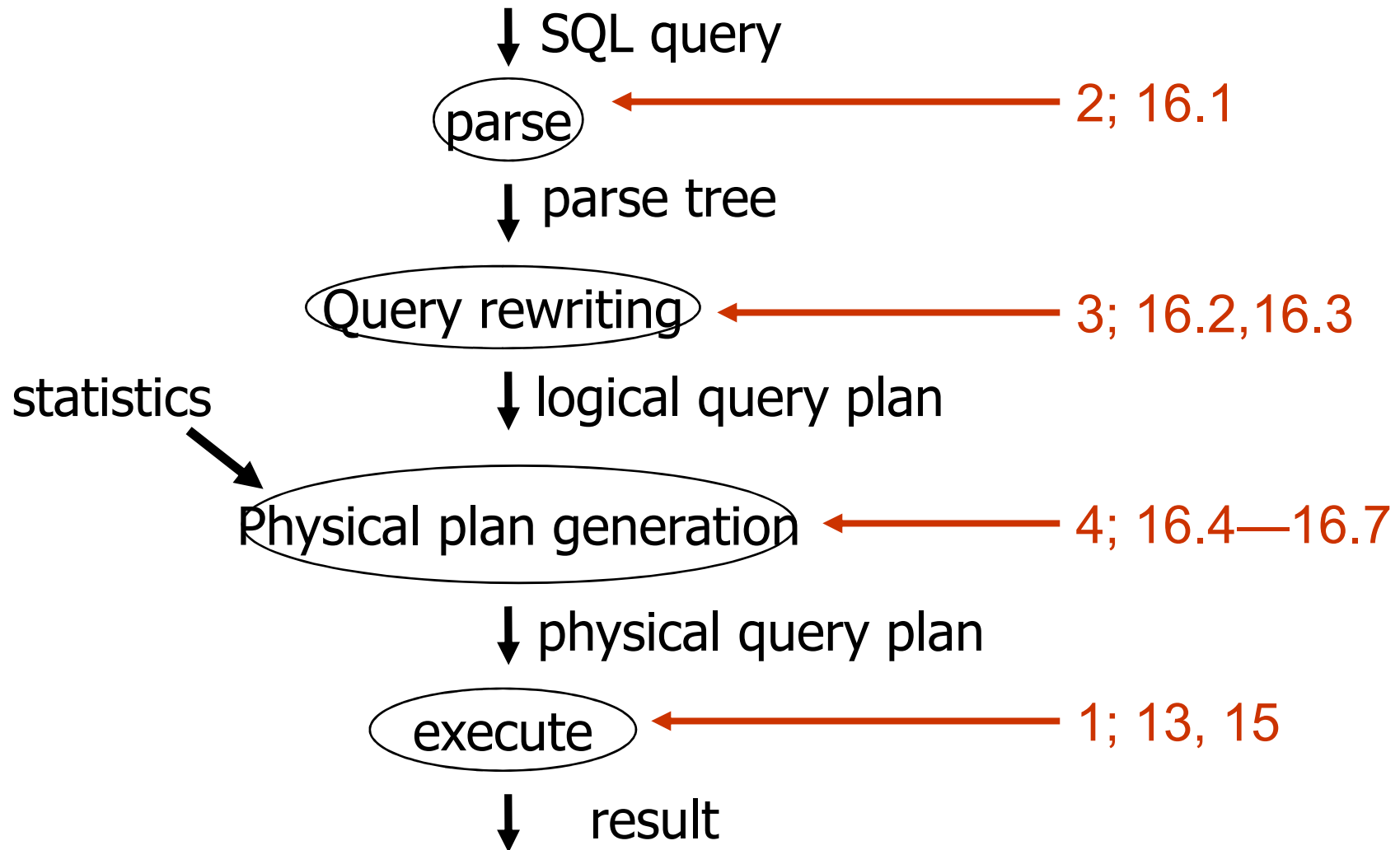


# CPS216: Advanced Database Systems

## **Notes 07: Query Execution (Sort and Join operators)**

Shivnath Babu

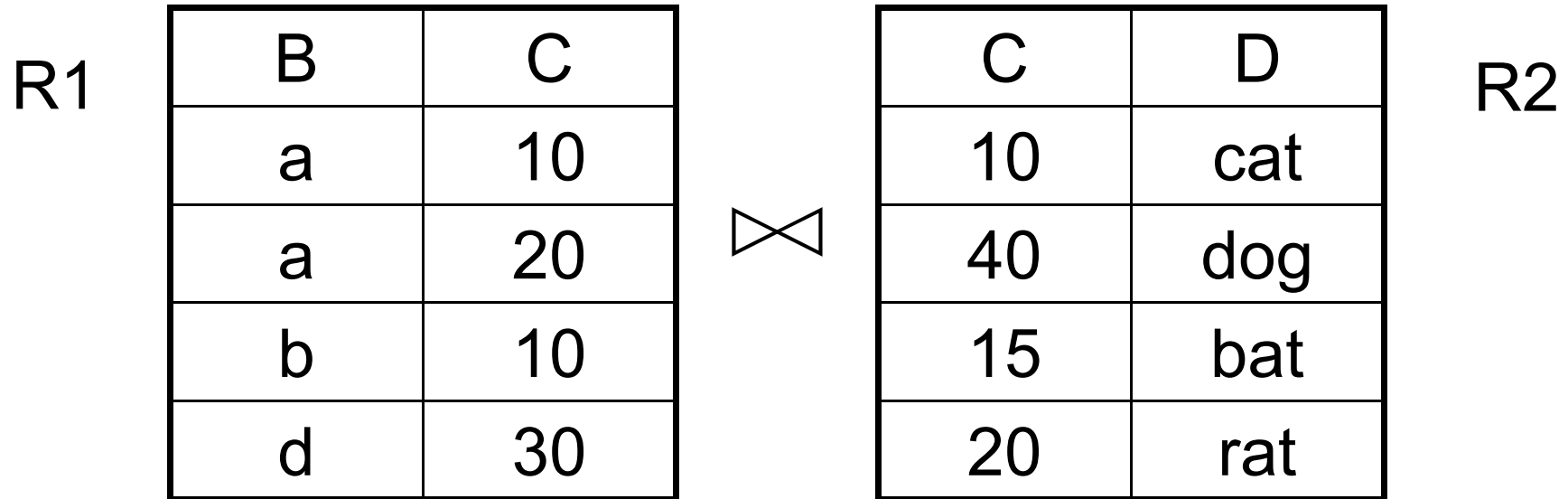
# Query Processing - In class order



# Roadmap

- A simple operator: Nested Loop Join
- Preliminaries
  - Cost model
  - Clustering
  - Operator classes
- Operator implementation (with examples from joins)
  - Scan-based
  - Sort-based
  - Using existing indexes
  - Hash-based
- Buffer Management
- Parallel Processing

# Nested Loop Join (NLJ)



- NLJ (conceptually)
  - for each  $r \in R1$  do
    - for each  $s \in R2$  do
      - if  $r.C = s.C$  then output  $r,s$  pair

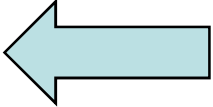
# Nested Loop Join (contd.)

- Tuple-based
- Block-based
- Asymmetric

# Implementing Operators

- Basic algorithm
  - Scan-based (e.g., NLJ)
  - Sort-based
  - Using existing indexes
  - Hash-based (building an index on the fly)
- Memory management
  - Tradeoff between memory and #IOs
- Parallel processing

# Roadmap

- A simple operator: Nested Loop Join
- Preliminaries 
  - Cost model
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# Operator Cost Model

- **Simplest:** Count # of disk blocks read and written during operator execution
- Extends to query plans
  - Cost of query plan = Sum of operator costs
- Caution: Ignoring CPU costs



# Assumptions

- Single-processor-single-disk machine
  - Will consider parallelism later
- Ignore cost of writing out result
  - Output size is independent of operator implementation
- Ignore # accesses to index blocks

# Parameters used in Cost Model


$B(R)$  = # blocks storing  $R$  tuples

$T(R)$  = # tuples in  $R$

$V(R,A)$  = # distinct values of attr  $A$  in  $R$

$M$  = # memory blocks available

# Roadmap

- A simple operator: Nested Loop Join
- Preliminaries
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# Notions of clustering

- Clustered file organization

R1 R2 S1 S2

R3 R4 S3 S4

.....

- Clustered relation

R1 R2 R3 R4

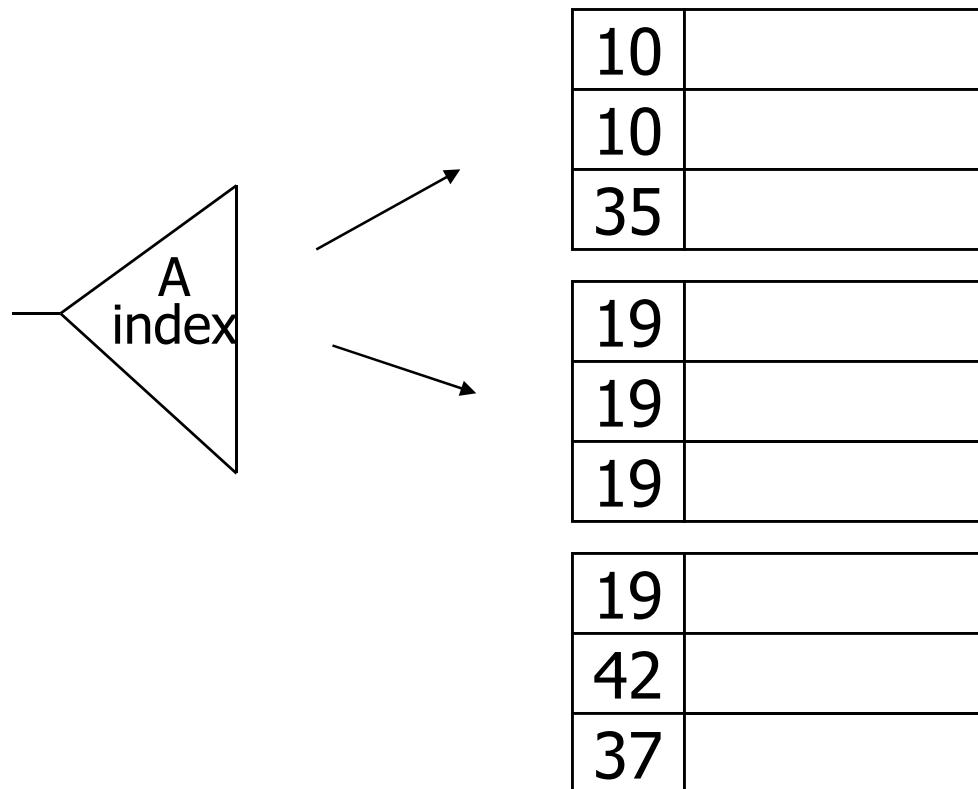
R5 R5 R7 R8

.....

