

CS 564: Database Management Systems

Lecture 36: Distributed Analytical Database

Xiangyao Yu 4/22/2024

### Announcement

### Past final exams posted on canvas

#### Final exam

- May 6<sup>th</sup>, 10:05am-12:05pm
- Room 1: VAN VLECK B130 (lastname ≤ 'Mao')
- Room 2: NOLAND 132 (lastname > 'Mao')
- Two cheat sheets allowed; can reuse the midterm one
- Cumulative with roughly 70% content after midterm
- Contact instructor by May 1st if need McBurney accommodation

Course evaluation: <a href="https://heliocampusac.wisc.edu/">https://heliocampusac.wisc.edu/</a>

# Outline

### **Distributed architecture**

Data partitioning

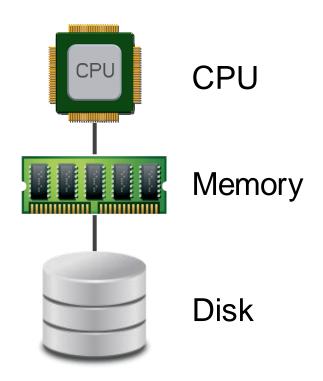
Query execution

Merge and Split operators

Join execution

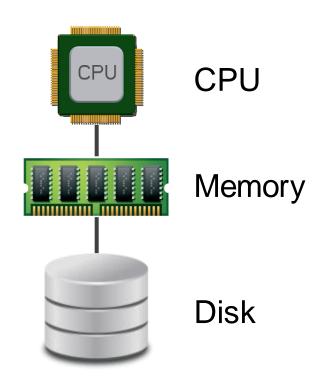
- Co-partitioned
- Partition-based
- Broadcast-based

