# COMP 430 Intro. to Database Systems

Query optimization

### Questions we want to address

- How to understand queries' performance?
- What queries can system optimize for us?
  - How can we tell?
  - How does it optimize?
- What queries do we need to optimize ourselves?
  - How can we tell?
  - How do we optimize?

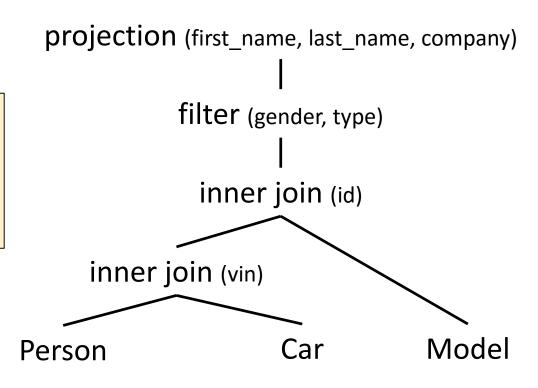
Need to understand some basic optimization strategies, but not the details.

### Example query

```
SELECT Person.first_name, Person.last_name, Model.company FROM Person
INNER JOIN Car ON Person.car_vin = Car.vin
INNER JOIN Company ON Car.model_id = Model.id
WHERE Person.gender = 'F' AND Car.type IN ('sedan', 'coupe');
```

### A simple query tree

SELECT Person.first\_name, Person.last\_name, Model.company
FROM Person
INNER JOIN Car ON Person.car\_vin = Car.vin
INNER JOIN Company ON Car.model\_id = Model.id
WHERE Person.gender = 'F' AND Car.type IN ('sedan', 'coupe');

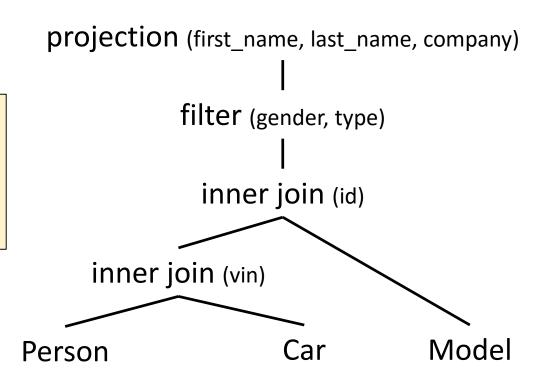


Tree form of *relational algebra* expression

### Activity – Generate equivalent query trees

SELECT Person.first\_name, Person.last\_name, Model.company
FROM Person
INNER JOIN Car ON Person.car\_vin = Car.vin
INNER JOIN Company ON Car.model\_id = Model.id
WHERE Person.gender = 'F' AND Car.type IN ('sedan', 'coupe');

Suggestions?



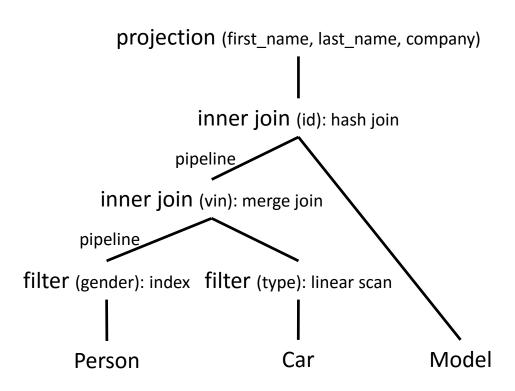
### Generate execution plans

projection (first\_name, last\_name, company) Notate with implementation inner join (id): hash join methods. Possibly multiple plans per tree. pipeline inner join (vin): merge join pipeline filter (gender): index filter (type): linear scan Model Person Car

### Analyze costs of each plan

- Algorithm cost for each node
- Data amount
  - #tuples
  - Total #bytes or #disk blocks
  - Selectivity of filters
- Data access
  - Method index, linear scan, binary search
  - Source memory, disk

Data-dependent – determined at run time. Use & cache lowest cost plan.



