# **CSE 344**

# FEBRUARY 16<sup>TH</sup> – DISK I/O AND ESTIMATION

# **ADMINISTRIVIA**

- HW6/OQ5 out after class
  - HW6 Due Wednesday, Feb 28th
  - OQ5 Due Friday, Feb 23<sup>rd</sup>
- Office hours for exam regrades
  - Additional HW5 OH on Wednesday

# HW6 AWS

- Making account
  - Use accurate information (matching academic records)
  - Create full account not 'starter code'
  - Be sure to terminate services when done



An additional file, that allows fast access to records in the data file given a search key

#### The index contains (key, value) pairs:

- The key = an attribute value (e.g., student ID or name)
- The value = a pointer to the record

**Could have many indexes for one table** 

# **KEYS IN INDEXING**

**Different keys:** 

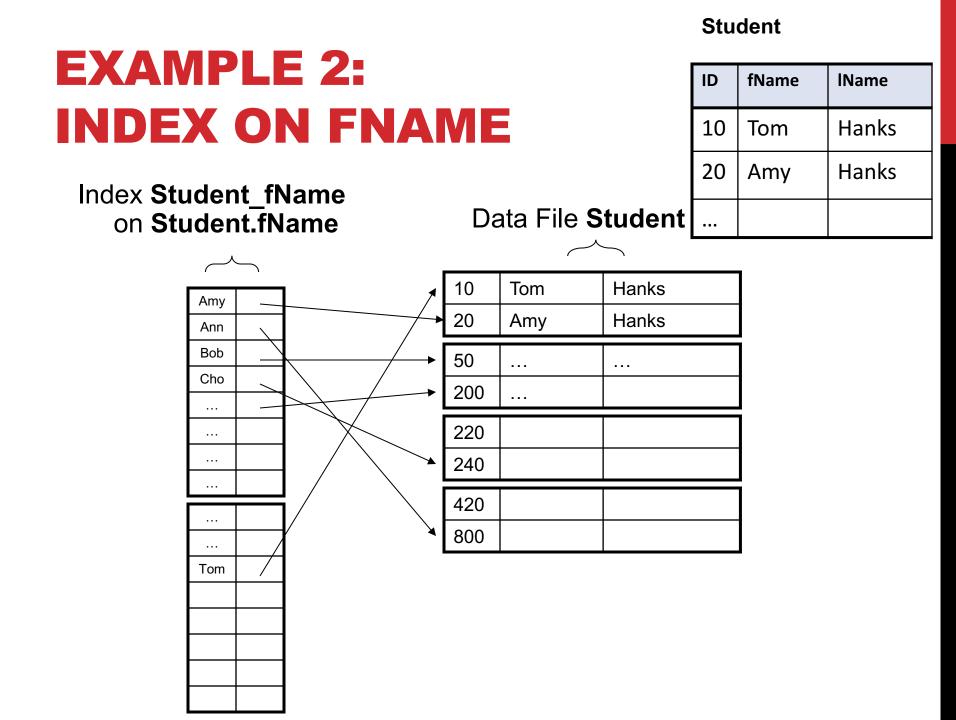
**Primary key** – uniquely identifies a tuple

Key of the sequential file – how the data file is sorted, if at all

**Index key** – how the index is organized

#### **EXAMPLE 1: IName** ID **fName INDEX ON ID** 10 Tom Hanks 20 Amy Hanks Data File **Student** .... Index Student\_ID on Student.ID 10 Tom Hanks 10 Hanks 20 Amy 20 50 50 . . . . . . 200 200 . . . 220 240 220 420 240 800 420 950 800 ...

Student



# **INDEX ORGANIZATION**

We need a way to represent indexes after loading into memory so that they can be used

Several ways to do this:

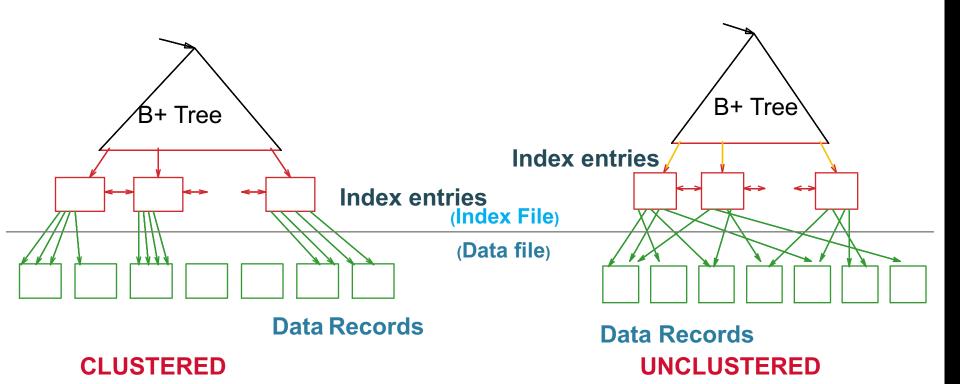
Hash table

#### **B+ trees – most popular**

- They are search trees, but they are not binary instead have higher fanout
- Will discuss them briefly next

Specialized indexes: bit maps, R-trees, inverted index

# CLUSTERED VS UNCLUSTERED



# INDEX CLASSIFICATION

### **Clustered/unclustered**

- Clustered = records close in index are close in data
  - Option 1: Data inside data file is sorted on disk
  - Option 2: Store data directly inside the index (no separate files)
- Unclustered = records close in index may be far in data

### **Primary/secondary**

- Meaning 1:
  - Primary = is over attributes that include the primary key
  - Secondary = otherwise
- Meaning 2: means the same as clustered/unclustered

### **Organization B+ tree or Hash table**

# SCANNING A DATA FILE

### **Disks are mechanical devices!**

- Technology from the 60s; density much higher now
- Read only at the rotation speed!