# Centralized vs. Distributed Database

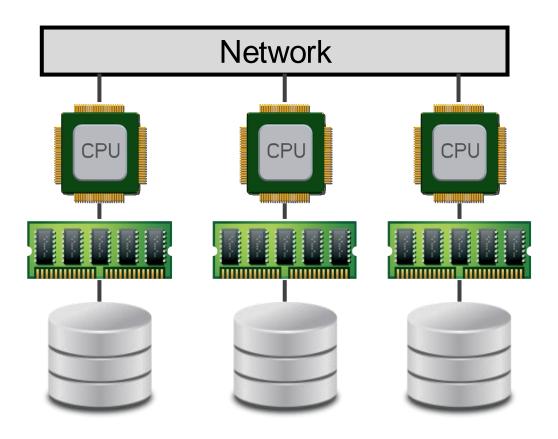


**Centralized database** 

## Centralized vs. Distributed Database

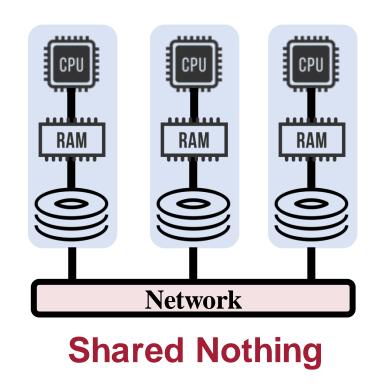


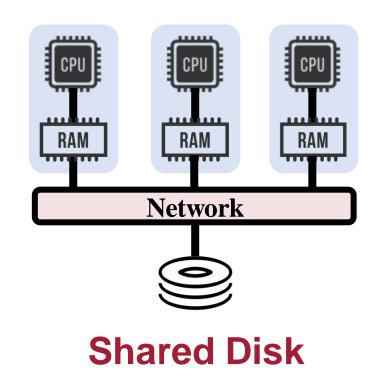
Centralized database

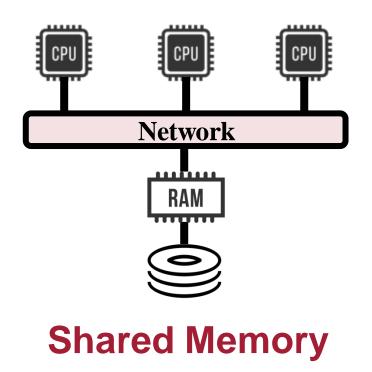


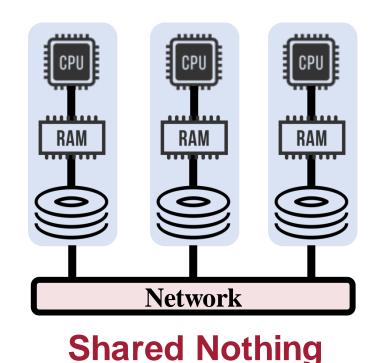
**Distributed database** 

- Multiple nodes connected using network
- Communication cost can be significant







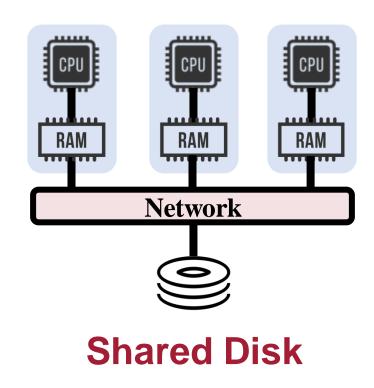


Each memory and disk is owned by some processor that acts as a server for that data

- Scales to thousands of servers and beyond

Important optimization goal: minimize network data transfer

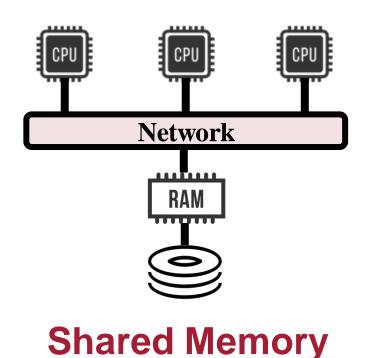
This lecture focuses on shared-nothing



Each processor has a private memory but has direct access to all disks

Scale to tens of servers

Example: Network attached storage (NAS) and storage area network (SAN)



All processors share direct access to a common global memory and to all disks

Does not scale beyond a single server

Example: multicore processors

# Outline

### Distributed architecture

### **Data partitioning**

### Query execution

Merge and Split operators

### Join execution

- Co-partitioned
- Partition-based
- Broadcast-based

## Parallel SQL Database

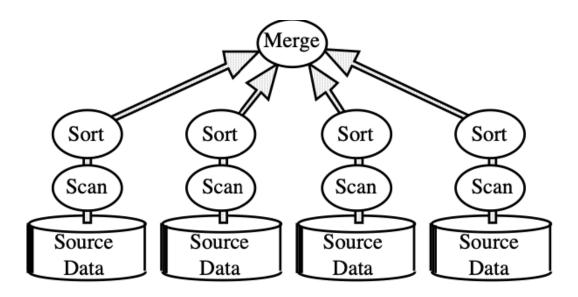
Most database programs are written in relational language SQL

- Goal: make SQL work on distributed system without rewriting
- Benefits of a high-level programming interface

# Parallel SQL Database

Most database programs are written in relational language SQL

- Goal: make SQL work on distributed system without rewriting
- Benefits of a high-level programming interface



**Partitioned Parallelism** 

### Partitioned Parallelism

#### Partition data across nodes

- Partition a table using the partition key (e.g., a single or multiple columns)
- Each record is mapped to one node

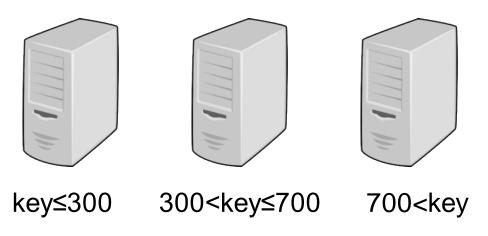
A single operator is executed collectively across multiple nodes