### Trees $\rightarrow$ Plans $\rightarrow$ Costs

#### Steps integrated

- Can be too many trees/plans to generate all of them.
- Cost heuristics guide tree/plan generation.

### Simple example of cost analysis

```
SELECT *
FROM Person
WHERE name IN ('Greiner', 'Halverhout') AND
zipcode = 77055 AND
birthdate > '19670212';
```

Test	CPU cost/record
name IN	100ns
zipcode =	1ns
birthdate >	1000ns

400,000,000 records

### Simple example of cost analysis: Test order

```
For each Person record:

If name IN ...

Then if zipcode = ...

Then if birthdate > ...

Then output record
```

How many possible orders? How to calculate cost for each?

## Simple example of cost analysis: Test order

400,000,000 records

Test	CPU cost/record	Selectivity	Cost for this test	simplified
name IN	100ns	0.00001	400,000,000 × 100ns	40,000,000,000ns
zipcode =	1ns	0.0001	400,000,000 × 0.00001 × 1ns	4,000ns
birthdate >	1000ns	0.3	400,000,000 × 0.00001 × 0.0001 × 1000ns	400ns
				40,000,004,400ns

Test	CPU cost/record	Selectivity	Cost for this test	simplified
zipcode =	1ns	0.0001	400,000,000 × 1ns	400,000,000ns
name IN	100ns	0.00001	400,000,000 × 0.0001 × 100ns	400,000ns
birthdate >	1000ns	0.3	400,000,000 × 0.00001 × 0.0001 × 1000ns	400ns
				400,400,400ns

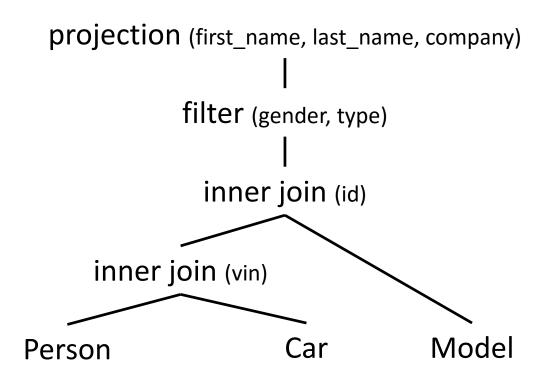
### Simple example of cost analysis: Test order

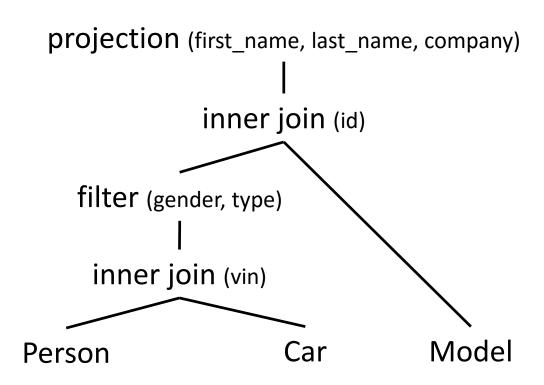
How to search through all the plans efficiently?

**Best:** in decreasing order by  $\frac{1 - \text{Selectivity(predicate})}{\text{Cost(predicate)}}$ , i.e., the fraction of data eliminated per unit time.

# Transforming query trees

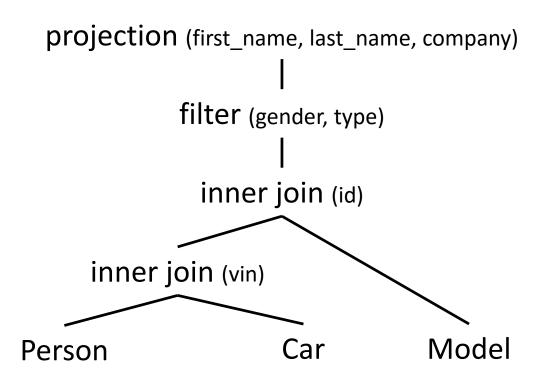
### Relational algebra equivalences: example

