# CPS216: Data-intensive Computing Systems Query Optimization (Costbased optimization)

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### **Query Optimization Problem**

Pick the best plan from the space of physical plans

### **Cost-Based Optimization**

- Prune the space of plans using heuristics
- Estimate cost for remaining plans

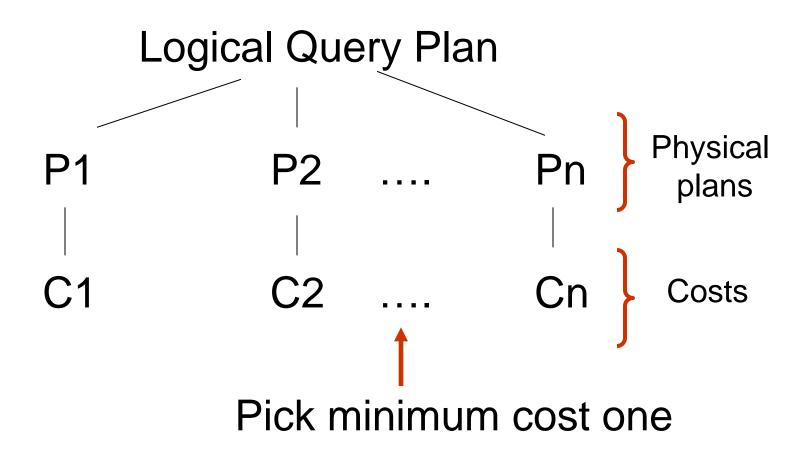
   Be smart about how you iterate through plans
- Pick the plan with least cost

Focus on queries with joins

### Heuristics for pruning plan space

- Predicates as early as possible
- Avoid plans with cross products
- Only left-deep join trees

#### **Physical Plan Selection**



### **Review of Notation**

- T (R) : Number of tuples in R
- B (R) : Number of blocks in R

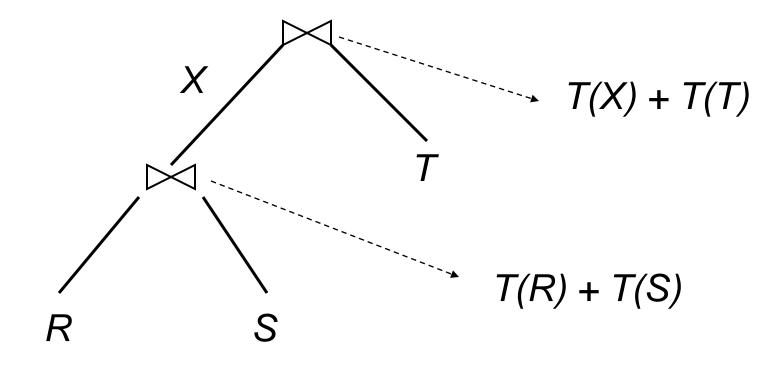
### Simple Cost Model

Cost (R  $\bowtie$  S) = T(R) + T(S)

All other operators have 0 cost

Note: The simple cost model used for illustration only

### **Cost Model Example**

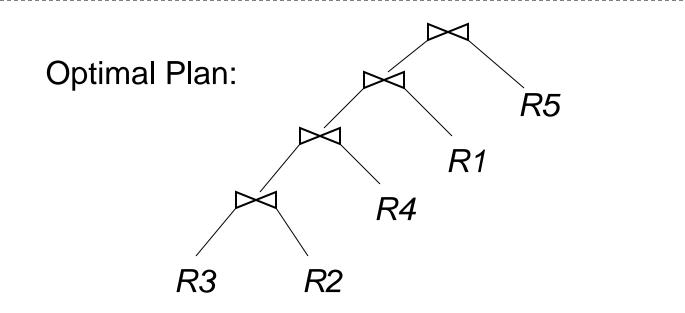


Total Cost: T(R) + T(S) + T(T) + T(X)