# Partitioned Parallelism

#### Partition data across nodes

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A single operator is executed collectively across multiple nodes



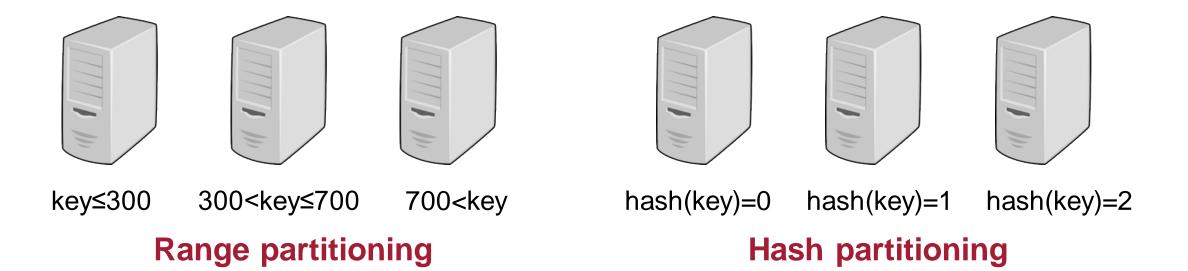
Range partitioning

### Partitioned Parallelism

#### Partition data across nodes

- Partition a table using the partition key (e.g., a single or multiple columns)
- Each record is mapped to one node

A single operator is executed collectively across multiple nodes



# Outline

Distributed architecture

Data partitioning

### **Query execution**

Merge and Split operators

### Join execution

- Co-partitioned
- Partition-based
- Broadcast-based

# Distributed Relational Operators

**Split** a relation into smaller relations

Merge multiple small relations into one bigger relation

# Distributed Relational Operators

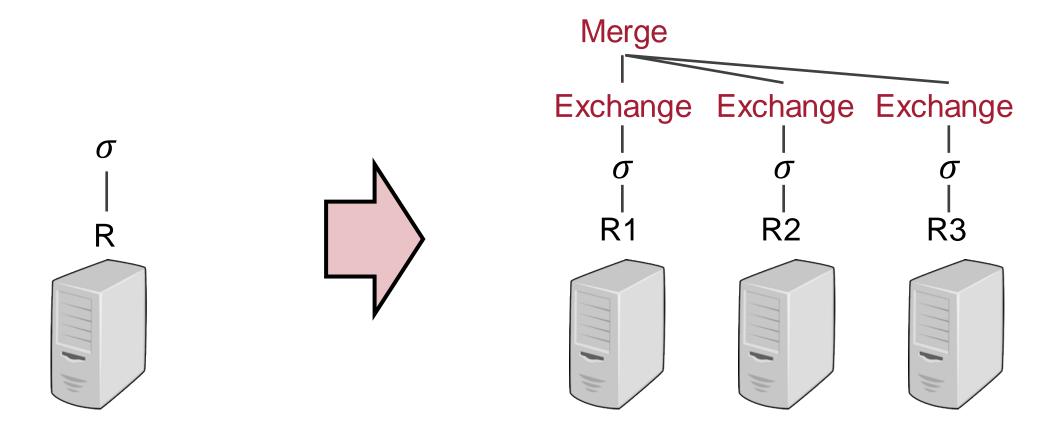
**Split** a relation into smaller relations

Merge multiple small relations into one bigger relation

Exchange: move relations between distributed nodes

- Broadcast: Send the source relation to all the nodes
- Shuffle: Split the relation and send each partition to the corresponding node

# Distributed Selection



Step 1: Each partition is filtered locally

Step 2: Filtered partitions are exchanged and merged in one node

# Outline

Distributed architecture

Data partitioning

Query execution

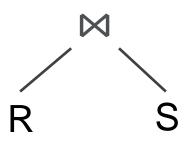
Merge and Split operators

### Join execution

- Co-partitioned
- Partition-based
- Broadcast-based

R (A, B), S (B, C)