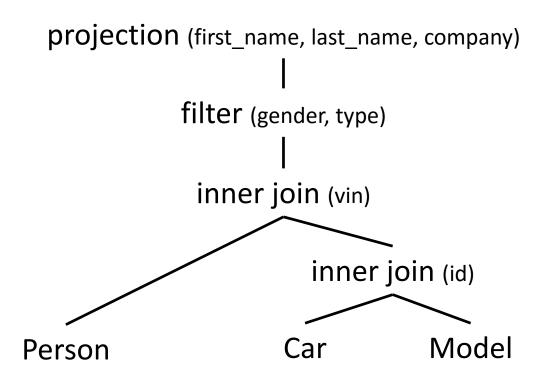




 $filter(inner join(A,B)) = filter_{AB}(inner join(filter_{A}(A),filter_{B}(B))$ 

### Relational algebra equivalences: example





Inner joins are associative & commutative.

### Too many trees to enumerate all

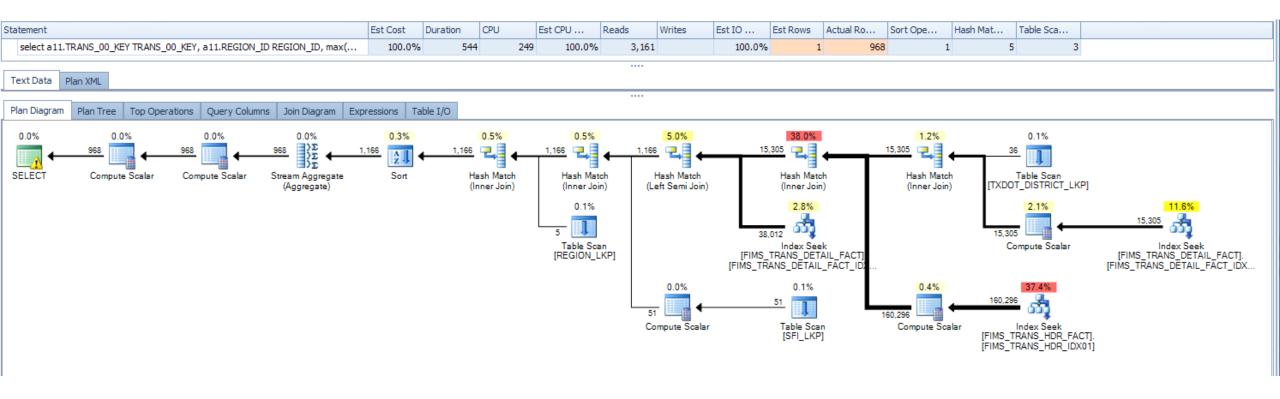
Example: inner join of n tables with commutativity & associativity → how many trees?

 $n! 2^n$ 

# Implementation Strategies

### Viewing & understanding execution plan

Helpful to be able to understand plan and any bottlenecks.



### Scan implementations

- Indexing previously discussed
- Binary search if sorted on search key
- Linear search

## What to do with each operation's results?

#### Materialization

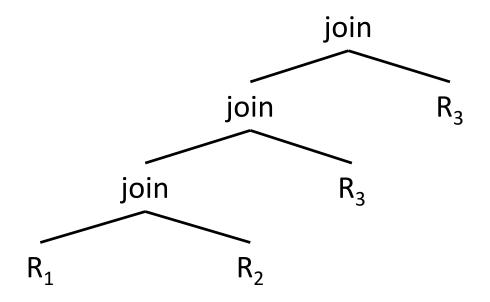
- Data saved to disk (or memory, if small enough)
- Disk access
- + Can reuse data if needed multiple times

#### Pipelining

- Consumer uses data as producer generates it
- Use buffer to allow consumer to be faster than producer
- + No disk access

## Left-deep trees have one long pipeline

Common to consider only *left-deep* trees of joins.