- Clustering index

# Clustering Index

Tuples with a given value of the search key packed in as few blocks as possible



# Examples

$$T(R) = 10,000$$

$$B(R) = 200$$

If R is **clustered**, then # R tuples per block =  
 $10,000/200 = 50$

$$\text{Let } V(R,A) = 40$$

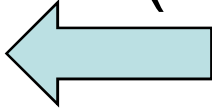
→ If I is a **clustering index** on R.A, then # IOs to  
access  $\sigma_{R.A = "a"}(R) = 250/50 = 5$

→ If I is a **non-clustering index** on R.A, then #  
IOs to access  $\sigma_{R.A = "a"}(R) = 250 (> B(R))$

# Operator Classes

	Tuple-at-a-time	Full-relation
Unary	Select	Sort
Binary		Difference

# Roadmap

- A simple operator: Nested Loop Join
  - Preliminaries
    - Cost model
    - Clustering
    - Operator classes
  - Operator implementation (with examples from joins)
    - Scan-based
    - Sort-based
    - Using existing indexes
    - Hash-based
  - Buffer Management
  - Parallel Processing
- 



# Implementing Tuple-at-a-time Operators

- One pass algorithm:
  - Scan
  - Process tuples one by one
  - Write output
- $\text{Cost} = B(R)$ 
  - Remember:  $\text{Cost} = \# \text{ IOs}$ , and we ignore the cost to write output

# Implementing a Full-Relation Operator, Ex: Sort

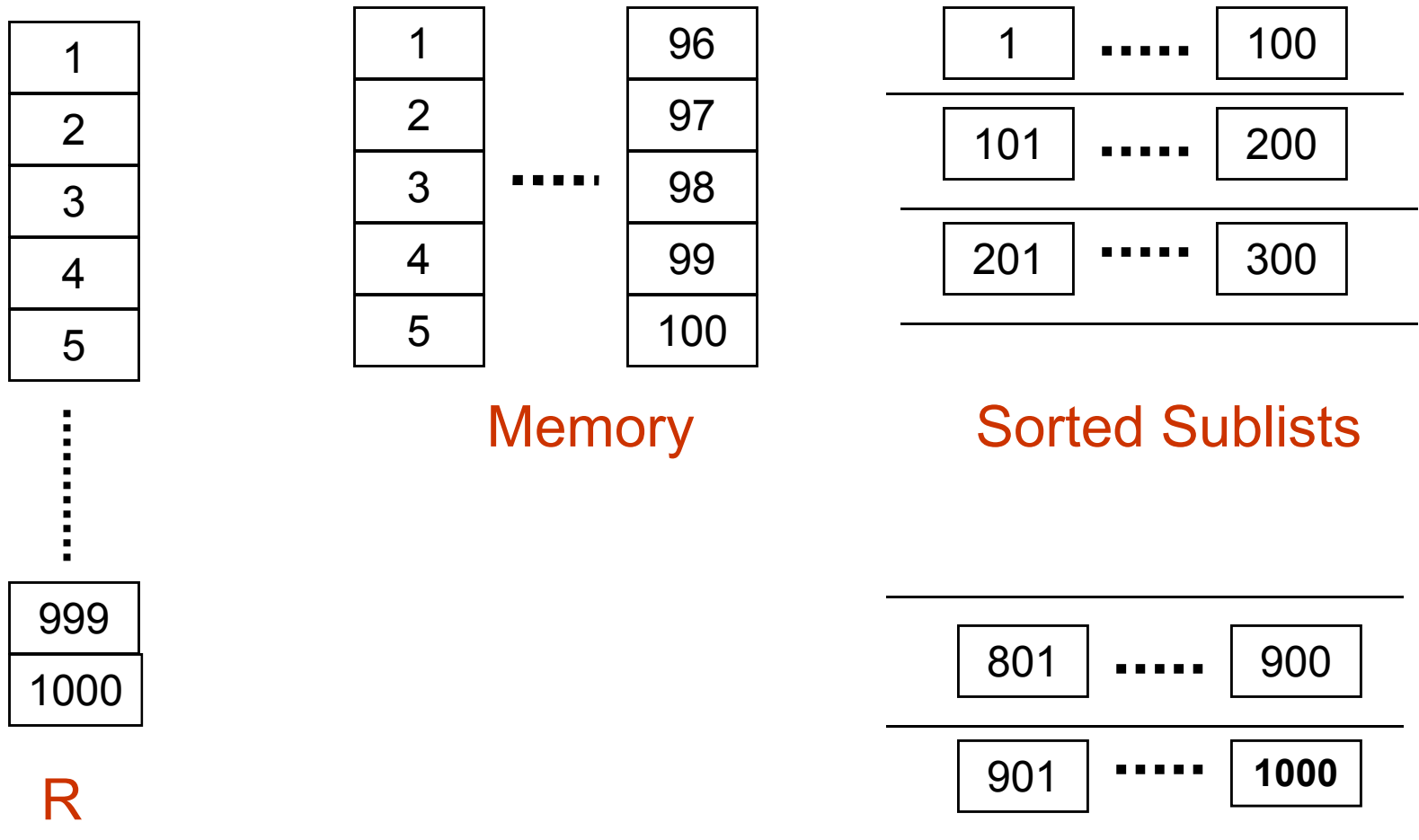
- Suppose  $T(R) \times \text{tupleSize}(R) \leq M \times |B(R)|$
- Read  $R$  completely into memory
- Sort
- Write output
- Cost =  $B(R)$

# Implementing a Full-Relation Operator, Ex: Sort

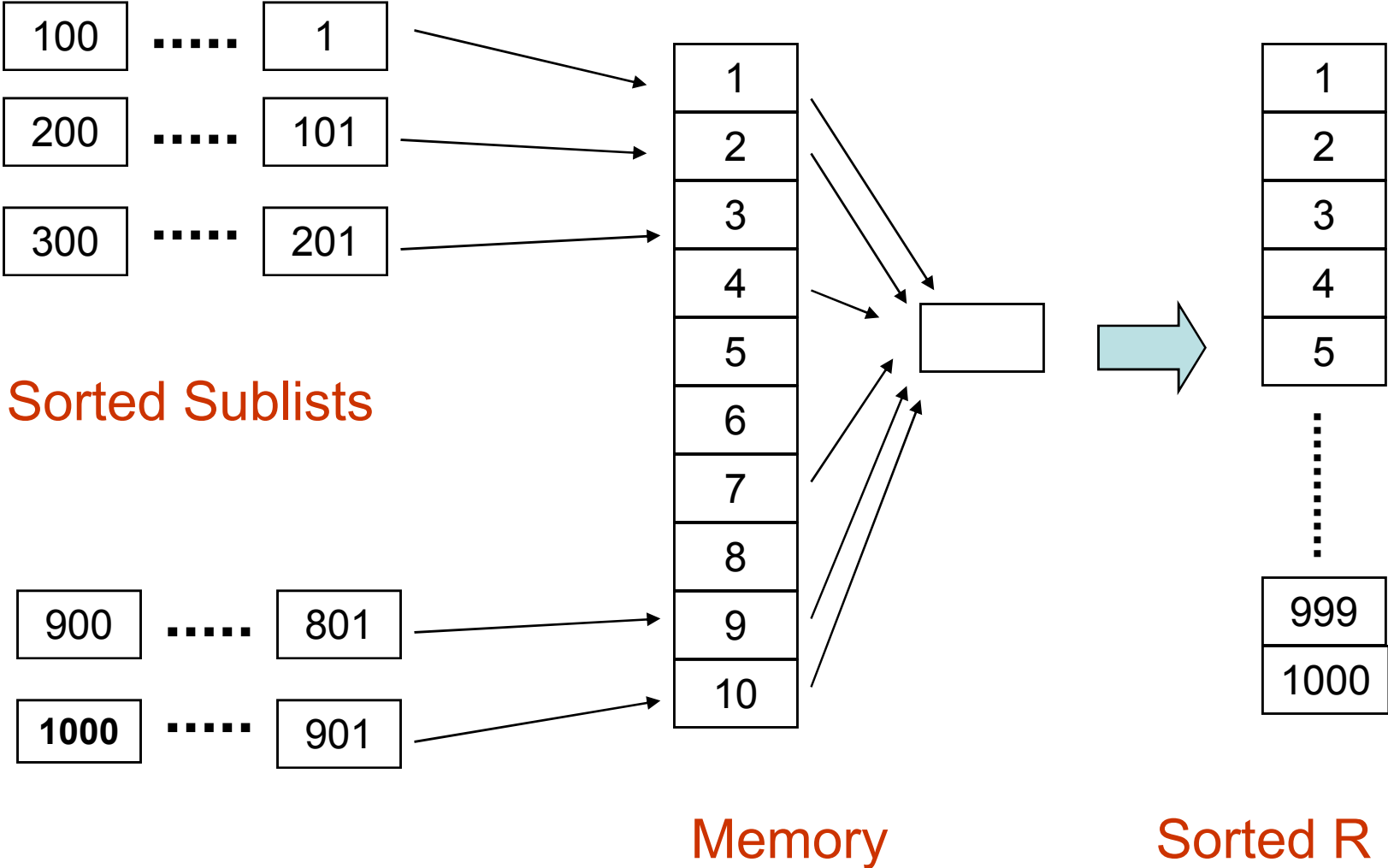
- Suppose  $R$  won't fit within  $M$  blocks
- Consider a two-pass algorithm for Sort; generalizes to a multi-pass algorithm
- Read  $R$  into memory in  $M$ -sized chunks
- Sort each chunk in memory and write out to disk as a **sorted sublist**
- Merge all sorted sublists
- Write output