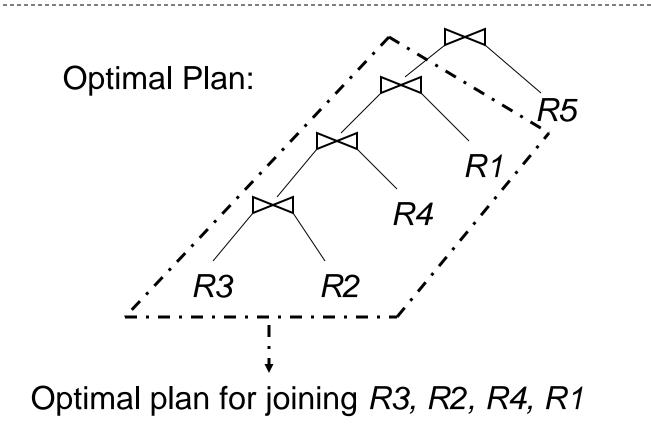
- Dynamic Programming based
- Dynamic Programming:
  - General algorithmic paradigm
  - Exploits "principle of optimality"
  - Useful reading:
    - Chapter 16, Introduction to Algorithms, Cormen, Leiserson, Rivest

Optimal for "whole" made up from optimal for "parts"

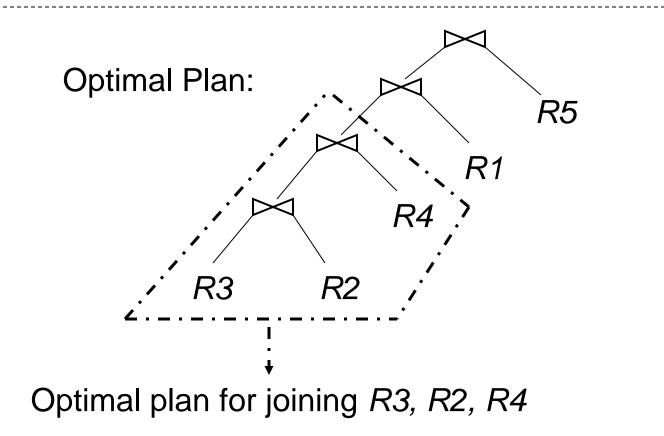
#### Query: $R1 \bowtie R2 \bowtie R3 \bowtie R4 \bowtie R5$



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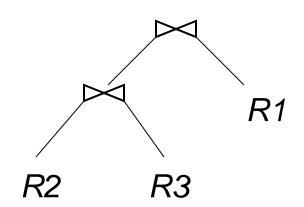