### Join implementations

Join(R, S) on R.A = S.A Let T(R) = #tuples in R, B(R) = #blocks in R

- Three main strategies
  - Nested joins + 2 variants
  - Sort-merge joins + 1 variant
  - Hash joins

### Nested loop join

#### For r in R:

For s in S:

If r.a == s.a, yield (r,s)

$$Cost = B(R) + T(R)B(S) + B(result)$$

#### For s in S:

For r in R:

If r.a == s.a, yield (r,s)

$$Cost = B(S) + T(S)B(R) + B(result)$$

### Block nested loop join – I/O aware

Let B+1 = #blocks in memory

For each group of B-1 blocks br of R:

For block bs of S:

For r in br:

For s in bs:

If r.a == s.a, yield (r,s)

Cost = 
$$B(R) + \frac{B(R)}{B-1}B(S) + B(\text{result})$$

Choose R = smaller relation

### Index nested loop join

Assumes S indexed on A.

For r in R:

If r.a == index(s,a), yield (r,s)

$$Cost = B(R) + T(R)C + B(result)$$

C = I/O to access all distinct values in index. Not constant, but typically < 10.

### Sort-merge join

- 1. Sort R and S each on A.
  - Use external merge sort (disk-based).
  - But, R and/or S might already be sorted on A.
- 2. "Merge" R and S.
  - Linear scan R.
  - Linear scan S, backing up on duplicates.

R: (1,a) (2,b) (2,c) (3,d)

S: (1,v)(2,w)(2,x)(3,y)(4,z)

Out: (1,a,v) (2,b,w) (2,b,x) (2,c,w) (2,c,x), (3,d,y)

$$R\left(\log_B \frac{R}{B+1}\right)$$

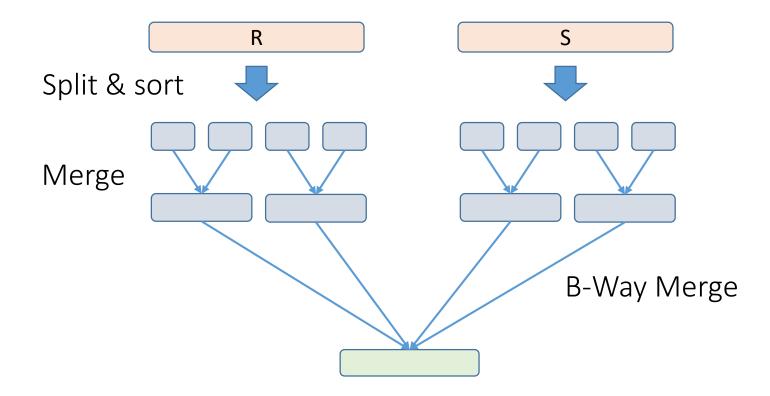
$$S\left(\log_B \frac{S}{B+1}\right)$$

$$B(R) + B(S)$$
 ...  $B(R)B(S)$ 

Not too sensitive to skew. Leaves data sorted!

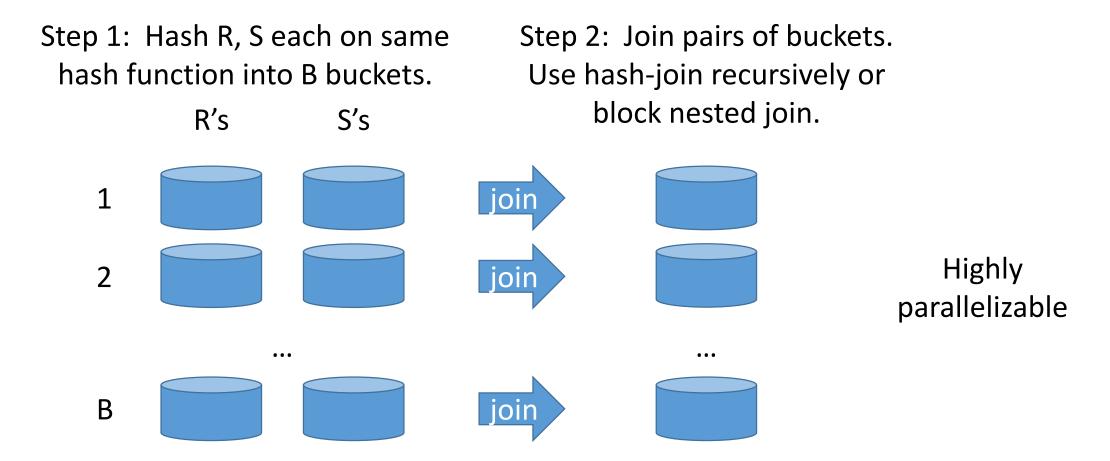
## Sort-merge join – I/O aware

Integrate the mergesort's merging with the "merge" of R,S



### Hash-join

Use hashing to decompose one large join into many small joins.