# Two-phase Sort: Phase 1

Suppose  $B(R) = 1000$ ,  $R$  is clustered, and  $M = 100$



# Two-phase Sort: Phase 2



# Analysis of Two-Phase Sort

- Cost =  $3xB(R)$  if  $R$  is clustered,  
=  $B(R) + 2B(R')$  otherwise
- Memory requirement  $M \geq B(R)^{1/2}$

# Duplicate Elimination

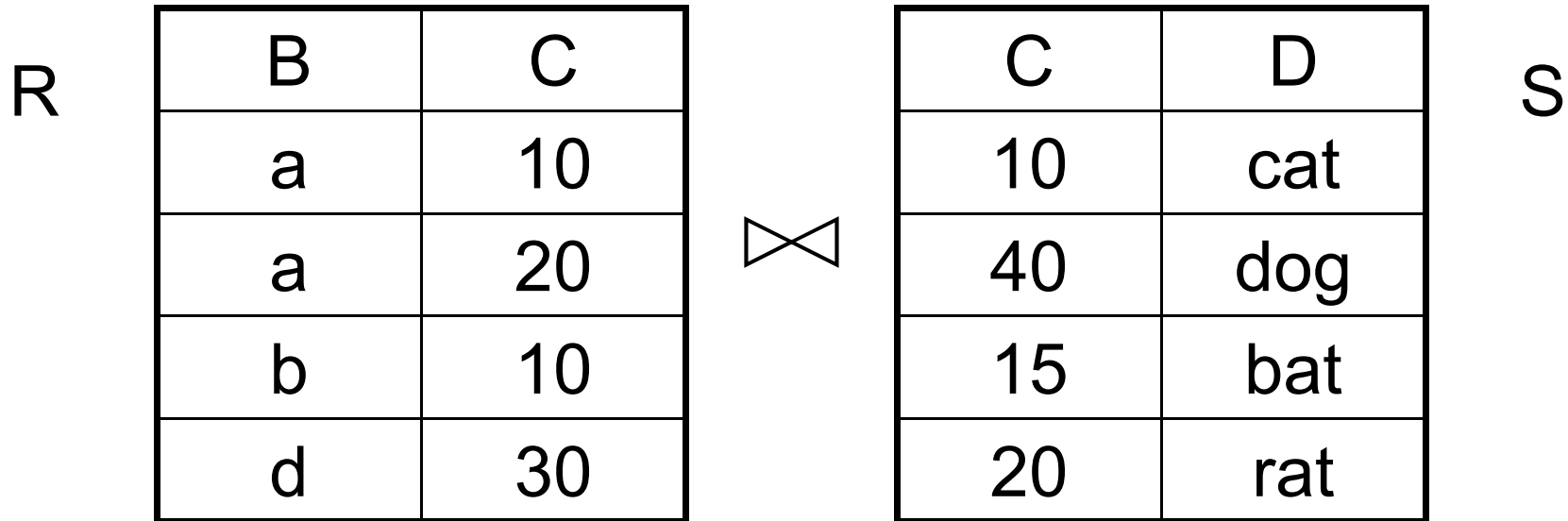
- Suppose  $B(R) \leq M$  and  $R$  is clustered
- Use an in-memory index structure
- Cost =  $B(R)$
- Can we do with less memory?
  - $B(\delta(R)) \leq M$
  - Aggregation is similar to duplicate elimination

# Duplicate Elimination Based on Sorting

- Sort, then eliminate duplicates
- Cost = Cost of sorting +  $B(R)$
- Can we reduce cost?
  - Eliminate duplicates during the merge phase



# Back to Nested Loop Join (NLJ)



- NLJ (conceptually)
  - for each  $r \in R$  do
    - for each  $s \in S$  do
      - if  $r.C = s.C$  then output  $r,s$  pair

# Analysis of Tuple-based NLJ

- Cost with R as outer =  $T(R) + T(R) \times T(S)$
- Cost with S as outer =  $T(S) + T(R) \times T(S)$
- $M \geq 2$

# Block-based NLJ

- Suppose R is outer
  - Loop: Get the next M-1 R blocks into memory
  - Join these with each block of S
- $B(R) + (B(R)/M-1) \times B(S)$
- What if S is outer?
  - $B(S) + (B(S)/M-1) \times B(R)$

## Let us work out an NLJ Example

- Relations are not clustered
- $T(R1) = 10,000$      $T(R2) = 5,000$   
10 tuples/block for R1; and for R2  
M = 101 blocks

Tuple-based NLJ Cost: for each R1 tuple:

[Read tuple + Read R2]

Total = 10,000 [1 + 5000] = 50,010,000 IOs

# Can we do better when R,S are not clustered?

## Use our memory

- (1) Read 100 blocks worth of R1 tuples
- (2) Read all of R2 (1 block at a time) + join
- (3) Repeat until done

Cost: for each R1 chunk:

Read chunk: 1000 IOs

Read R2: 5000 IOs

Total/chunk = 6000

$$\text{Total} = \frac{10,000}{1,000} \times 6000 = 60,000 \text{ IOs}$$

[Vs. 50,010,000!]

- Can we do better?

➔ Reverse join order:  $R2 \bowtie R1$

$$\text{Total} = \frac{5000}{1000} \times (1000 + 10,000) =$$

$$5 \times 11,000 = 55,000 \text{ IOs}$$

[Vs. 60,000]

## Example contd. NLJ $R2 \bowtie R1$

- Now suppose relations are **clustered**

### Cost

For each R2 chunk:

Read chunk: 100 IOs

Read R1: 1000 IOs

Total/chunk = 1,100

Total= 5 chunks x 1,100 = 5,500 IOs

[Vs. 55,000]



# Joins with Sorting

- **Sort-Merge Join** (conceptually)
  - (1) if R1 and R2 not sorted, sort them
  - (2)  $i \leftarrow 1; j \leftarrow 1;$   
While  $(i \leq T(R1)) \wedge (j \leq T(R2))$  do
    - if  $R1\{i\}.C = R2\{j\}.C$  then **OutputTuples**
    - else if  $R1\{i\}.C > R2\{j\}.C$  then  $j \leftarrow j+1$
    - else if  $R1\{i\}.C < R2\{j\}.C$  then  $i \leftarrow i+1$

## Procedure **Output-Tuples**

While  $(R1\{i\}.C = R2\{j\}.C) \wedge (i \leq T(R1))$  do

[  $jj \leftarrow j$ ;

while  $(R1\{i\}.C = R2\{jj\}.C) \wedge (jj \leq T(R2))$  do

[ output pair  $R1\{i\}$ ,  $R2\{jj\}$ ;

$jj \leftarrow jj+1$  ]

$i \leftarrow i+1$  ]

# Example

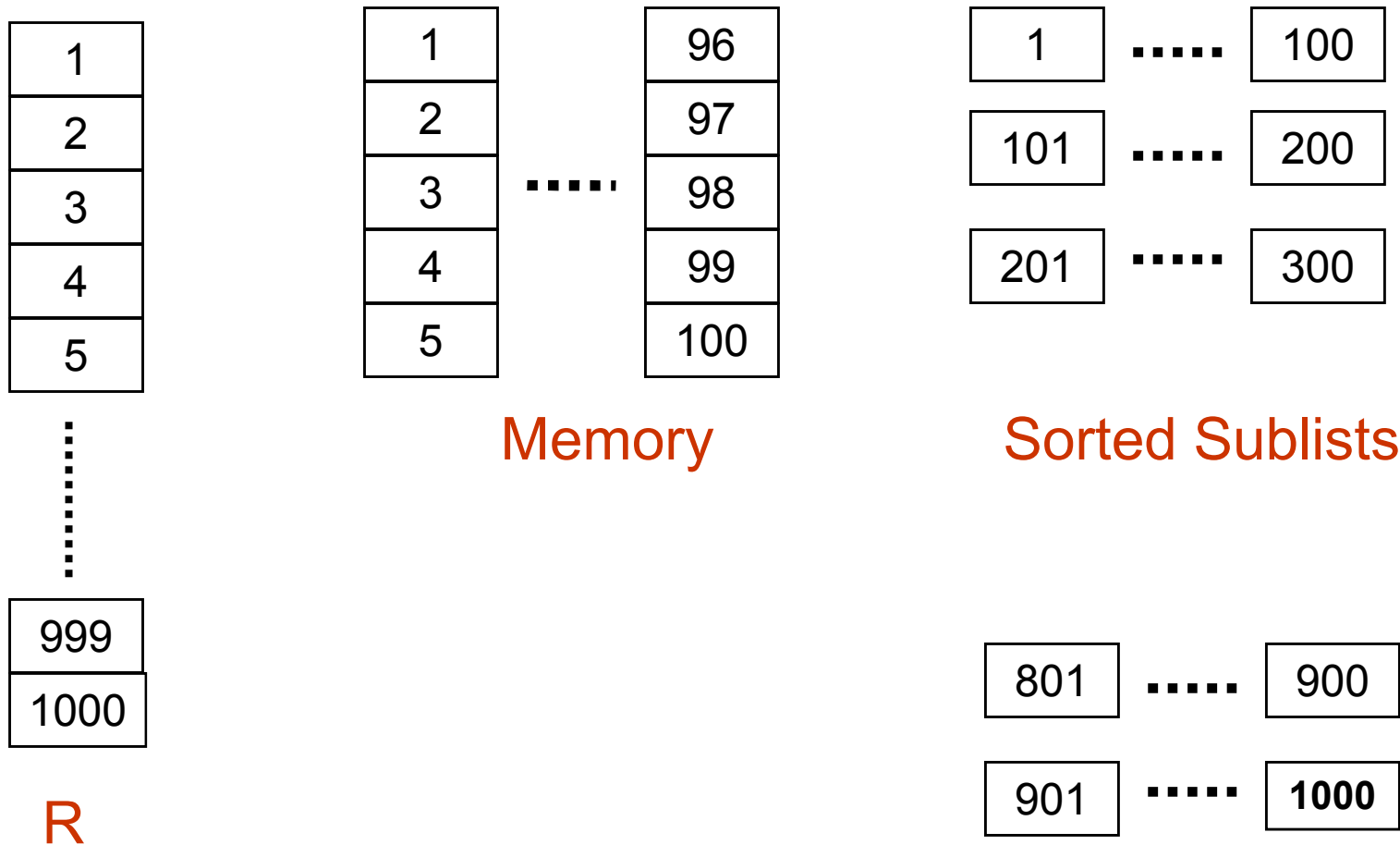
$i$	$R1\{i\}.C$	$R2\{j\}.C$	$j$
1	10	5	1
2	20	20	2
3	20	20	3
4	30	30	4
5	40	30	5
		50	6
		52	7