SELECT \*
FROM R, S
WHERE R.B = S.B

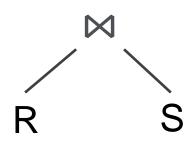


R (A, B), S (B, C)

SELECT \*
FROM R, S
WHERE R.B = S.B

### Assume R and S are co-partitioned

 Partitioned on the join key using the same partitioning function



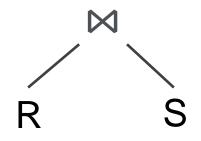
For example, rows in **R1** cannot possibly join with rows in **S2** 

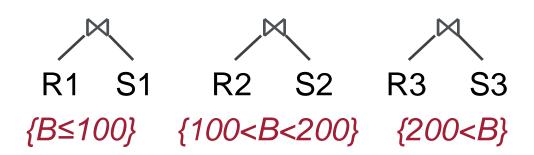
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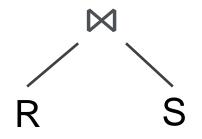




Join local partitions of R and S (i.e., R1⋈S1, R2⋈S2, R3⋈S3)

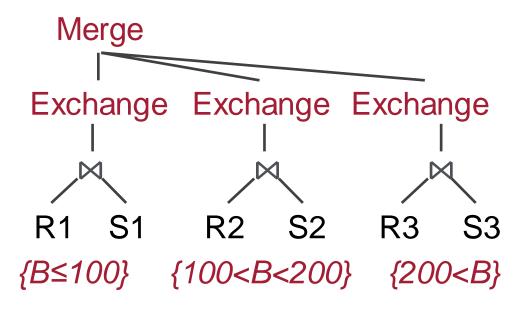
R(A, B), S(B, C)

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### Assume R and S are co-partitioned

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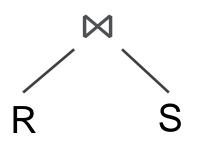
Join local partitions of R and S (i.e., R1⋈S1, R2⋈S2, R3⋈S3) Send join outputs to a single node and merge

R (A, B), S (B, C)

SELECT \*
FROM R, S
WHERE R.B = S.B

Relation R is range-partitioned on column B

Relation S is range-partitioned on column C



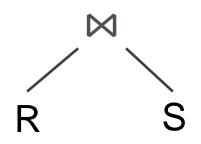
```
{B≤100} {100<B≤200} {200<B}
R1 S1 R2 S2 R3 S3
{C≤20} {20<C≤40} {40<C}
```

R (A, B), S (B, C)

SELECT \*
FROM R, S
WHERE R.B = S.B

Relation R is range-partitioned on column B

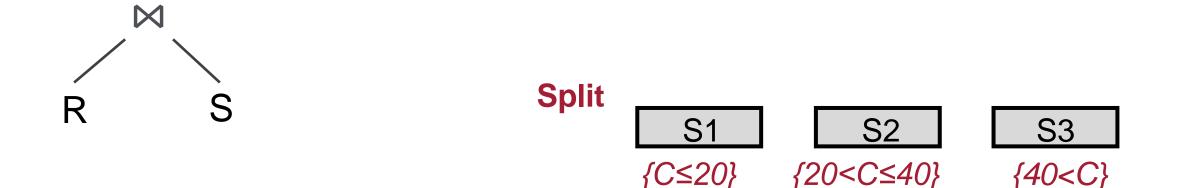
Relation S is range-partitioned on column C



$$\{B \le 100\}$$
  $\{100 < B \le 200\}$   $\{200 < B\}$   
R1 S1 R2 S2 R3 S3  
 $\{C \le 20\}$   $\{20 < C \le 40\}$   $\{40 < C\}$ 

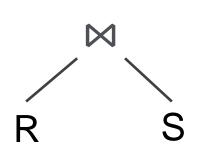
R (A, B), S (B, C)

