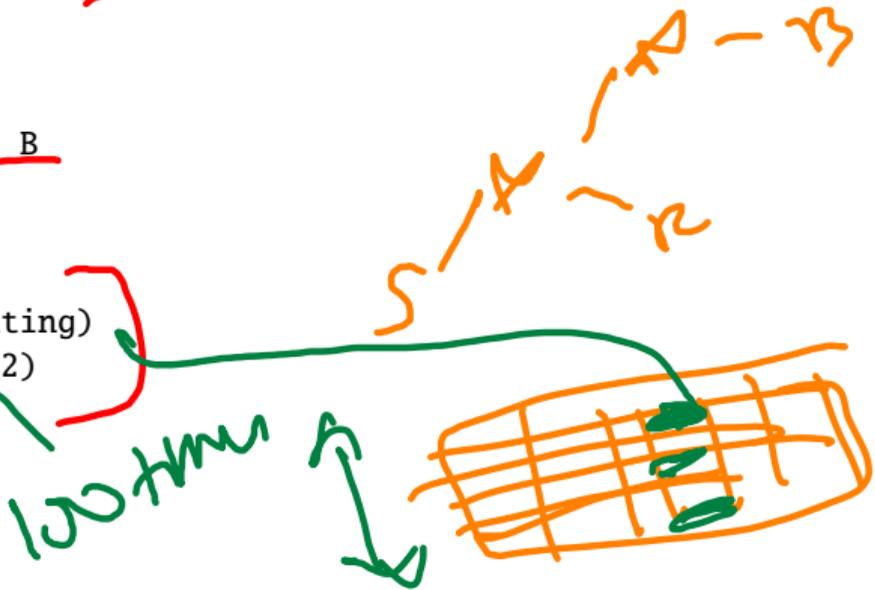
inner

```
SELECT name
FROM sailors AS S, reserves AS R
WHERE S.sid = R.sid
AND R.day = '2018-10-15'
```

Nested Sub-Queries: Decompose

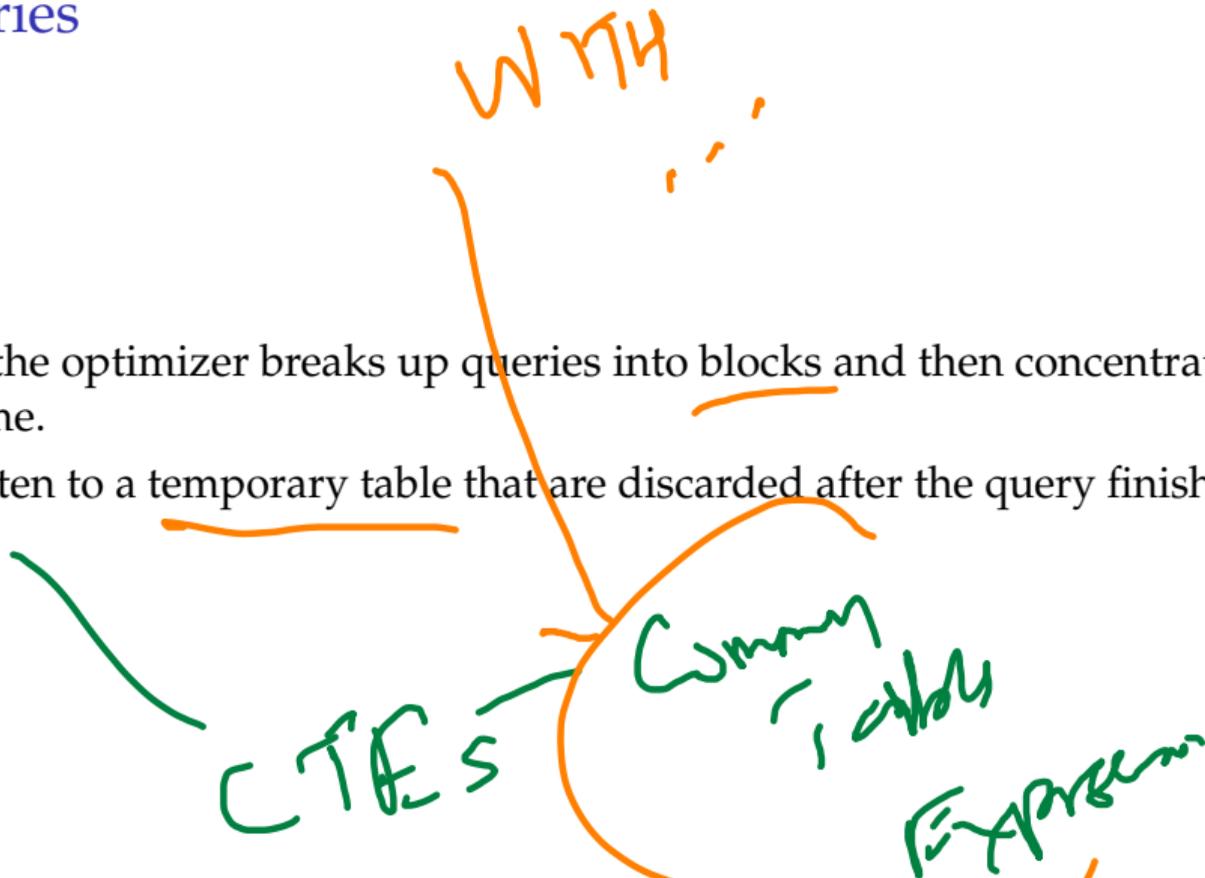
- For each sailor with the highest rating (over all sailors) and at least two reservations for red boats, find the sailor id and the earliest date on which the sailor has a reservation for a red boat. /

```
SELECT S.sid, MIN(R.day)
FROM sailors S, reserves R, boats B
WHERE S.sid = R.sid
AND R.bid = B.bid
AND B.color = 'red'
AND S.rating = (SELECT MAX(S2.rating)
                FROM sailors S2)
GROUP BY S.sid
HAVING COUNT(*) > 1
```



Decomposing Queries

- For harder queries, the optimizer breaks up queries into blocks and then concentrates on one block at a time.
- Sub-queries are written to a temporary table that are discarded after the query finishes.



Decomposing Queries

```
SELECT S.sid, MIN(R.day) --- Outer Block
FROM sailors S, reserves R, boats B
WHERE S.sid = R.sid
      AND R.bid = B.bid
      AND B.color = 'red'
      AND S.rating = () --- Result of Nested Query
GROUP BY S.sid
HAVING COUNT(*) > 1
```

500

```
SELECT MAX(rating) FROM sailors
```

Conclusion

Query Optimization

- We can use static rules and heuristics to optimize a query plan without needing to understand the contents of the database.
- Filter as early as possible.

Expert System

R1
R2
...
R100



Next Class

- Cost-based Query Optimization