# Block-based Sort-Merge Join

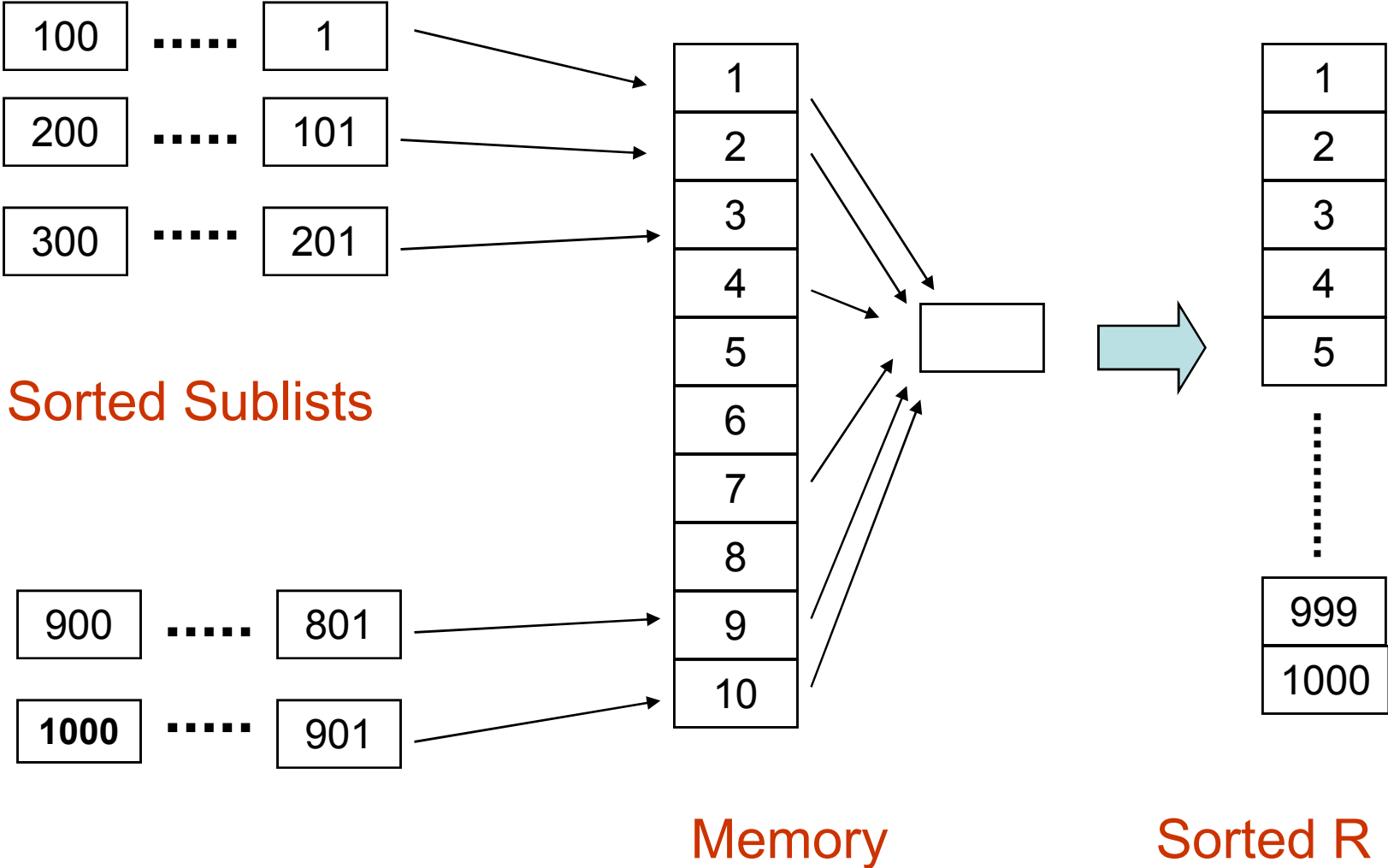
- Block-based **sort**
- Block-based **merge**

# Two-phase Sort: Phase 1

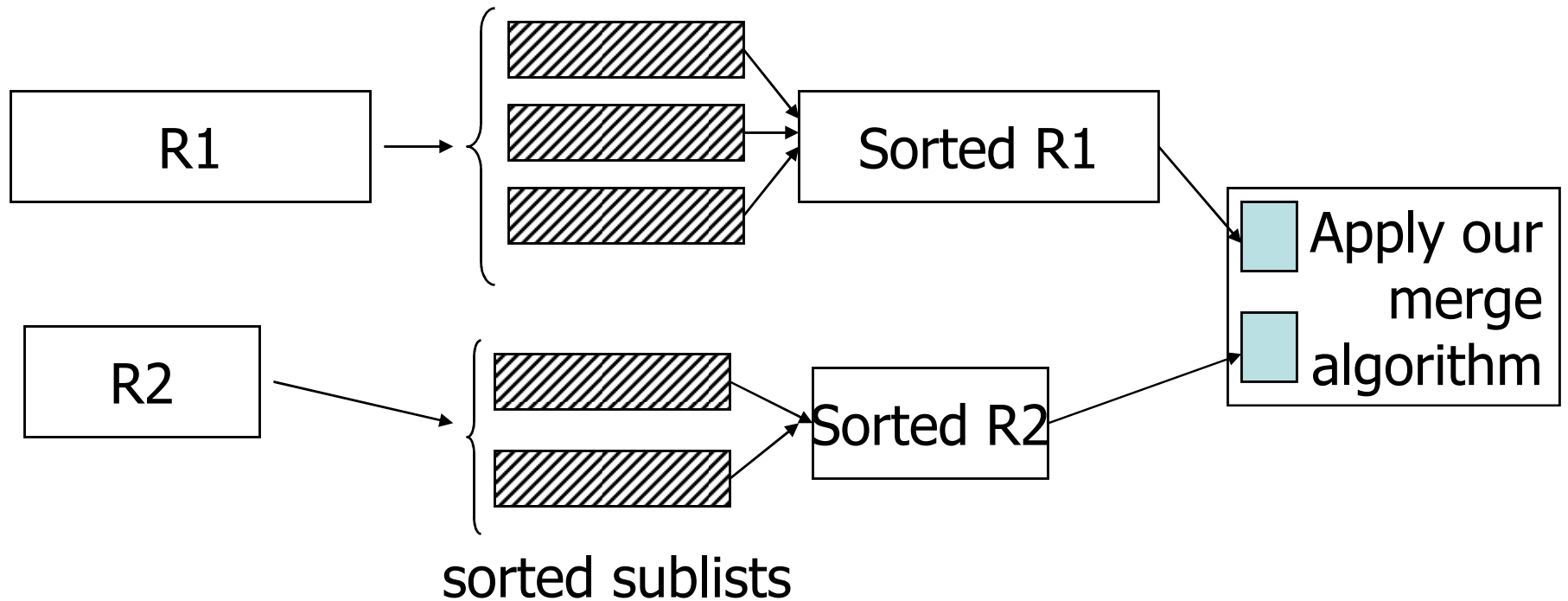
Suppose  $B(R) = 1000$  and  $M = 100$



# Two-phase Sort: Phase 2



# Sort-Merge Join



# Analysis of Sort-Merge Join

- Cost =  $5 \times (B(R) + B(S))$
- Memory requirement:  
 $M \geq (\max(B(R), B(S)))^{1/2}$



## Continuing with our Example

R1,R2 clustered, but unordered

$$\begin{aligned}\text{Total cost} &= \text{sort cost} + \text{join cost} \\ &= 6,000 + 1,500 = 7,500 \text{ IOs}\end{aligned}$$