### Consequence: Sequential scan is MUCH FASTER than random reads

- Good: read blocks 1,2,3,4,5,...
- Bad: read blocks 2342, 11, 321,9, ...

### Rule of thumb:

 Random reading 1-2% of the file ≈ sequential scanning the entire file; this is decreasing over time (because of increased density of disks)

Solid state (SSD): \$\$\$ expensive; put indexes, other "hot" data there, still too expensive for everything

# **SUMMARY SO FAR**

# Index = a file that enables direct access to records in another data file

- B+ tree / Hash table
- Clustered/unclustered

### Data resides on disk

- Organized in blocks
- Sequential reads are efficint
- Random access less efficient
- Random read 1-2% of data worse than sequential

## **CREATING INDEXES IN SQL**

CREATE TABLE V(M int, N varchar(20), P int);

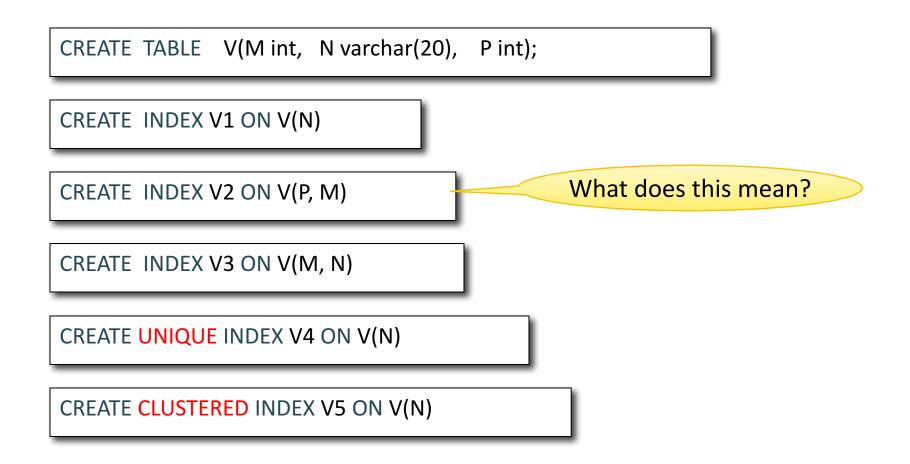
CREATE INDEX V1 ON V(N)

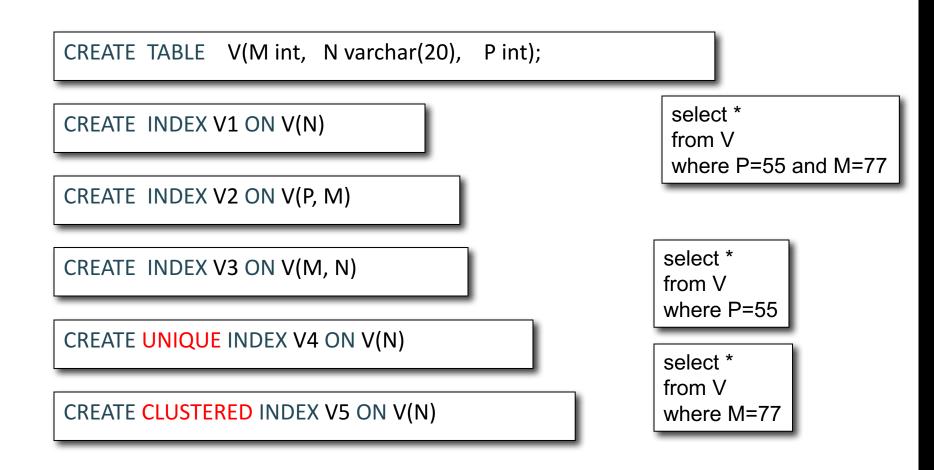
CREATE INDEX V2 ON V(P, M)