## Size estimation

Costs depend on amount of data

### Estimating sizes

- Size of each relation
- Selectivity of each join & filter condition

- How calculated?
  - Gather statistics
  - Use statistics to estimate selectivity

### Size statistics per relation R

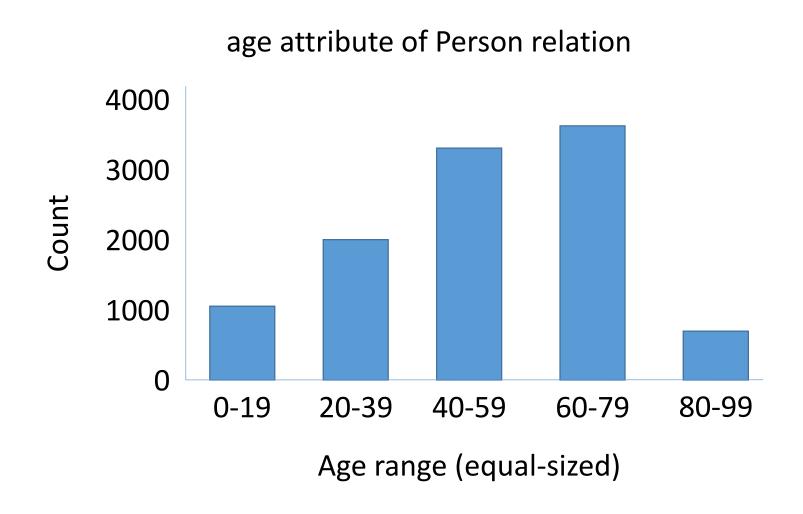
- tuples(R): #tuples
- *length*(*R*): #bytes per tuple
- bfactor(R): #tuples that fit into a block (blocking factor)
- blocks(R): #blocks if no fragmentation, =  $\left[\frac{tuples(R)}{bfactor(R)}\right]$

### Value statistics for attribute A

- *values*(*A*): #distinct values
- minval(A): minimum value
- maxval(A): maximum value

• Distribution of values for attribute A – estimated by histogram

### Equi-width histogram example



### Histograms summary

Various other types of histograms

- High-level goals
  - Time- & Space-efficient to compute
  - Useful in estimating selectivity of equality and range conditions

### Managing statistics

- Accuracy vs. resources
  - Update frequency
  - Build histograms for which attributes, with how much accuracy
- Automatically by DBMS
- Some manual control
  - SQL Server: CREATE STATISTICS, UPDATE STATISTICS
- Can view statistics
  - SQL Server: DBCC SHOW\_STATISTICS

### Size estimates for various operations

- CrossJoin(R,S)  $tuples(R) \times tuples(S)$
- InnerJoin(R,S) tuples(R)
- OuterJoin(R,S) tuples(R) + tuples(S)

Assuming join on R's FK & S's PK

- $R \cup S$  tuples(R) + tuples(S)
- $R \cap S$  min(tuples(R), tuples(S))
- R-S tuples(R)

- WHERE R.A = v:
  - Without histogram:  $s = \frac{1}{values(R.A)}$
  - With equal-width histogram:  $s = \frac{count(range\_containing\_A,R)}{tuples(R) \times range\_width}$

- WHERE R.A = S.B.
  - Without histograms:  $s = \min\left(\frac{1}{values(R.A)}, \frac{1}{values(S.B)}\right) = \frac{1}{\max(values(R.A), values(S.B))}$
  - With histograms: Generalize the previous.