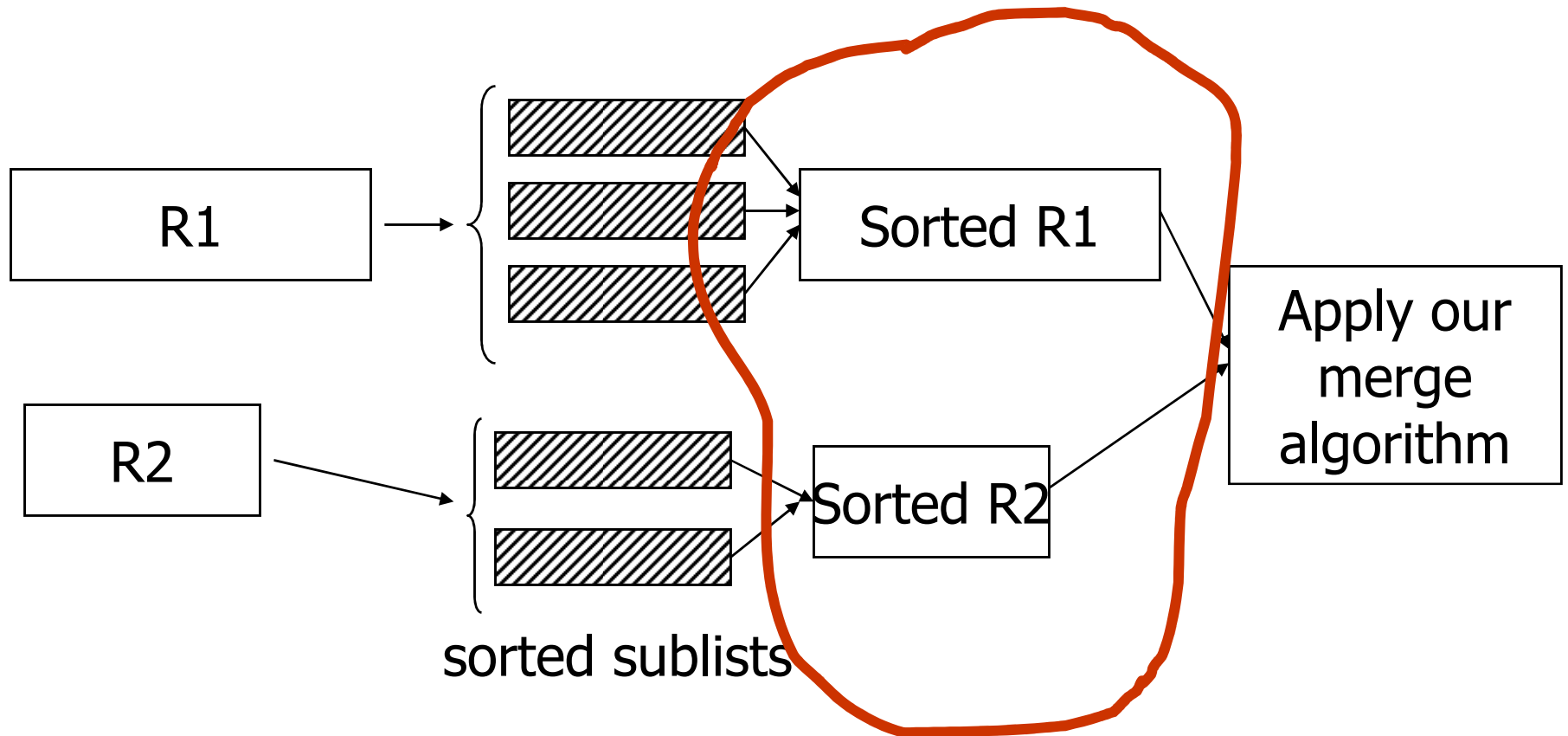
But: NLJ cost = 5,500

So merge join does not pay off!

## However ...

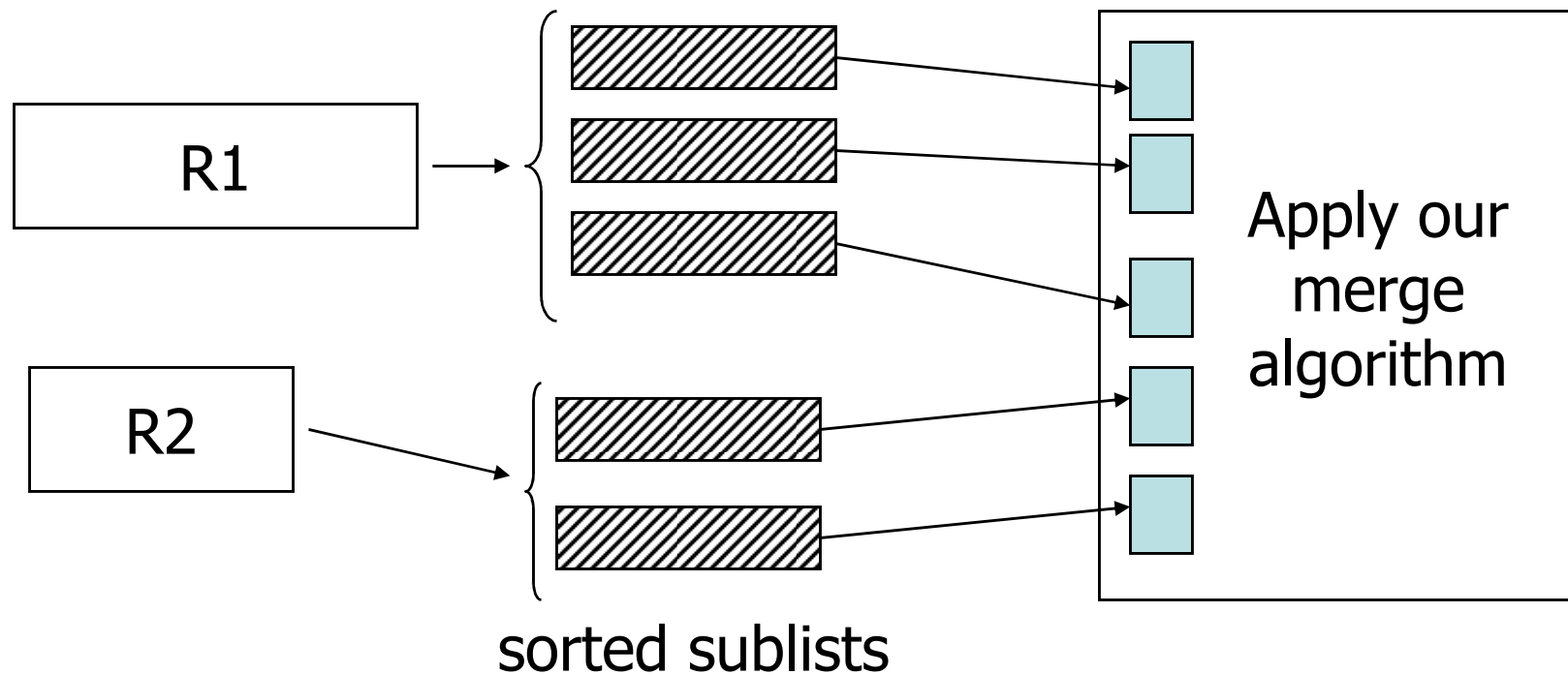
- NLJ cost =  $B(R) + B(R)B(S)/M-1 = O(B(R)B(S))$  [Quadratic]
- Sort-merge join cost =  $5 \times (B(R) + B(S)) = O(B(R) + B(S))$  [Linear]

# Can we Improve Sort-Merge Join?



Do we need to create the sorted R1, R2?

# A more “Efficient” Sort-Merge Join



# Analysis of the “Efficient” Sort-Merge Join

- Cost =  $3 \times (B(R) + B(S))$   
[Vs.  $5 \times (B(R) + B(S))$ ]
- Memory requirement:  
 $M \geq (B(R) + B(S))^{1/2}$   
[Vs.  $M \geq (\max(B(R), B(S)))^{1/2}$ ]

Another catch with the more “Efficient” version: Higher chances of **thrashing!**

## Cost of “Efficient” Sort-Merge join:

$$\begin{aligned} \text{Cost} &= \text{Read R1} + \text{Write R1 into sublists} \\ &\quad + \text{Read R2} + \text{Write R2 into sublists} \\ &\quad + \text{Read R1 and R2 sublists for Join} \\ &= 2000 + 1000 + 1500 = 4500 \end{aligned}$$

[Vs. 7500]

## Memory requirements in our Example

$$B(R1) = 1000 \text{ blocks}, 1000^{1/2} = 31.62$$

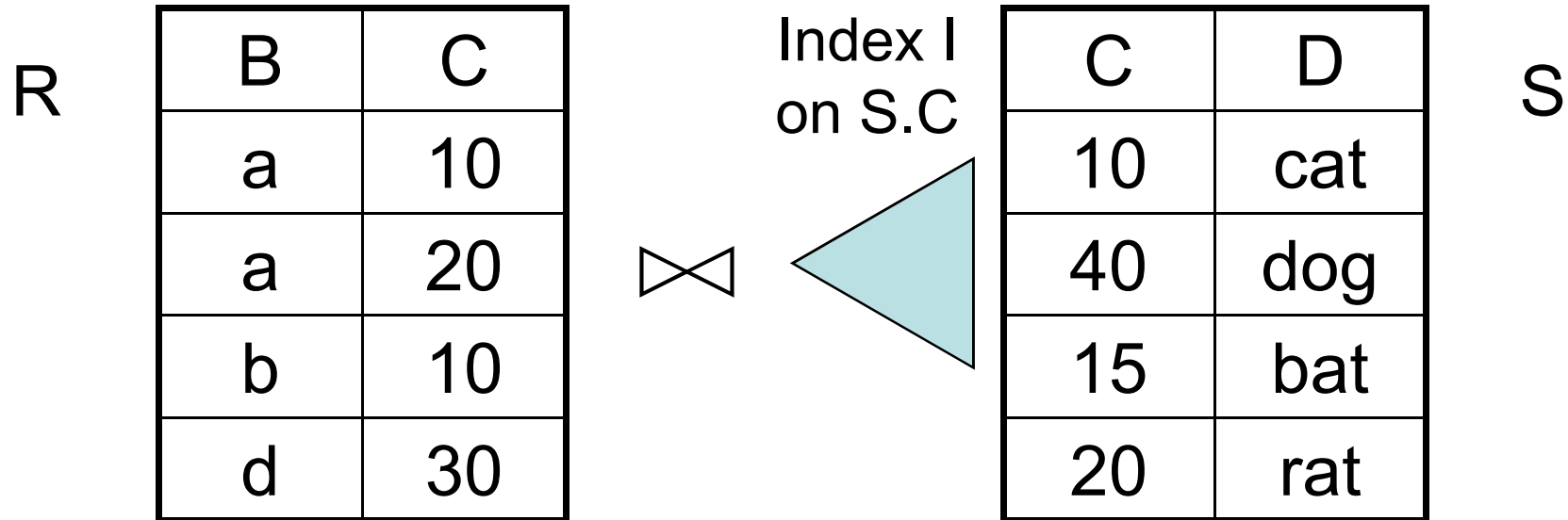
$$B(R2) = 500 \text{ blocks}, 500^{1/2} = 22.36$$

$$B(R1) + B(R2) = 1500, 1500^{1/2} = 38.7$$

$M > 32$  buffers for simple sort-merge join

$M > 39$  buffers for efficient sort-merge join

# Joins Using Existing Indexes



- Indexed NLJ (conceptually)

for each  $r \in R$  do

for each  $s \in S$  that matches **probe(I,r.C)** do

output r,s pair



# Continuing with our Running Example

- Assume R1.C index exists; 2 levels
- Assume R2 clustered, unordered
- Assume R1.C index fits in memory