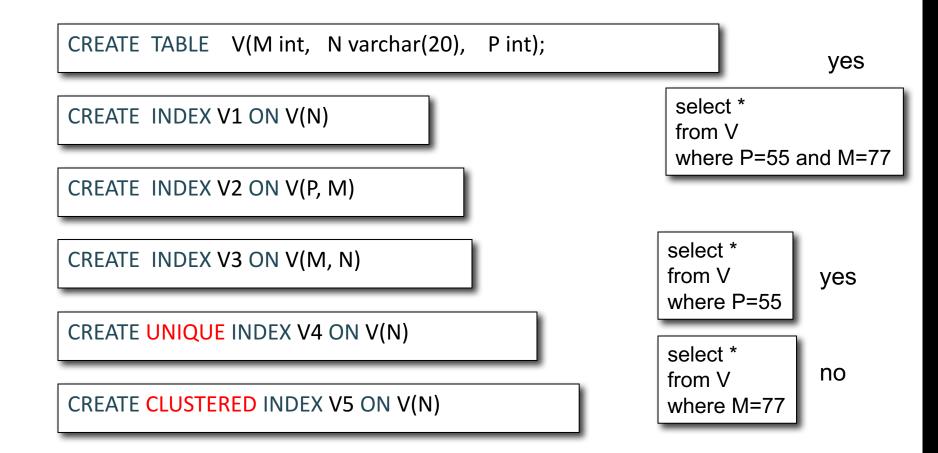
CREATE INDEX V3 ON V(M, N)

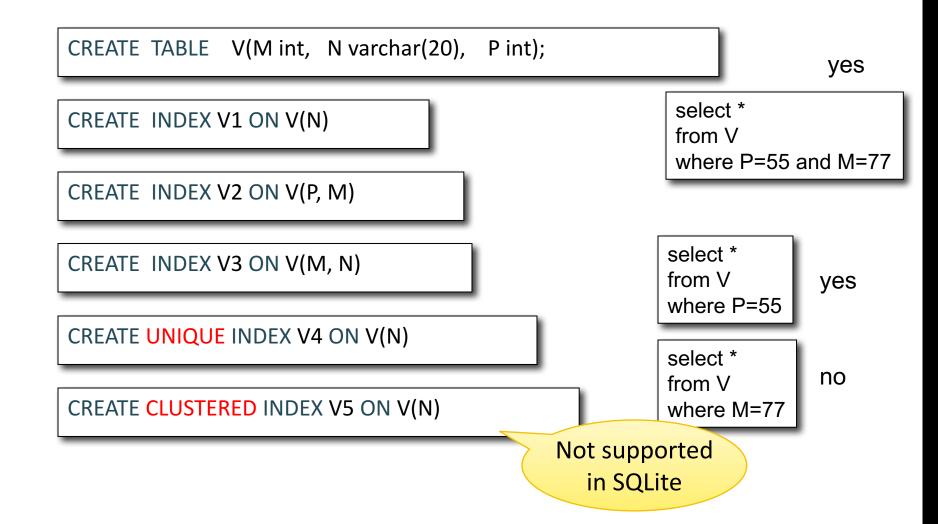
CREATE UNIQUE INDEX V4 ON V(N)

CREATE CLUSTERED INDEX V5 ON V(N)









#### Student

ID	fName	IName
10	Tom	Hanks
20	Amy	Hanks

# **WHICH INDEXES?**

#### The index selection problem

 Given a table, and a "workload" (big Java application with lots of SQL queries), decide which indexes to create (and which ones NOT to create!)

#### Who does index selection:

- The database administrator DBA
- Semi-automatically, using a database administration tool

## INDEX SELECTION: WHICH SEARCH KEY

Make some attribute K a search key if the WHERE clause contains:

- An exact match on K
- A range predicate on K
- A join on K

V(M, N, P);

Your workload is this

100000 queries:

SELECT \* FROM V WHERE N=? 100 queries:

SELECT \* FROM V WHERE P=?

V(M, N, P);

Your workload is this

100000 queries:

SELECT \* FROM V WHERE N=? 100 queries:

SELECT \* FROM V WHERE P=?

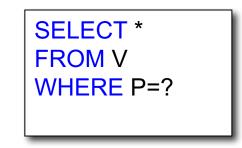
What indexes ?

V(M, N, P);

Your workload is this

100000 queries:

SELECT \* FROM V WHERE N=? 100 queries:



A: V(N) and V(P) (hash tables or B-trees)

V(M, N, P);

Your workload is this

100000 queries:

100 queries:

SELECT \* FROM V WHERE N>? and N<? SELECT \* FROM V WHERE P=? 100000 queries:

INSERT INTO V VALUES (?, ?, ?)

What indexes ?

V(M, N, P);

Your workload is this

100000 queries:

100 queries:

100000 queries:

SELECT \* FROM V WHERE N>? and N<? SELECT \* FROM V WHERE P=? INSERT INTO V VALUES (?, ?, ?)

### A: definitely V(N) (must B-tree); unsure about V(P)

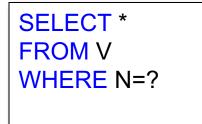
V(M, N, P);

Your workload is this

100000 queries:

1000000 queries:

100000 queries:



SELECT \* FROM V WHERE N=? and P>? INSERT INTO V VALUES (?, ?, ?)

### What indexes ?

V(M, N, P);

Your workload is this

100000 queries:

1000000 queries:

100000 queries:

SELECT \* FROM V WHERE N=?

SELECT \* FROM V WHERE N=? and P>? INSERT INTO V VALUES (?, ?, ?)



How does this index differ from:

- 1. Two indexes V(N) and V(P)?
- 2. An index V(P, N)?