#### Query: $R1 \bowtie R2 \bowtie R3 \bowtie R4 \bowtie R5$

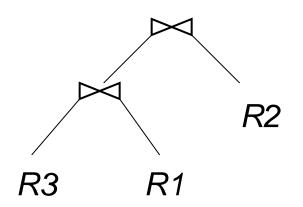


## **Exploiting Principle of Optimality**



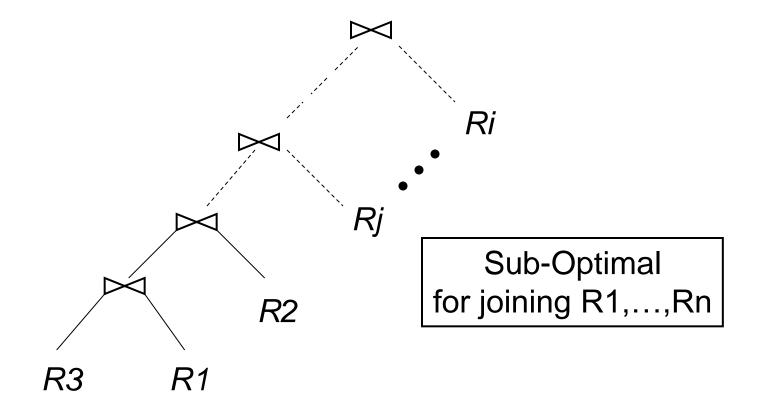


Optimal for joining *R1, R2, R3* 

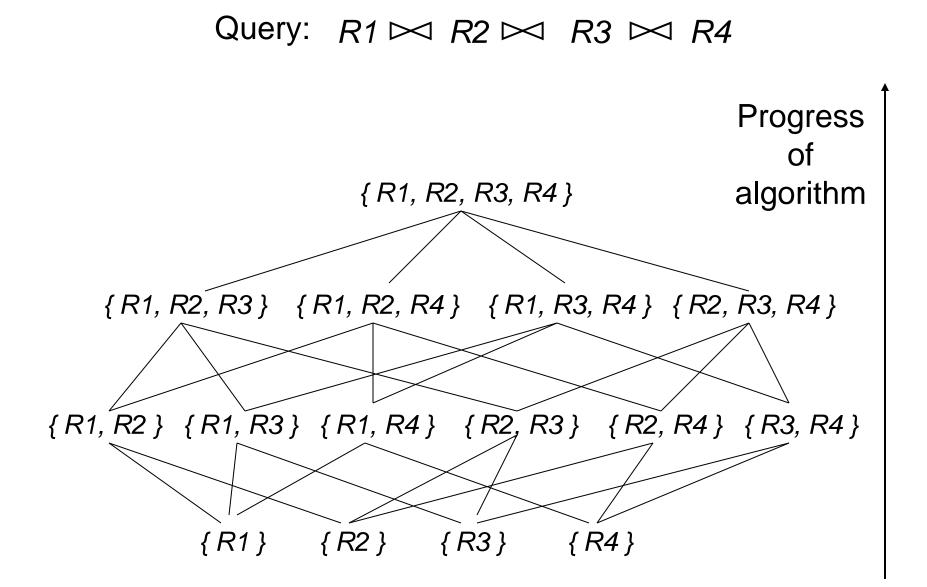


Sub-Optimal for joining *R1, R2, R3* 

## **Exploiting Principle of Optimality**



A sub-optimal sub-plan cannot lead to an optimal plan



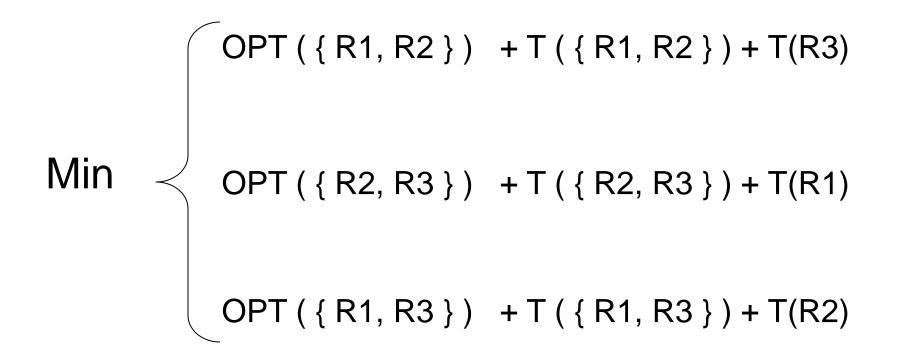
### Notation

# OPT ({ *R1, R2, R3*}): Cost of optimal plan to join *R1,R2,R3*

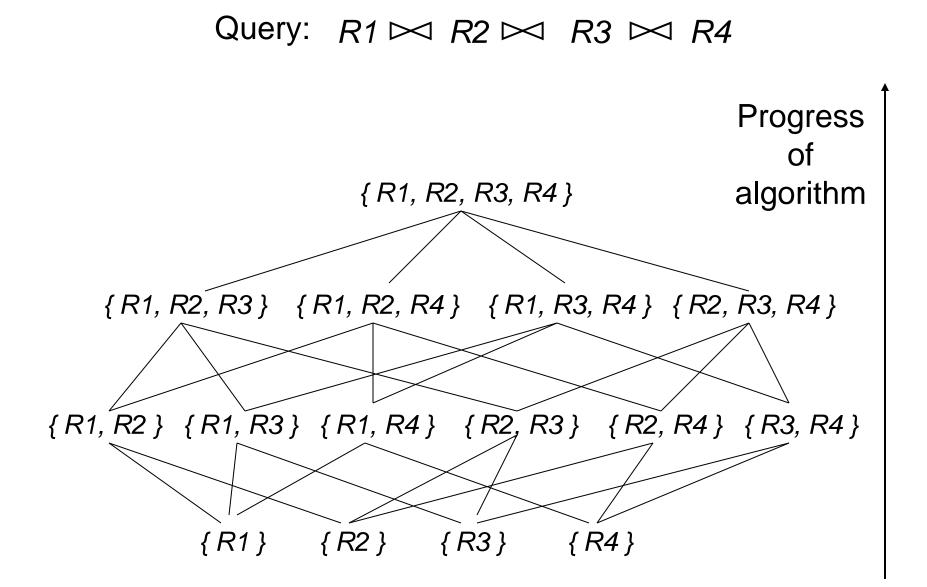
#### T ( { *R1, R2, R3* } ):