Cost: R2 Reads: 500 IOs

for each R2 tuple:

- probe index - free
- if match, read R1 tuple

→ # R1 Reads depends on:

- # matching tuples
- clustering index or not

## What is expected # of matching tuples?

(a) say R1.C is key, R2.C is foreign key  
then expected = 1 tuple

(b) say  $V(R1,C) = 5000$ ,  $T(R1) = 10,000$   
with **uniform assumption**  
expect =  $10,000/5,000 = 2$

## What is expected # of matching tuples?

(c) Say  $\text{DOM}(R1, C) = 1,000,000$

$$T(R1) = 10,000$$

with assumption of **uniform distribution**  
**in domain**

$$\text{Expected} = \frac{10,000}{1,000,000} = \frac{1}{100} \text{ tuples}$$

## Total cost with Index Join with a Non-Clustering Index

(a) Total cost =  $500 + 5000(1) = 5,500$

(b) Total cost =  $500 + 5000(2) = 10,500$

(c) Total cost =  $500 + 5000(1/100) = 550$

Will any of these change if we have a clustering index?

## What if index does not fit in memory?

Example: say R1.C index is 201 blocks

- Keep root + 99 leaf nodes in memory
- Expected cost of each index access is

$$E = (0)\frac{99}{200} + (1)\frac{101}{200} \approx 0.5$$

Total cost (including Index Probes)

$$= 500+5000 \text{ [Probe + Get Records]}$$

$$= 500+5000 \text{ [0.5+2]}$$

$$= 500+12,500 = 13,000 \quad (\text{Case b})$$

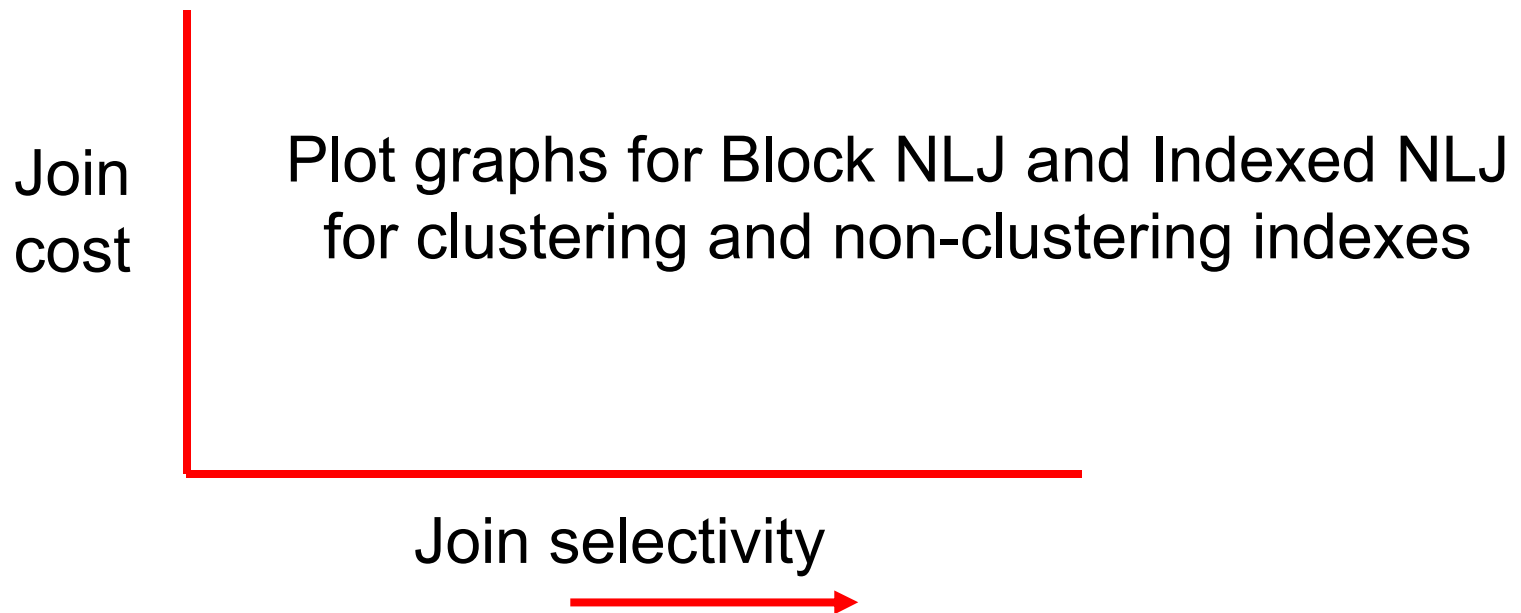
For **Case (c)**:

$$= 500+5000[0.5 \times 1 + (1/100) \times 1]$$

$$= 500+2500+50 = 3050 \text{ IOs}$$

# Block-Based NLJ Vs. Indexed NLJ

- Wrt #joining records
- Wrt index clustering





# Sort-Merge Join with Indexes

- Can avoid sorting
- Zig-zag join

# So far

not clustered

{	NLJ R2 ⋈ R1	55,000 (best)
	Merge Join	_____
	Sort+ Merge Join	_____
	R1.C Index	_____
	R2.C Index	_____

clustered

---

{	NLJ R2 ⋈ R1	5500
	Merge join	1500
	Sort+Merge Join	7500 → 4500
	R1.C Index	5500, 3050, 550
	R2.C Index	_____

# Building Indexes on the fly for Joins

- Hash join (conceptual)
  - Hash function  $h$ , range  $1 \rightarrow k$
  - Buckets for R1:  $G_1, G_2, \dots, G_k$
  - Buckets for R2:  $H_1, H_2, \dots, H_k$

## Algorithm

(1) Hash R1 tuples into  $G_1--G_k$

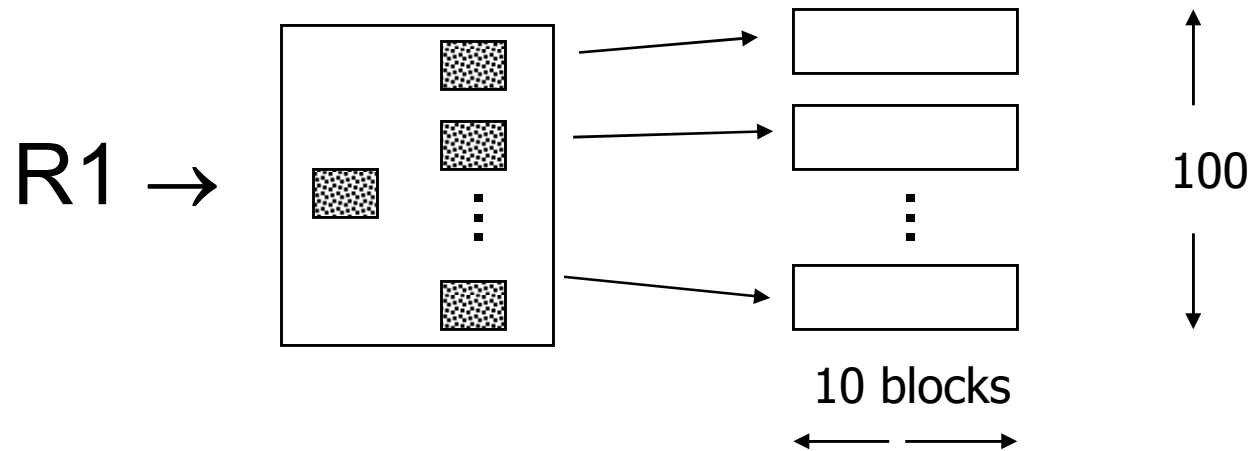
(2) Hash R2 tuples into  $H_1--H_k$

(3) For  $i = 1$  to  $k$  do

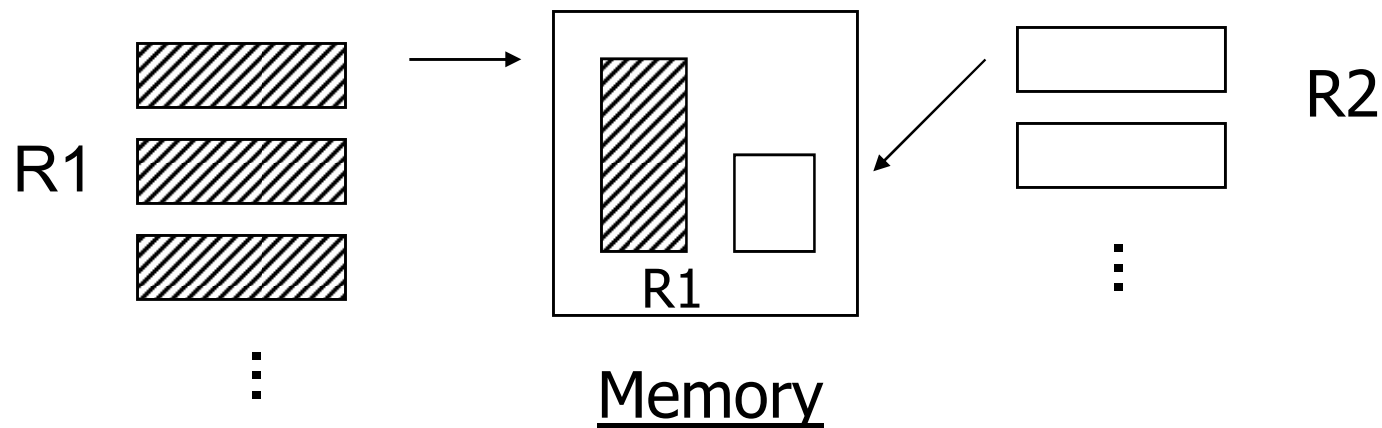
    Match tuples in  $G_i, H_i$  buckets