V(M, N, P);

Your workload is this

1000 queries:

SELECT \* FROM V WHERE N>? and N<? 100000 queries:

SELECT \* FROM V WHERE P>? and P<?



V(M, N, P);

Your workload is this

1000 queries:

SELECT \* FROM V WHERE N>? and N<? 100000 queries:

SELECT \* FROM V WHERE P>? and P<?

A: V(N) secondary, V(P) primary index

# TWO TYPICAL KINDS OF QUERIES

SELECT \* FROM Movie WHERE year = ? • Point queries

• What data structure should be used for index?

SELECT \* FROM Movie WHERE year >= ? AND year <= ?

- Range queries
- What data structure should be used for index?

## BASIC INDEX SELECTION GUIDELINES

**Consider queries in workload in order of importance** 

### **Consider relations accessed by query**

• No point indexing other relations

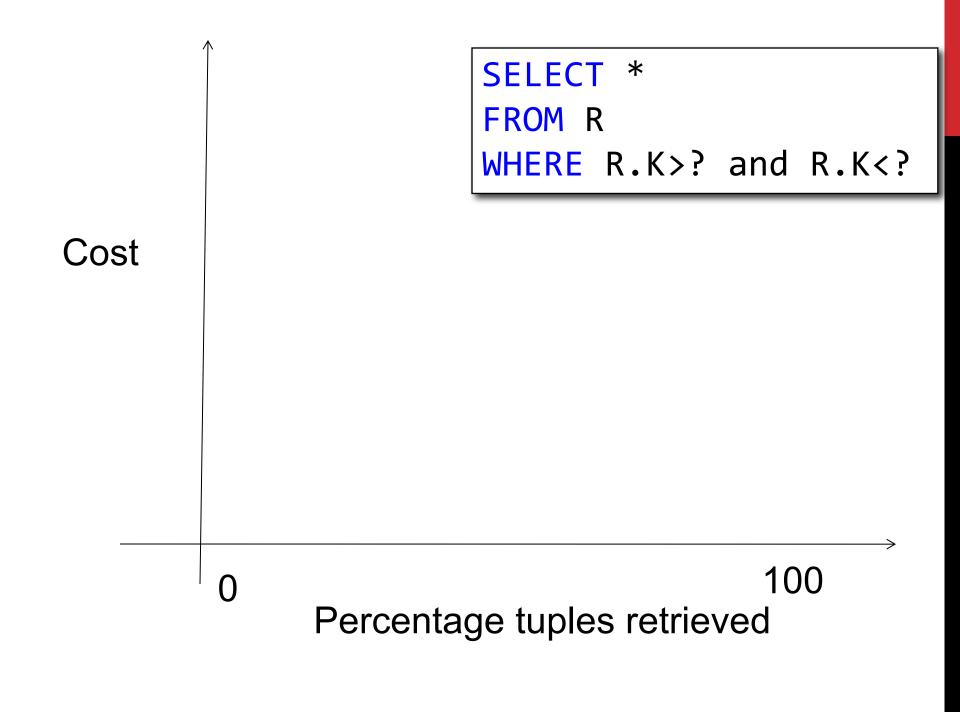
### Look at WHERE clause for possible search key

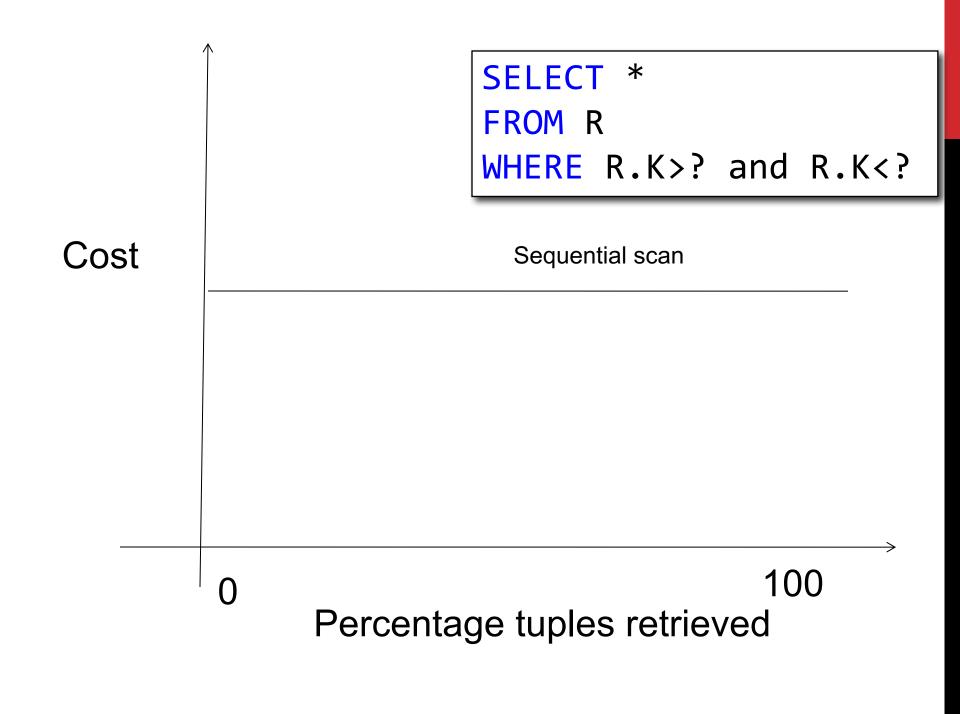
Try to choose indexes that speed-up multiple queries

