# Principles of Distributed and Parallel Database Systems: Part 2 Total Cost of a Query Execution Plan



# **Objectives**



Objective

Explain the reliability & fault tolerance concepts in distributed data processing systems

#### **Total Cost**

Total cost = CPU cost + I/O cost + communication cost

CPU cost= unit instruction cost \* no. of instructions

I/O cost= unit disk I/O cost \* no. of disk I/Os

communication cost = message initiation +
transmission

## **Response Time**

Response time = CPU time + I/O time + communication time

CPU time= unit instruction time \* no. of sequential instructions

I/O time= unit I/O time\*no. of sequential I/Os

communication time= unit msg initiation time\*no. of sequential msg

+ unit transmission time\*no. of sequential bytes

# **Optimization Statistics**

Primary cost factor:size of intermediate relations

Need to estimate their sizes

Make them precise

⇒ more costly to

maintain

Simplifying assumption: uniform distribution of attribute values in a relation

#### **Statistics**

For each relation  $R[A_1, A_2, ..., A_n]$  fragmented as  $R_1, ..., R_r$ 

- length of each attribute:length(A<sub>i</sub>)
- The number of distinct values for each attribute in each fragment:  $card(\Pi_{Ai}R_i)$
- maximum and minimum values in the domain of each attribute: min(A<sub>i</sub>), max(A<sub>i</sub>)
- the cardinalities of each domain:card(dom[A<sub>i</sub>])

The cardinalities of each fragment: $card(R_j)$  Selectivity factor (SF) of each operation for relations

- For joins

$$SF \bowtie (R,S) = \frac{card(R \bowtie S)}{card(R) * card(S)}$$

#### Selection

- size(R) =card(R)
   ×length(R)
- $card(\sigma_F(R)) = SF_{\sigma}(F)$ ×card(R)

$$SF_{\sigma}(A=value) = ?$$

$$SF_{\sigma}(A>value) = ?$$

$$SF_{\sigma}(A < value) =$$

length of each attribute:  $length(A_i)$ 

number of distinct values for each attribute in each fragment:  $card(\Pi_{Ai}R_i)$ 

maximum and minimum values in the domain of each attribute: $min(A_i)$ , $max(A_i)$ 

the cardinalities of each domain: card(dom[A<sub>i</sub>])

$$size(R) = card(R) \times length(R)$$
  
 $card(\sigma_F(R)) = SF_{\sigma}(F) \times card(R)$ 

#### where

$$SF_{\sigma}(A = value) = \frac{1}{card(\prod_{A}(R))}$$

$$SF_{\sigma}(A > value) = \frac{max(A) - value}{max(A) - min(A)}$$

$$SF_{\sigma}(A < value) = \frac{value - min(A)}{max(A) - min(A)}$$

$$SF_{\sigma}(p(A_i) \land p(A_i)) = SF_{\sigma}(p(A_i)) \times SF_{\sigma}(p(A_i))$$

$$SF_{\sigma}(p(A_i) \vee p(A_j)) = SF_{\sigma}(p(A_i)) + SF_{\sigma}(p(A_j)) - (SF_{\sigma}(p(A_i)) \times SF_{\sigma}(p(A_j)))$$

#### Projection

 $- card(\Pi_A(R)) = card(R)$ 

#### Cartesian Product

 $- card(R \times S) = ?$ 

#### Union

- upper bound:card(R∪S) =?
- lower
  bound:card(R∪S) =?

#### Set Difference

- upper bound:?
- lower bound:?

#### Projection

-  $card(\Pi_{A}(R)) = card(R)$ 

#### Cartesian Product

-  $card(R \times S) = card(R)^* card(S)$ 

#### Union

- upper bound:card(R∪S)=card(R) +card(S)
- lower bound:card(R∪S)=max{card(R),card(S)}

#### Set Difference

- upper bound:card(R-S)=card(R)
- lower bound: 0

### **Intermediate Relation Size: Join**

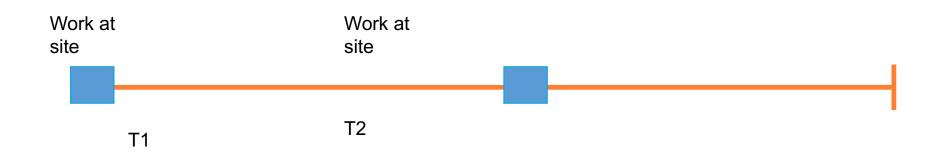
# Special case: Ais a key of Rand Bis a foreign key of S

- card( $R\bowtie_{A=B}S$ ) = card(S)

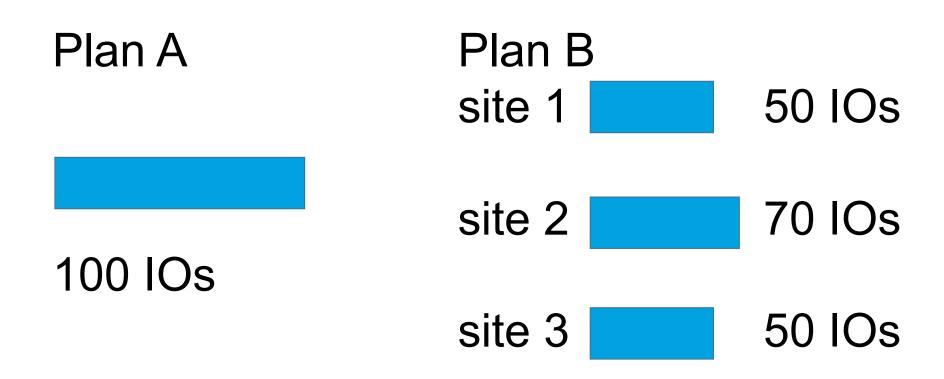
#### More general:

- card(R⋈S) =SF<sub>∞</sub>\*card(R)×card(S)

# **IOs May Not be the Best Metric**



# **IOs May Not be the Best Metric**

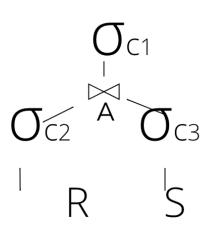


# **Query Separation**

Separate query into 2 or more steps

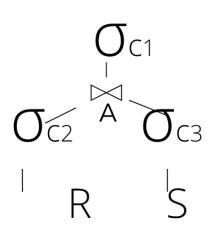
Optimize each step independently

# **Example: Simple Queries**



- 1. Compute R'=  $\Pi$ A[ $\sigma$ c2R] S'=  $\Pi$ A[ $\sigma$ c3S]
- 2. 2. Compute J = R' S'
- 3. Compute  $J = R' \bowtie S'$

# **Example: Simple Queries**



1. Compute R'=  $\Pi$ A[ $\sigma$ c2R]

$$S' = \Pi A[\sigma c3S]$$

- 2. Compute  $J = R' \bowtie S'$
- 3. Compute

Ans =  $\sigma c1\{[J \bowtie \sigma c2R] \ [J \bowtie \sigma c3S]\}$ 

# **Simple Query**

- Relations have a single attribute
- Output has a single attribute

$$J \leftarrow R' \bowtie S'$$

#### Decompose query into

- Local processing
- Simple query (or queries)
- Final processing

Optimize simple query

# Philosophy

- Hard part is distributed join
- Do this part with only keys; get rest of data later
- Simpler to optimize simple queries