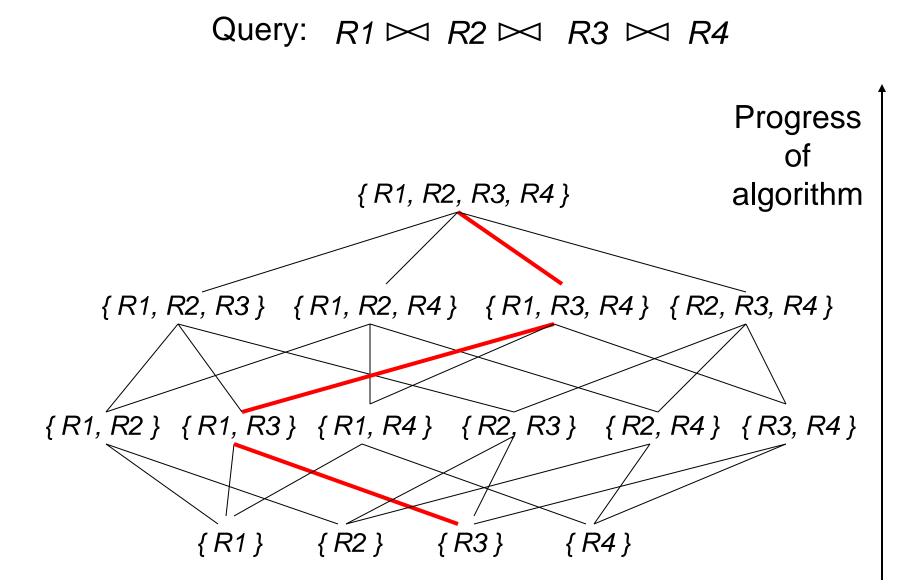
Number of tuples in  $R1 \bowtie R2 \bowtie R3$ 

#### OPT ( { R1, R2, R3 } ):

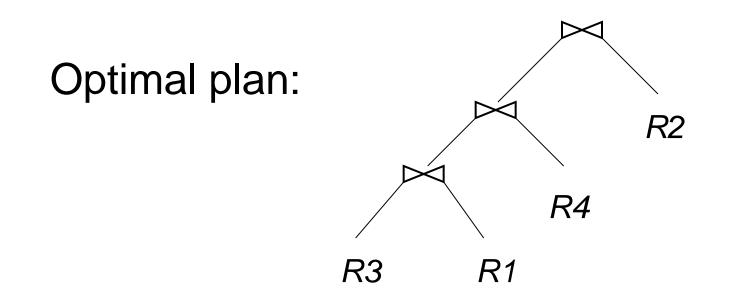


Note: Valid only for the simple cost model





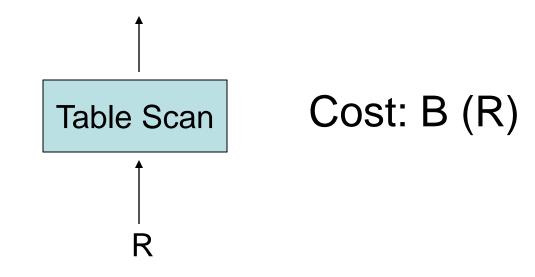
#### Query: $R1 \bowtie R2 \bowtie R3 \bowtie R4$