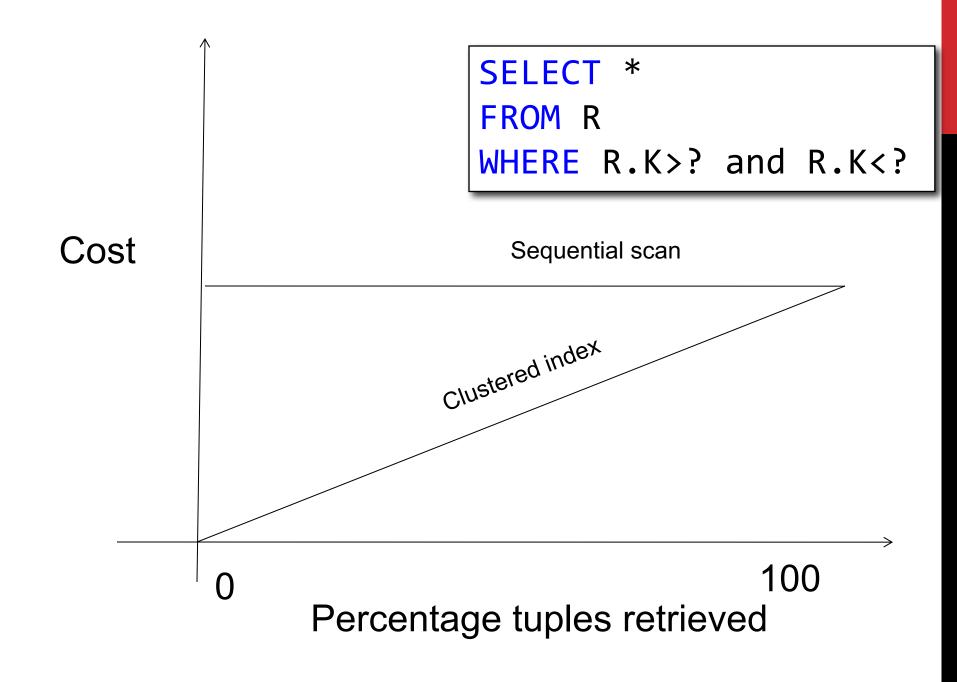
# **TO CLUSTER OR NOT**

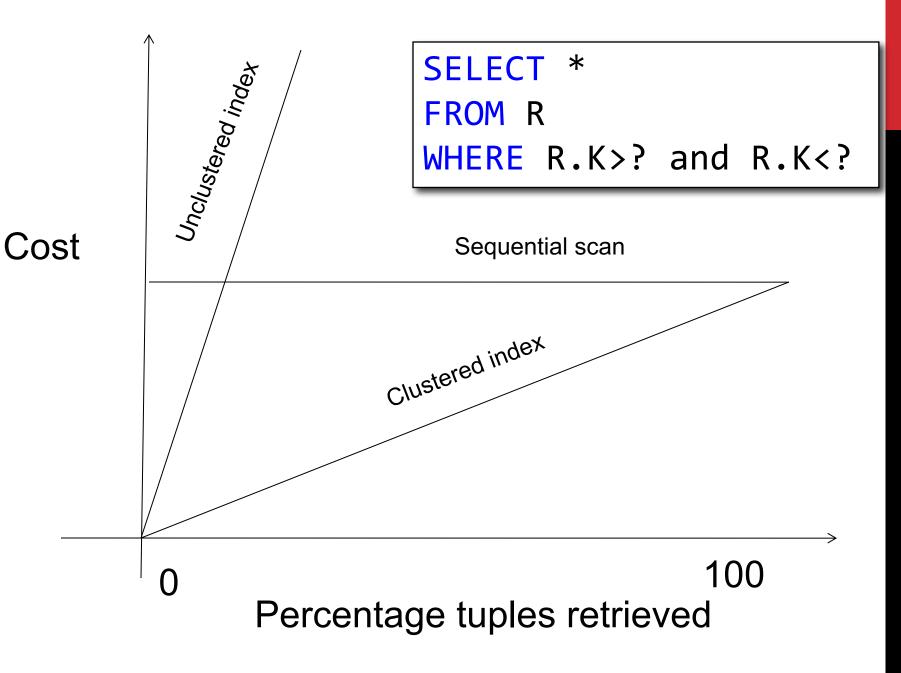
Range queries benefit mostly from clustering

Point indexes do *not* need to be clustered: they work equally well unclustered









# CHOOSING INDEX IS NOT ENOUGH

To estimate the cost of a query plan, we still need to consider other factors:

- How each operator is implemented
- The cost of each operator
- Let's start with the basics