R.A values(R.A)=3	S.B values(S.B)=2	Match?
1	1	Yes
2	2	Yes
3	1	No
1	2	No
2	1	No
3	2	No

I.e., expect 1/3 to match.

#### • WHERE R.A ≥ v:

- Without statistics:  $s=\frac{1}{2}$  Without histogram:  $s=\begin{cases} 0 & \text{if } v>maxval(R.A) \\ \frac{maxval(R.A)-v}{maxval(R.A)-minval(R.A)} & \text{otherwise} \end{cases}$
- With equal-width histogram: Generalize the previous.

• Conjunction: 
$$s = \prod s_i$$

• Disjunction: 
$$s = 1 - \prod (1 - s_i)$$

• Negation: 
$$s = 1 - s'$$

# User-level Query Optimization

So, what should **you** the SQL programmer do?

# Just a bit of terminology

Query optimization – what the system does

Query tuning – what the programmer does

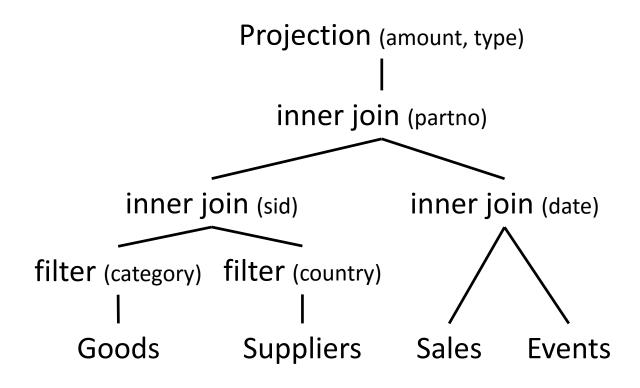
# Understanding current performance

What is the bottleneck? Why?

- Query plan
- Statistics DBCC SHOW\_STATISTICS
- Estimate statistics yourself
- Experiment with timings

# How would you investigate slow query?

SELECT Sales.amount, Events.type
FROM Sales, Events, Goods, Suppliers
WHERE Sales.date = Events.date AND
Sales.partno = Goods.partno AND
Suppliers.sid = Goods.sid AND
Goods.category = 'engine' AND
Suppliers.country = 'US'



# Technique summary

- Improve system resources
- Improve query optimization
- Don't get in optimizer's way
- Reduce computation
- Reduce data usage
- Denormalization next topic
- Eliminating constraints & triggers generally disrecommended

Not a prioritized list!

#### Improving system resources

Not always feasible, but can improve EVERY query.

- Faster CPU
- More physical memory
  - Reduce need for I/O
- Faster I/O
  - SSDs instead of disks
  - Distribute over multiple drives to increase bandwidth

#### Improving query optimization

Help system optimize better.

- Use simpler synthetic keys
- Create indices CREATE INDEX
- Manage index growth FILL FACTOR
- Create specialized statistics CREATE/UPDATE STATISTICS
- Partition or data stripe table to improve I/O
- Check whether benefits outweigh overheads.

# Overriding the query optimizer

Sometimes, the query optimizer gets it wrong.

- Force it to use index: SELECT ... FROM ... WITH INDEX ...
- Force join order

However, the query optimizer is usually smarter than you.

Don't get in optimizer's way

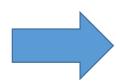
### Use WHERE expressions on the raw attribute

SELECT ...

**FROM Account** 

WHERE YEAR(created\_on) = 2016 AND MONTH(created\_on) = 1;

Blocks usage of both index & statistics.



SELECT ...

**FROM Account** 

WHERE created on BETWEEN '1/1/2016' AND '2/1/2016';

Other functions commonly used like this: ISNULL() and implicit type conversions.

# Don't loop over SQL calls

for each person in person\_list:
INSERT INTO Person VALUES (this person's data)

Use one INSERT INTO with a list of values.

Such examples generally stem from not understanding SQL's capabilities.

Use one SELECT with a join, grouping on product\_id.

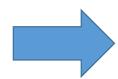
# Don't use cursors unnecessarily

Many newbie examples just duplicate SELECT's features.