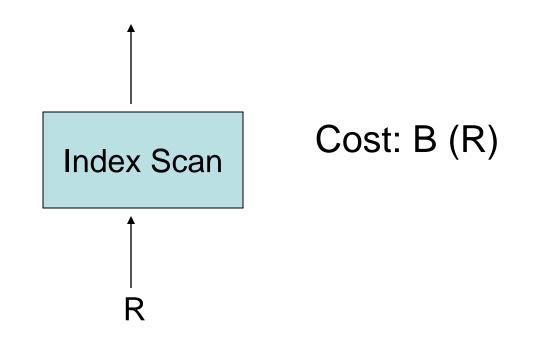
# More Complex Cost Model

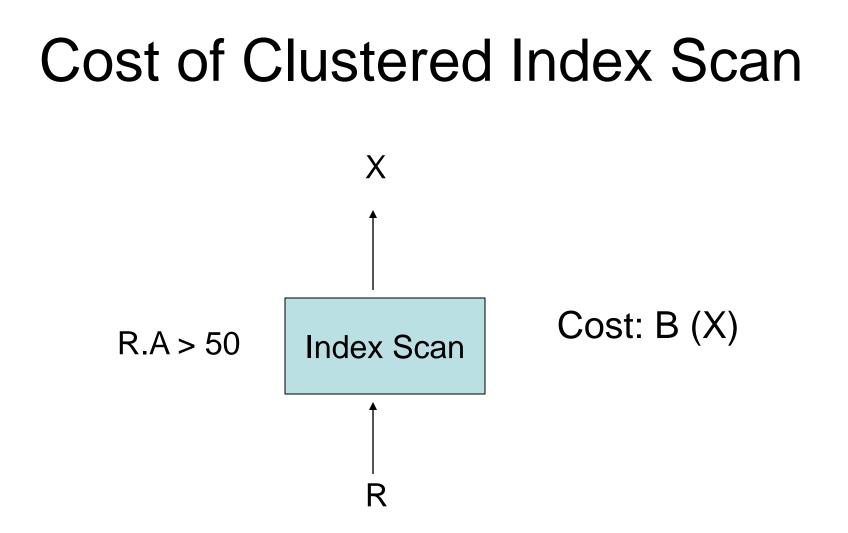
- DB System:
  - Two join algorithms:
    - Tuple-based nested loop join
    - Sort-Merge join
  - Two access methods
    - Table Scan
    - Index Scan (all indexes are in memory)
  - Plans pipelined as much as possible
- Cost: Number of disk I/O s