## **COST PARAMETERS**

#### Cost = I/O + CPU + Network BW

• We will focus on I/O in this class

#### **Parameters (a.k.a. statistics):**

- **B(R)** = # of blocks (i.e., pages) for relation R
- T(R) = # of tuples in relation R
- V(R, a) = # of distinct values of attribute a

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When **a** is a key, **V(R,a) = T(R)** When **a** is not a key, **V(R,a)** can be anything <= **T(R)** 

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# DBMS collects statistics about base tables must infer them for intermediate results

#### **SELECTIVITY FACTORS FOR CONDITIONS**

$$A = c$$
 /\*  $\sigma_{A=c}(R)$  \*/

Selectivity = 1/V(R,A)

#### $A < C \qquad /* \sigma_{A < c}(R)*/$

Selectivity = (c - min(R, A))/(max(R,A) - min(R,A))

c1 < A < c2 /\*  $\sigma_{c1 < A < c2}(R)^*$ /

• Selectivity = (c2 - c1)/(max(R,A) - min(R,A))

## COST OF READING DATA FROM DISK

Sequential scan for relation R costs B(R)

#### Index-based selection

- Estimate selectivity factor **f** (see previous slide)
- Clustered index: f\*B(R)
- Unclustered index f\*T(R)

#### Note: we ignore I/O cost for index pages

Example:

B(R) = 2000 T(R) = 100,000 V(R, a) = 20

cost of 
$$\sigma_{a=v}(R) = ?$$

Table scan:

Example:

B(R) = 2000 T(R) = 100,000 V(R, a) = 20

cost of  $\sigma_{a=v}(R) = ?$ 

Table scan: B(R) = 2,000 I/Os

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Table scan: B(R) = 2,000 I/Os

- If index is clustered:
- If index is unclustered:

Example:

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cost of  $\sigma_{a=v}(R) = ?$ 

Table scan: B(R) = 2,000 I/Os

- If index is clustered: B(R) \* 1/V(R,a) = 100 I/Os
- If index is unclustered:

Example:

B(R) = 2000 T(R) = 100,000 V(R, a) = 20

cost of  $\sigma_{a=v}(R) = ?$ 

Table scan: B(R) = 2,000 I/Os

- If index is clustered: B(R) \* 1/V(R,a) = 100 I/Os
- If index is unclustered: T(R) \* 1/V(R,a) = 5,000 I/Os

Example:

B(R) = 2000 T(R) = 100,000 V(R, a) = 20

cost of  $\sigma_{a=v}(R) = ?$ 

Table scan: B(R) = 2,000 I/Os

#### Index based selection:

- If index is clustered: B(R) \* 1/V(R,a) = 100 I/Os
- If index is unclustered: T(R) \* 1/V(R,a) = 5,000 I/Os

Lesson: Don't build unclustered indexes when V(R,a) is small !



#### Join operator algorithms

- One-pass algorithms (Sec. 15.2 and 15.3)
- Index-based algorithms (Sec 15.6)

#### Note about readings:

- In class, we discuss only algorithms for joins
- Other operators are easier: read the book

## **JOIN ALGORITHMS**

Hash join

**Nested loop join** 

Sort-merge join

#### **HASH JOIN**

Hash join:  $\mathbf{R} \bowtie \mathbf{S}$ 

Scan R, build buckets in main memory

Then scan S and join

Cost: B(R) + B(S)

Which relation to build the hash table on?