# Reduce computation

#### Avoid constant-valued subqueries

SELECT first\_name, last\_name
FROM Person
WHERE age = (SELECT Max(age)
FROM Person);

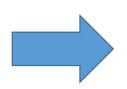


SELECT @oldest\_age = MAX(age) FROM Person;

SELECT first\_name, last\_name FROM Person WHERE age = @oldest\_age;

# Avoid *correlated* subqueries when possible

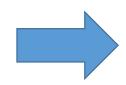
```
SELECT name,
city,
(SELECT company_name FROM Company WHERE id = Customer.company_id)
FROM Customer;
```



```
SELECT name,
city,
company_name
FROM Customer
INNER JOIN Company ON Company.id = Customer.company_id;
```

### Avoid *correlated* subqueries when possible

```
SELECT id, first_name, last_name, salary
FROM Employee e
WHERE EXISTS (SELECT 1
FROM Orders o
WHERE e.id = o.sales_rep_id AND
o.customer_id = 123);
```

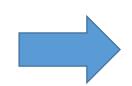


```
SELECT id, first_name, last_name, salary
FROM Employee e
WHERE id IN (SELECT sales_rep_id
FROM Orders
WHERE o.customer_id = 123);
```

#### Avoid DISTINCT when it's unnecessary

Find departments with employees

SELECT DISTINCT Dept.id, Dept.name FROM Dept, Employee WHERE Dept.id = Employee.dept\_id;



SELECT id, name
FROM Dept
WHERE id IN (SELECT dept\_id
FROM Employee);

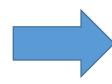
A poor variation:

SELECT id, name
FROM Dept
WHERE id IN (SELECT DISTINCT dept\_id
FROM Employee);

Similarly, avoid UNION when UNION ALL is sufficient.

#### Don't COUNT to check for existence

```
SELECT id, name
FROM Dept
WHERE (SELECT COUNT(*)
FROM Employee
WHERE Dept.id = Employee.dept_id)
> 0;
```



SELECT id, name
FROM Dept
WHERE id IN (SELECT dept\_id
FROM Employee);

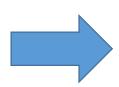
#### Use CASE to avoid multiple passes

SELECT COUNT (\*)

FROM Employee WHERE salary < 20000;

SELECT COUNT (\*)
FROM Employee
WHERE salary BETWEEN 20000
AND 50000;

SELECT COUNT (\*)
FROM Employee
WHERE salary > 50000;



SELECT COUNT (CASE WHEN salary < 20000
THEN 1 ELSE null END),
COUNT (CASE WHEN salary BETWEEN 2000
AND 50000
THEN 1 ELSE null END),
COUNT (CASE WHEN salary > 50000
THEN 1 ELSE null END);

# Don't ORDER BY unnecessarily

• Don't sort in nested/helper queries. Only sort your final results.

Watch for this when using a pre-existing query as a helper.

# Reducing data usage

#### Don't select too much data

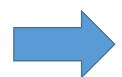
SELECT first\_name, last\_name, city FROM Person WHERE city = 'Houston'; SELECT first\_name, last\_name, city FROM Person;

... and then application code filters the data.

SELECT \*
FROM Person
WHERE city = 'Houston';

#### SELECT less data in subquery

```
SELECT id, first_name, last_name, salary
FROM Employee e
WHERE dept_id = 456 AND
id IN (SELECT sales_rep_id FROM Orders);
```



Probably sorts and/or indexes subquery.

In this case, a correlated subquery can be better.

- Subquery results in less data.
- Subquery can use indices on both attributes.

```
SELECT id, first_name, last_name, salary
FROM Employee e
WHERE dept_id = 456 AND
EXISTS (SELECT 1
FROM Orders
WHERE e.id = o.sales_rep_id);
```

# Don't update data with identical data

```
UPDATE Person

SET active = False

WHERE ... AND /* no recent transactions, website visits, or customer calls */

active = True;
```

Avoids many writes to Person. Reduces any logging of writes.