### **Cost of Table Scan**

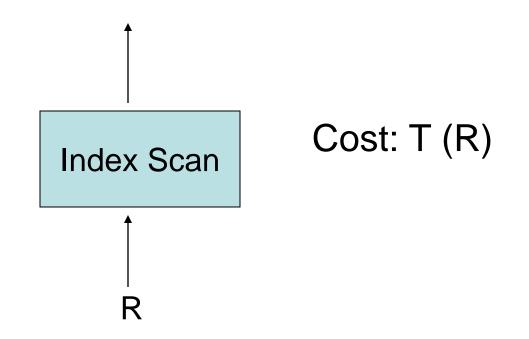


### **Cost of Clustered Index Scan**

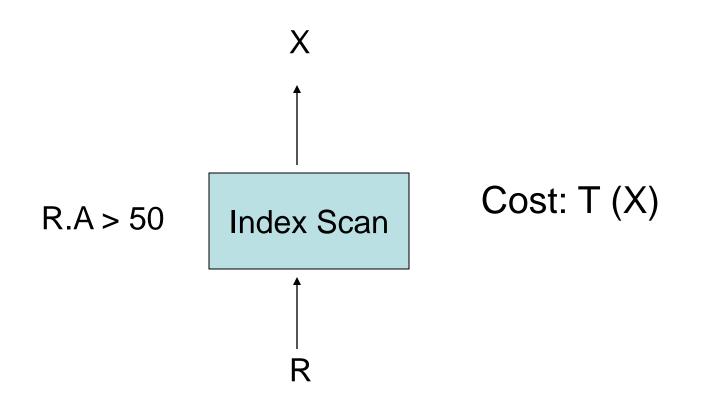




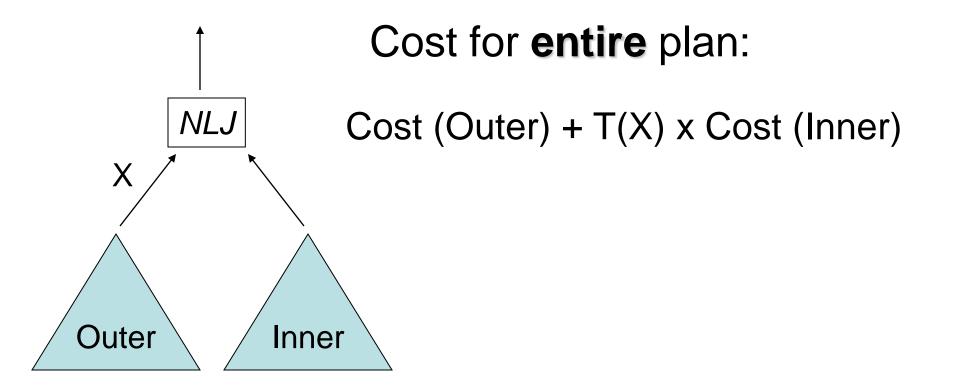
### Cost of Non-Clustered Index Scan

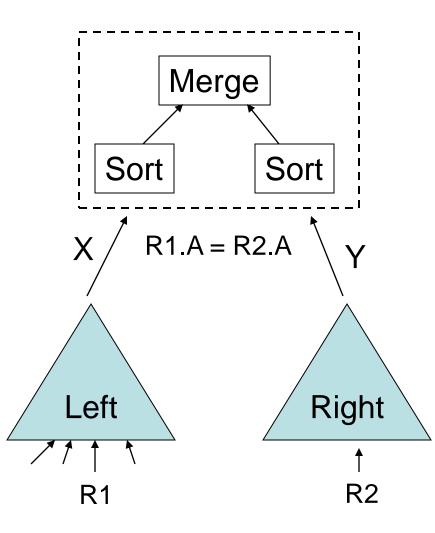