#### **HASH JOIN**

Hash join:  $\mathbf{R} \bowtie \mathbf{S}$ 

Scan R, build buckets in main memory

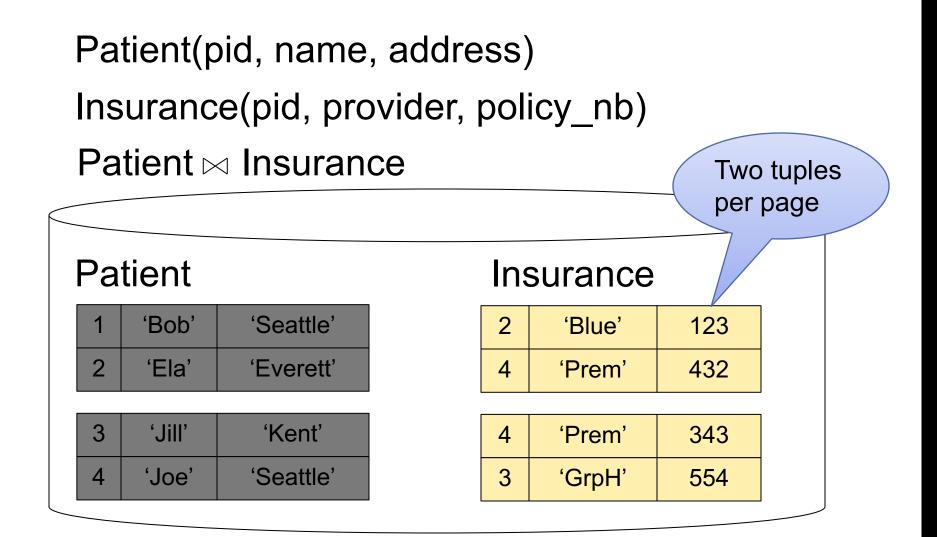
Then scan S and join

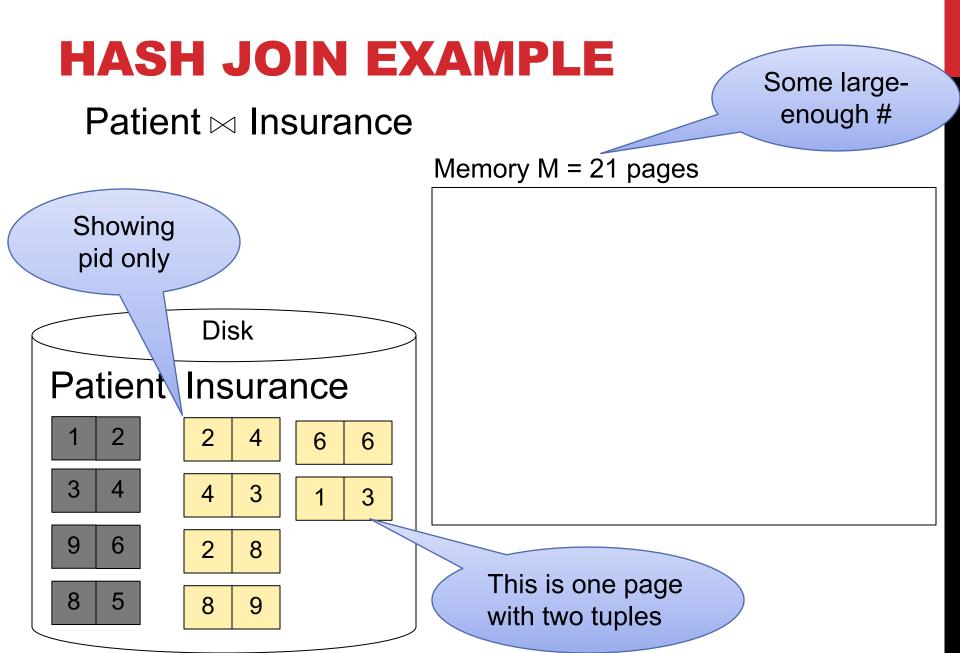
Cost: B(R) + B(S)

Which relation to build the hash table on?

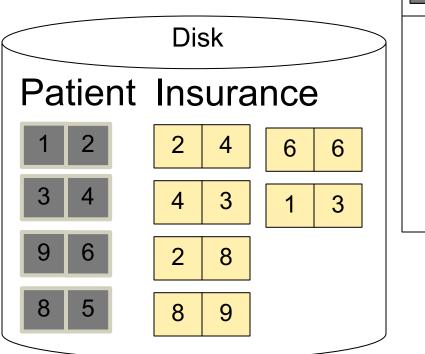
#### One-pass algorithm when $B(R) \le M$

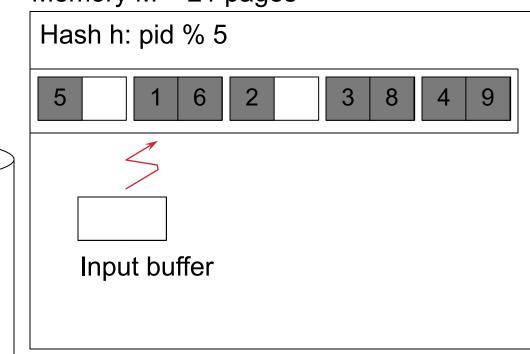
• M = number of memory pages available



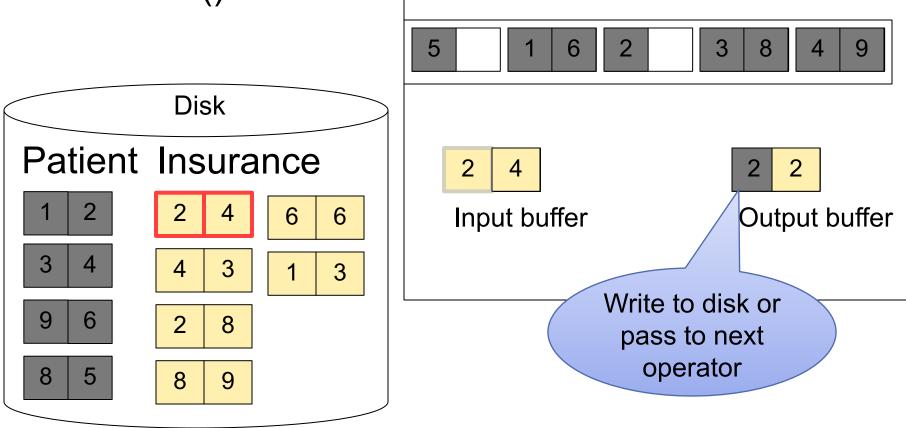


Step 1: Scan Patient and build hash table in memoryCan be done in<br/>method open()Memory M = 21 pagesHash h: pid % 5