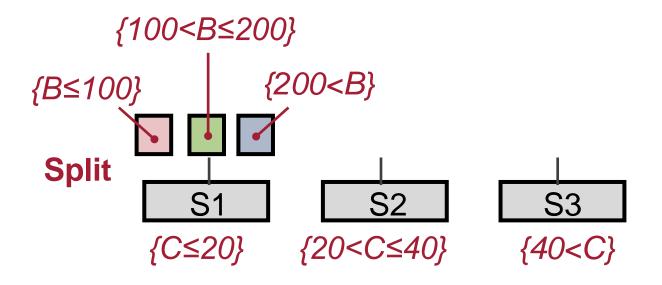
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```
R (A, B), S (B, C)
```

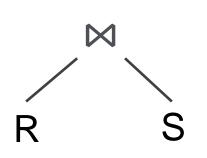
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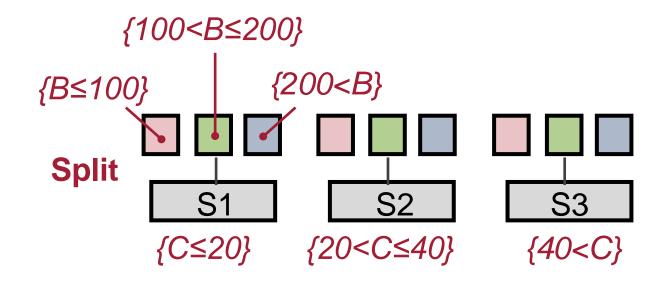




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R (A, B), S (B, C)
```

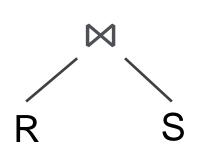
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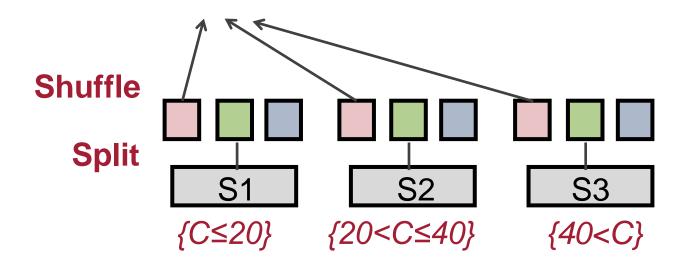




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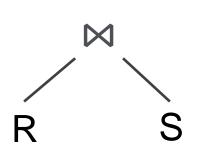
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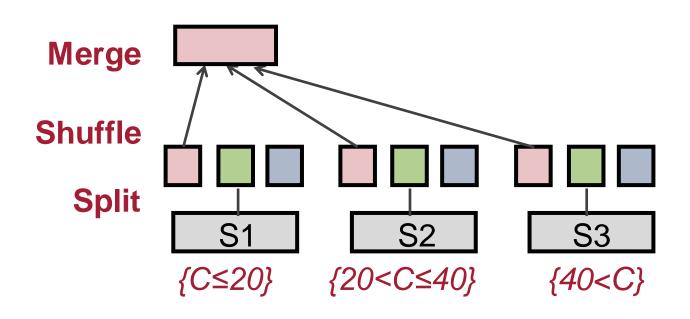




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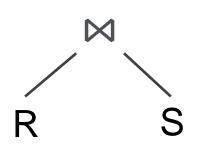
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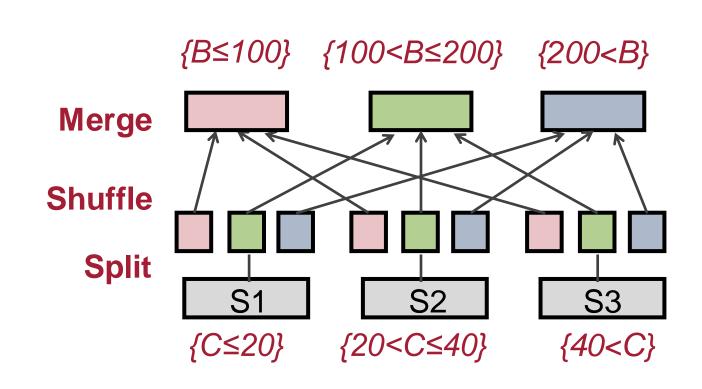