# Example Continued: Hash Join

- R1, R2 contiguous
- Use 100 buckets
- Read R1, hash, + write buckets



- > Same for R2
- > Read one R1 bucket; build memory hash table  
[R1 is called the **build** relation of the hash join]
- > Read corresponding R2 bucket + hash probe  
[R2 is called the **probe** relation of the hash join]



Then repeat for all buckets

## Cost:

"Bucketize:"      Read R1 + write

                            Read R2 + write

Join:                      Read R1, R2

$$\text{Total cost} = 3 \times [1000 + 500] = 4500$$

# Minimum Memory Requirements

Size of R1 bucket =  $(x/k)$

$k$  = number of buckets ( $k = M-1$ )

$x$  = number of R1 blocks

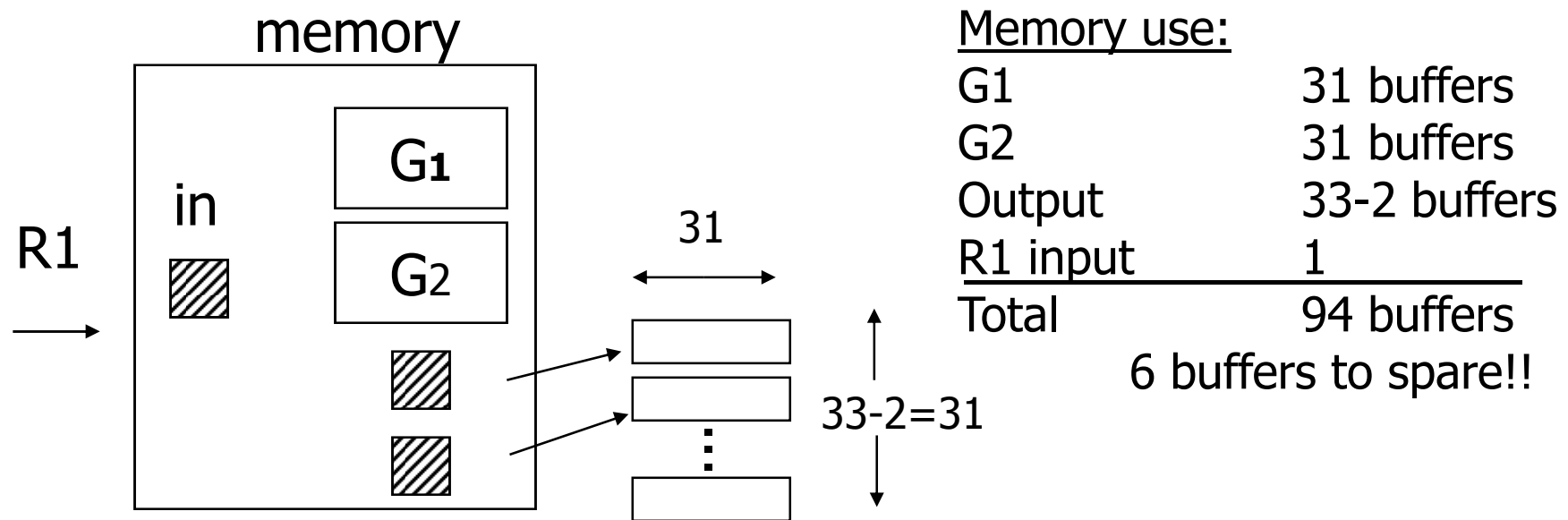
So...  $(x/k) \leq k \rightarrow k \geq \sqrt{x} \rightarrow M > \sqrt{x}$

Actually,  $M > \sqrt{\min(B(R), B(S))}$

[Vs.  $M > \sqrt{B(R)+B(S)}$  for Sort-Merge Join]

# Trick: keep some buckets in memory

E.g.,  $k'=33$  R1 buckets = 31 blocks  
keep 2 in memory

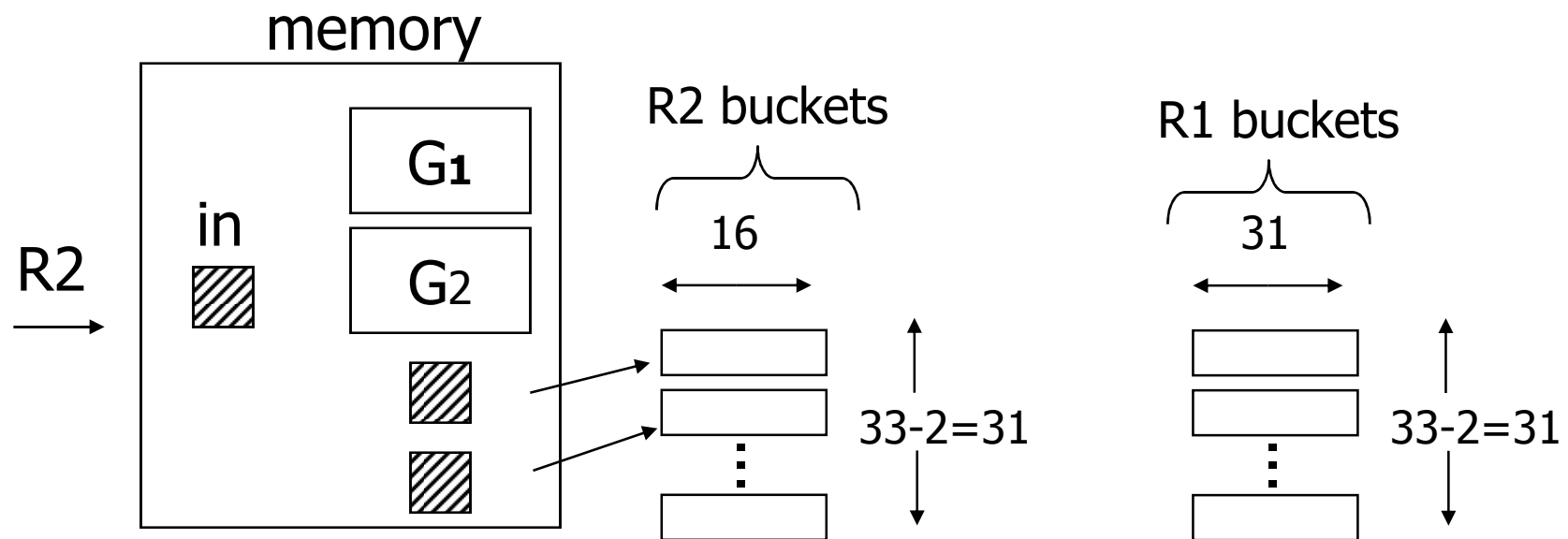


called **Hybrid Hash-Join**



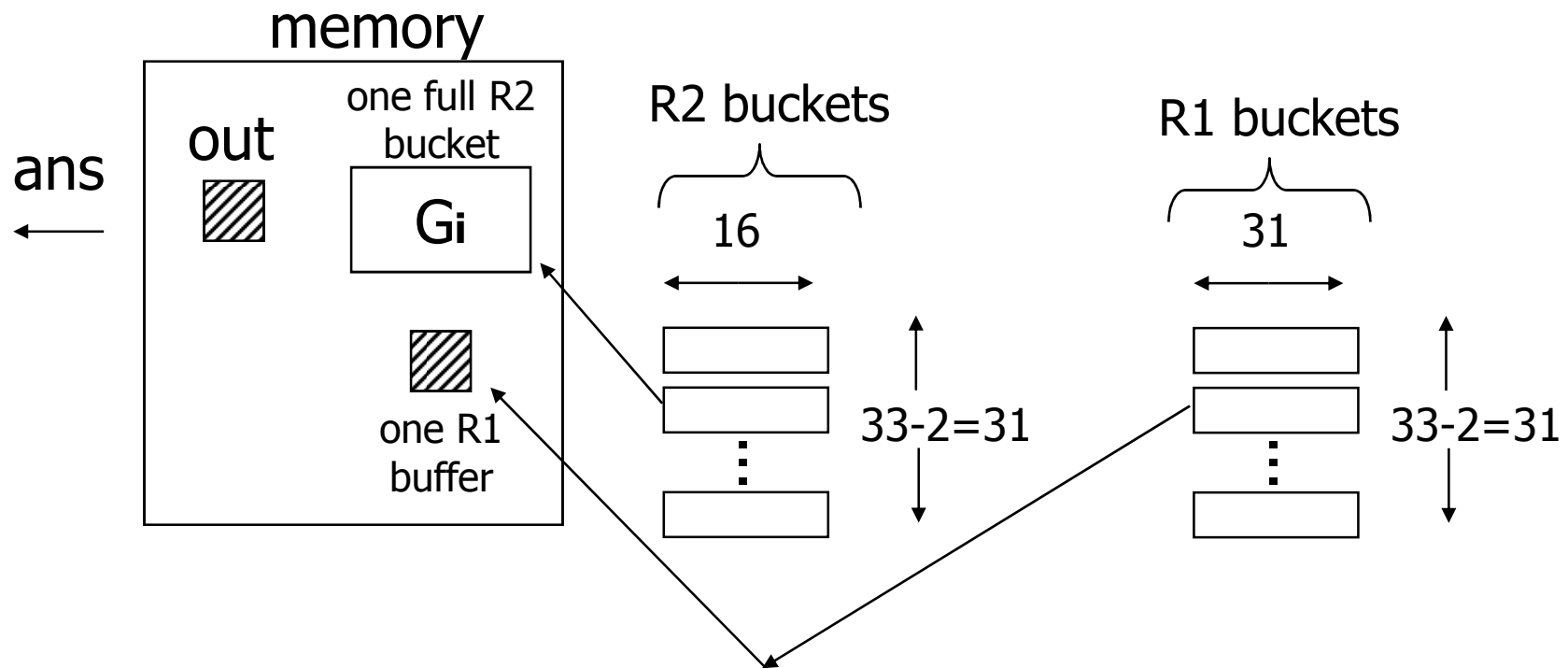
## Next: Bucketize R2

- R2 buckets =  $500/33 = 16$  blocks
- Two of the R2 buckets joined immediately with G1, G2