### Cost of Non-Clustered Index Scan



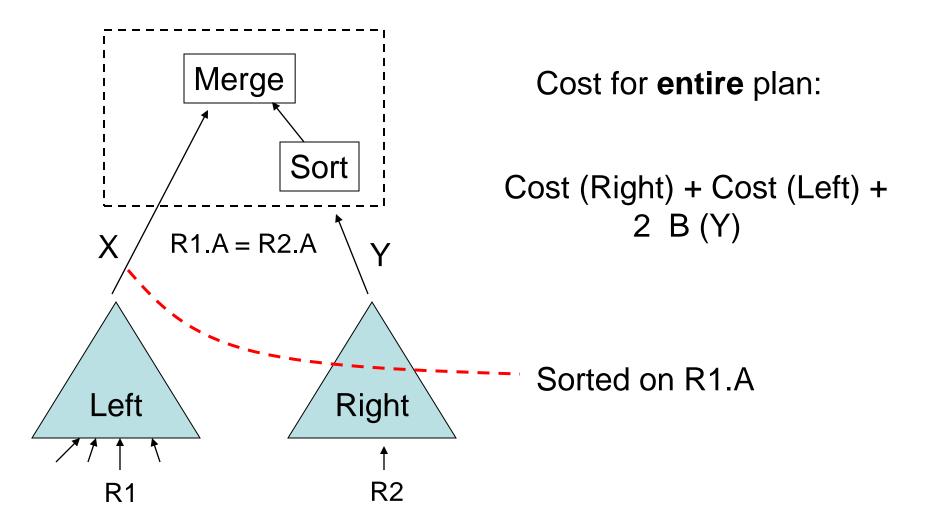
### Cost of Tuple-Based NLJ

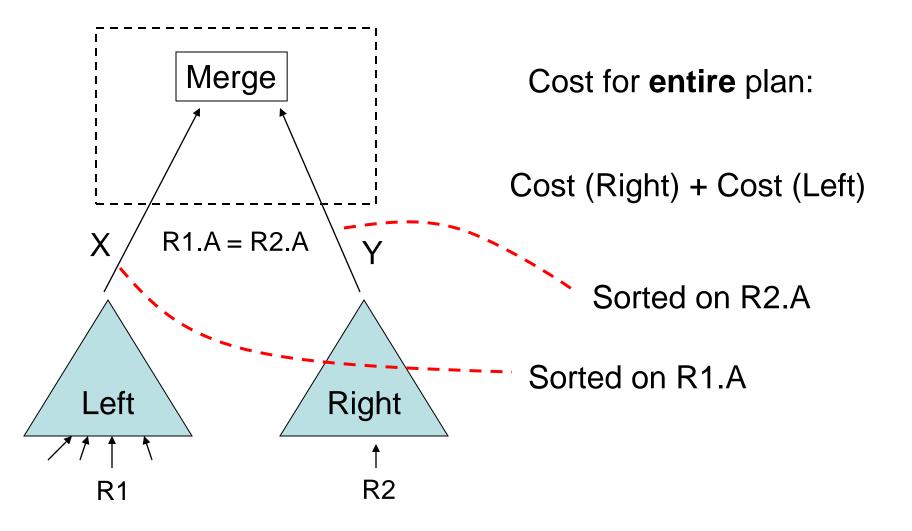




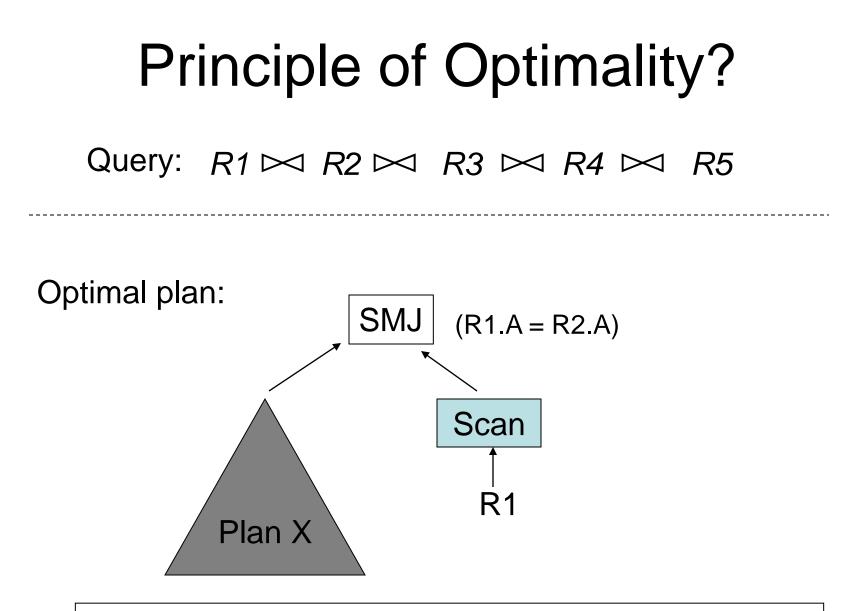
Cost for entire plan:

# Cost (Right) + Cost (Left) + 2 (B (X) + B (Y) )



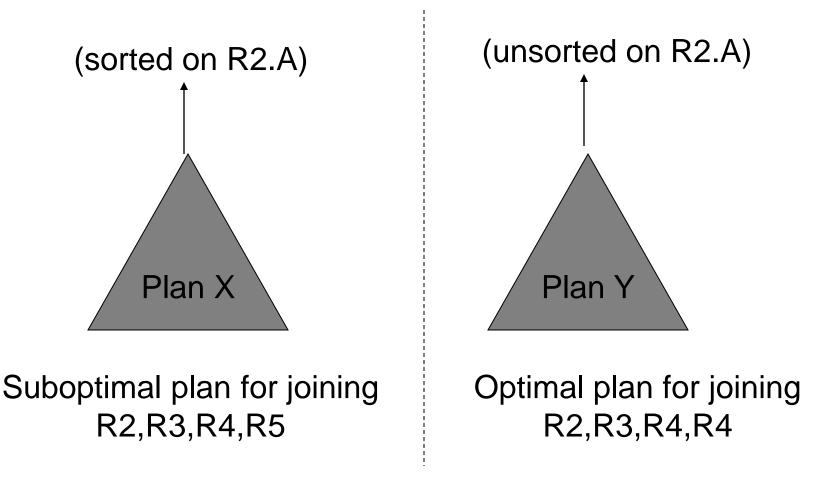


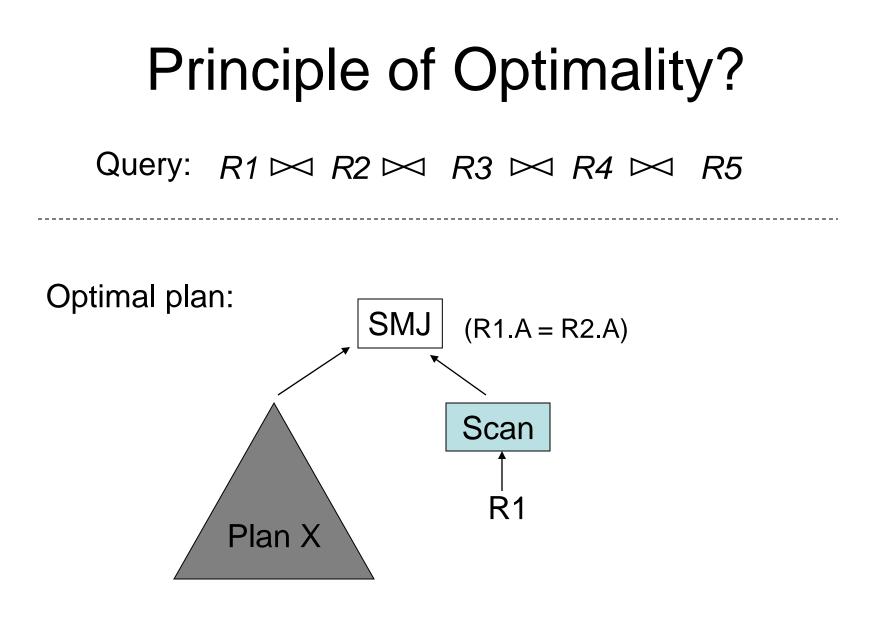
Bottom Line: Cost depends on sorted-ness of inputs



Is Plan X the optimal plan for joining R2,R3,R4,R5?

### Violation of Principle of Optimality





Can we assert anything about plan X?

# Weaker Principle of Optimality

If plan X produces output sorted on R2.A then plan X is the **optimal plan** for joining R2,R3,R4,R5 that produces output sorted on R2.A

If plan X produces output unsorted on R2.A then plan X is the **optimal plan** for joining R2, R3, R4, R5

# Interesting Order

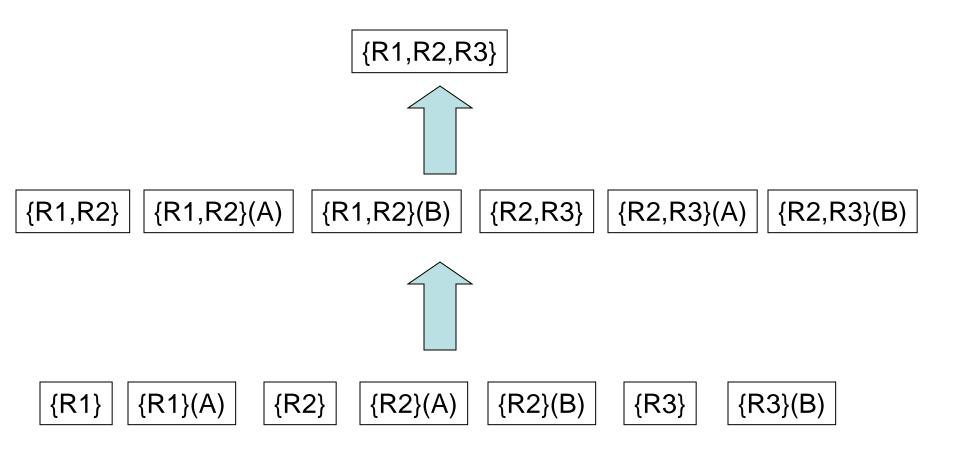
- An attribute is an **interesting order** if:
  - participates in a join predicate
  - Occurs in the Group By clause
  - Occurs in the Order By clause

### Interesting Order: Example

#### Select \* From R1(A,B), R2(A,B), R3(B,C) Where R1.A = R2.A and R2.B = R3.B

Interesting Orders: R1.A, R2.A, R2.B, R3.B

# Modified Selinger Algorithm



### Notation

Optimal way of joining R1, R2 so that output is sorted on attribute R2.C

#### Modified Selinger Algorithm $\{R1, R2, R3\}$ {R1, R2}(A) {R1,R2}(B) {R1,R2} {R2,R3} | {R2,R3}(A) | {R2, R3}(B) {R1}(A) {R2}(A) {R1} {R2} {R2}(B) {R3} {R3}(B)