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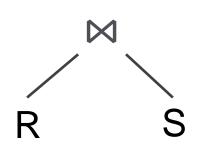
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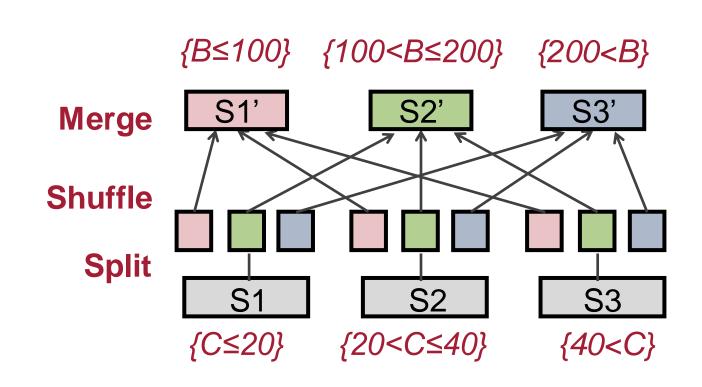




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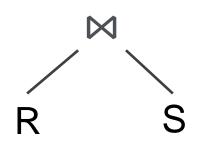


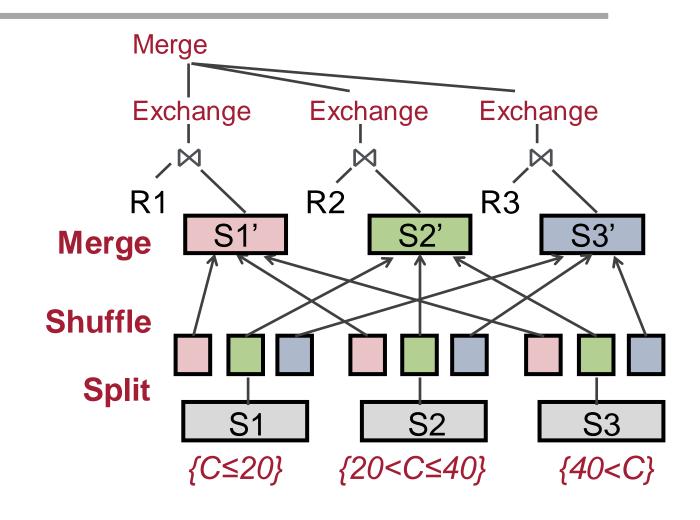


Now both R and S are co-partitioned on B

R (A, B), S (B, C)

SELECT \*
FROM R, S
WHERE R.B = S.B



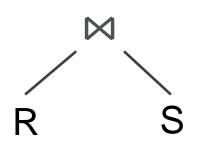


Now both R and S are co-partitioned on B Use the co-partitioned join algorithm to finish the query

R (A, B), S (B, C)

SELECT \*
FROM R, S
WHERE R.B = S.B

What if neither **R** nor **S** is partitioned on the join key?

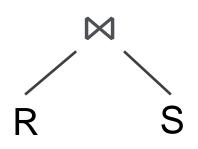


$$\{H(A)=0\}$$
  $\{H(A)=1\}$   $\{H(A)=2\}$   
R1 S1 R2 S2 R3 S3  
 $\{C \le 20\}$   $\{20 < C \le 40\}$   $\{40 < C\}$ 

R (A, B), S (B, C)

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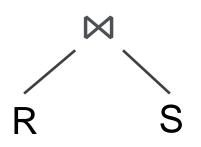


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R1 S1 R2 S2 R3 S3  
 $\{C \le 20\}$   $\{20 < C \le 40\}$   $\{40 < C\}$ 

Re-partition both R and S on column B using the same partition function!

```
R (A, B), S (B, C)
```