# Finally: Join remaining buckets

- for each bucket pair:
  - read one of the buckets into memory
  - join with second bucket

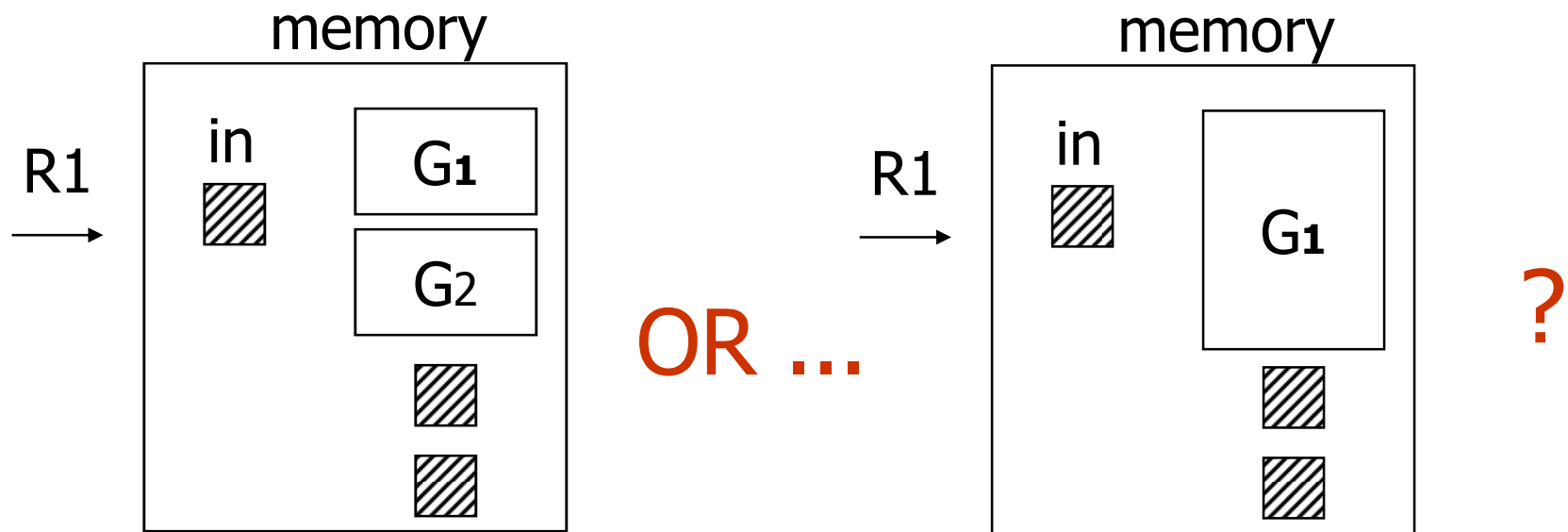


## Cost

- Bucketize R1 =  $1000+31\times 31=1961$
- To bucketize R2, only write 31 buckets:  
so, cost =  $500+31\times 16=996$
- To compare join (2 buckets already done)  
read  $31\times 31+31\times 16=1457$

Total cost =  $1961+996+1457 = 4414$

# How many Buckets in Memory?



☞ See textbook for an interesting answer ...

## Another hash join trick:

- Only write into buckets  
    <val,ptr> pairs
- When we get a match in join phase,  
    must fetch tuples

- To illustrate cost computation, assume:
  - 100 <val,ptr> pairs/block
  - expected number of result tuples is 100
- Build hash table for R2 in memory  
     5000 tuples  $\rightarrow$   $5000/100 = 50$  blocks
- Read R1 and match
- Read  $\sim$  100 R2 tuples

<u>Total cost</u> =	Read R2:	500
	Read R1:	1000
	Get tuples:	<u>100</u>
		1600

# So far:

clustered	NLJ	5500	
	Merge join	1500	
	Sort+merge joint	7500	
	R1.C index	5500	→ 550
	R2.C index	_____	
	Build R.C index	_____	
	Build S.C index	_____	
	Hash join	4500	
	with trick,R1 first	4414	
	with trick,R2 first	_____	
	Hash join, pointers	1600	

# Hash-based Vs. Sort-based Joins

- Some similarities (see textbook), some dissimilarities
- Non-equi joins
- Memory requirement
- Sort order may be useful later



# Summary

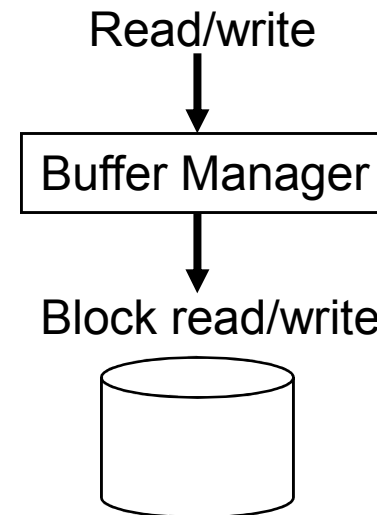
- **NLJ** ok for “small” relations  
(relative to memory size)
- For equi-join, where relations not sorted and no indexes exist,  
Hybrid Hash Join usually best

# Summary (contd.)

- **Sort-Merge Join** good for non-equi-join (e.g.,  $R1.C > R2.C$ )
- If relations already sorted, use **Merge Join**
- If index exists, it could be useful
  - Depends on expected result size and index clustering
- Join techniques apply to Union, Intersection, Difference

# Buffer Management

- DBMS Buffer Manager



- May control memory directly (i.e., does not allocate from virtual memory controlled by OS)

# Buffer Replacement Policies

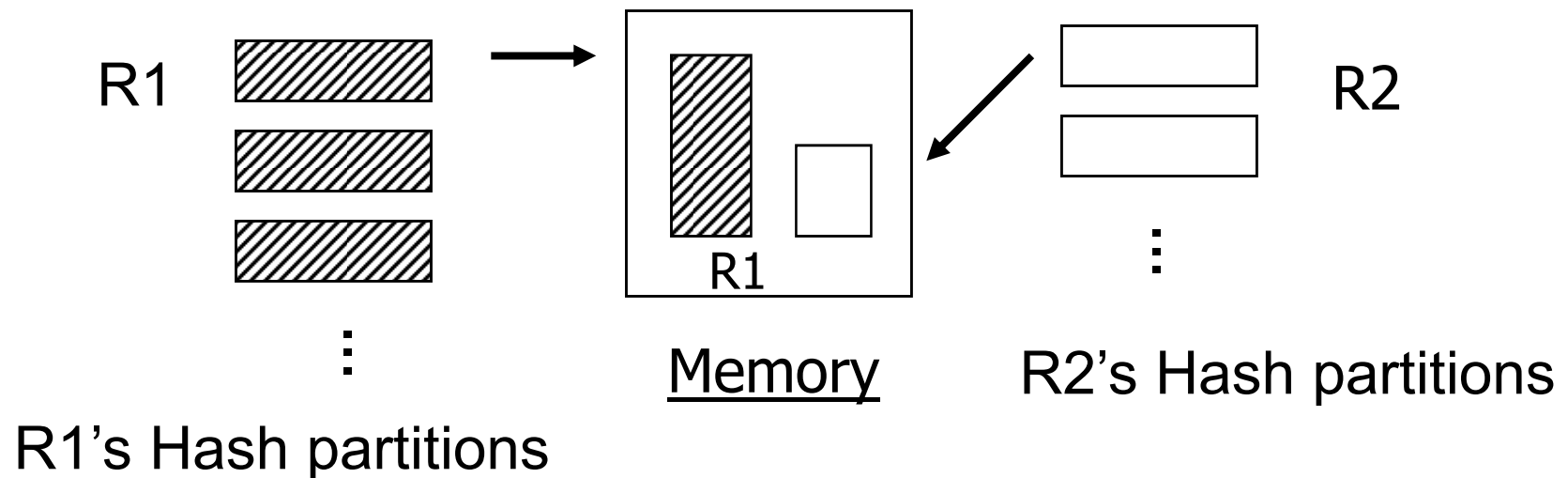
- Least Recently Used (LRU)
- Second-chance
- Most Recently Used (MRU)
- FIFO

# Interaction between Operators and Buffer Management

- Memory (our M parameter) may change while an operator is running
- Some operators can take advantage of specific buffer replacement policies
  - E.g., **Rocking** for Block-based NLJ

# Join Strategies for Parallel Processors

- May cover later if time permits
- We will see one example: Hash Join



# Textbook Material

- All of Chapter 15 except 15.8
  - 15.8 covers multi-pass sort and hash

# Roadmap

- A simple operator: Nested Loop Join
- Preliminaries
  - Cost model
  - Clustering
  - Operator classes
- Operator implementation (with examples from joins)
  - Scan-based
  - Sort-based
  - Using existing indexes
  - Hash-based
- Buffer Management
- Parallel Processing