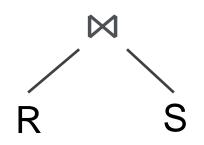
```
SELECT *
FROM R, S
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```



$$\{H(A)=0\}$$
  $\{H(A)=1\}$   $\{H(A)=2\}$   
R1 S1 R2 S2 R3 S3  
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```
R (A, B), S (B, C)
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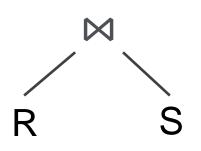
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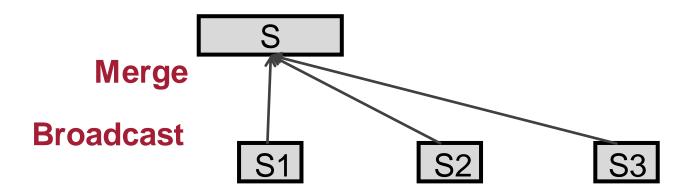


Broadcast S1 S2 S3

R (A, B), S (B, C)

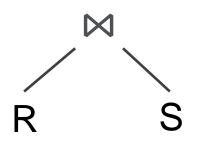
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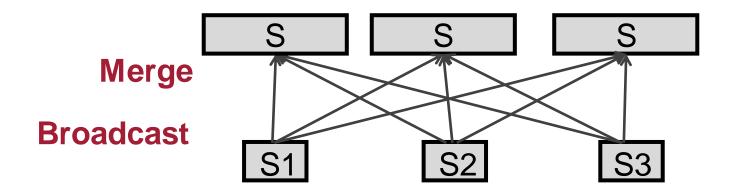




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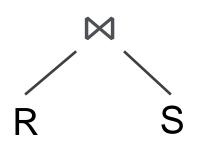
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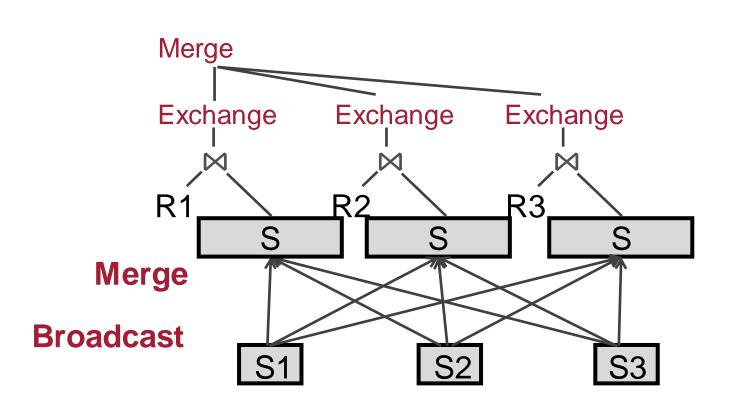




R (A, B), S (B, C)

SELECT \*
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### Distributed Join Methods

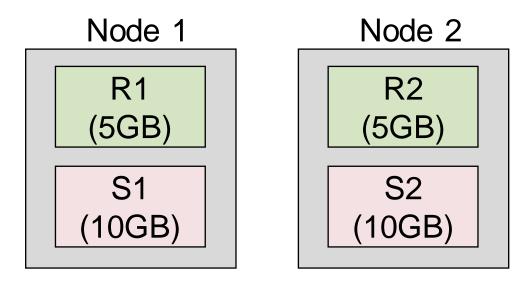
Co-partitioned: Works if the two relations are already co-partitioned on the join key (using the same partitioning function)

Partition-based: Partition one or two relations such that both relations are co-partitioned

Broadcast-based: Broadcast the smaller relation to all nodes

# Exercise

We execute R⋈S on two nodes with the data layout shown below. Assume S is partitioned on the join key, but R is not. What is the network IO cost for partition-based join and broadcast-based join, respectively? (we assume uniform distribution; we assume the join result size is negligible)



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