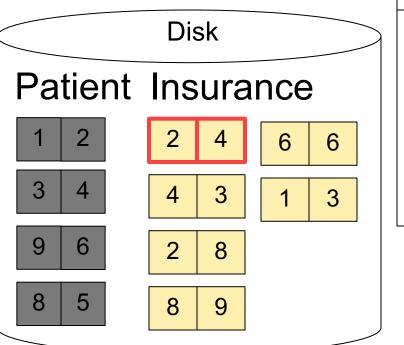


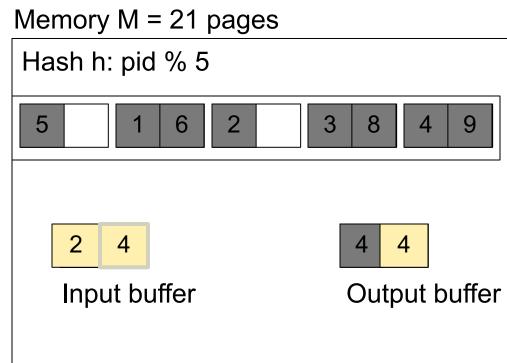


Step 2: Scan Insurance and probe into hash tableDone during<br/>calls to next()Memory M = 21 pagesHash h: pid % 5

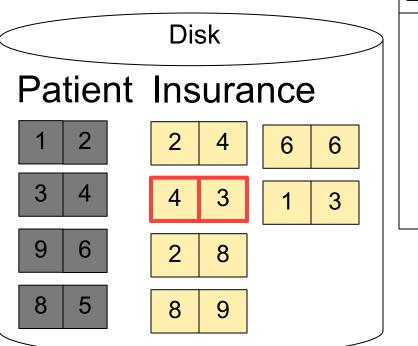


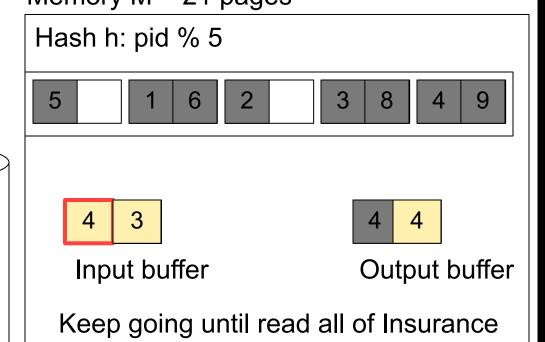
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Step 2: Scan Insurance and probe into hash tableDone during<br/>calls to next()Memory M = 21 pagesHash h: pid % 5





Cost: B(R) + B(S)

## **NESTED LOOP JOINS**

Tuple-based nested loop R 
Imple S

R is the outer relation, S is the inner relation

 $\begin{array}{l} \underline{\text{for}} \text{ each tuple } t_1 \text{ in } R \ \underline{\text{do}} \\ \underline{\text{for}} \text{ each tuple } t_2 \text{ in } S \ \underline{\text{do}} \\ \underline{\text{if}} \ t_1 \text{ and } t_2 \text{ join } \underline{\text{then}} \text{ output } (t_1, t_2) \end{array}$ 

## **NESTED LOOP JOINS**

Tuple-based nested loop R >> S

R is the outer relation, S is the inner relation

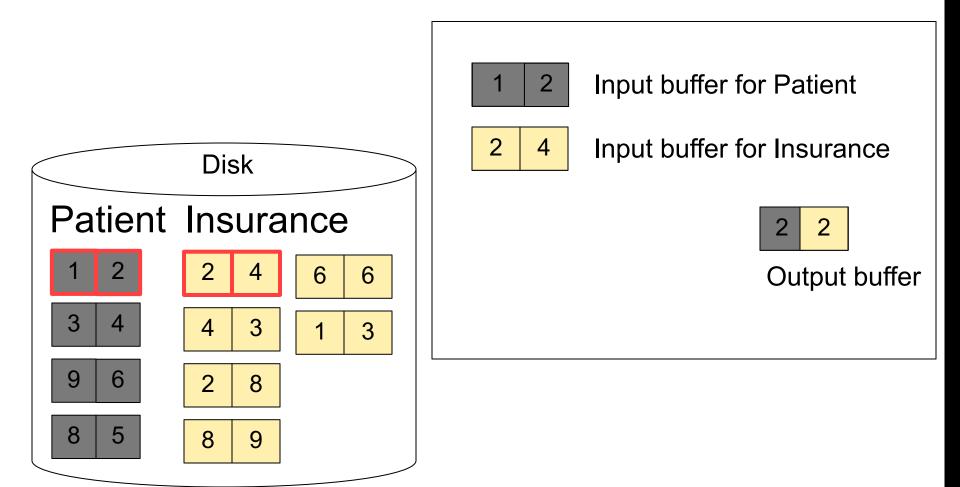
 $\begin{array}{l} \label{eq:total_for_second} \hline \begin{tabular}{l} for each tuple $t_1$ in $R$ $do$ \\ \hline \begin{tabular}{l} for each tuple $t_2$ in $S$ $do$ \\ \hline \begin{tabular}{l} if $t_1$ and $t_2$ join $\underline{then}$ output ($t_1,t_2$) \\ \hline \end{tabular} \end{array}$ 

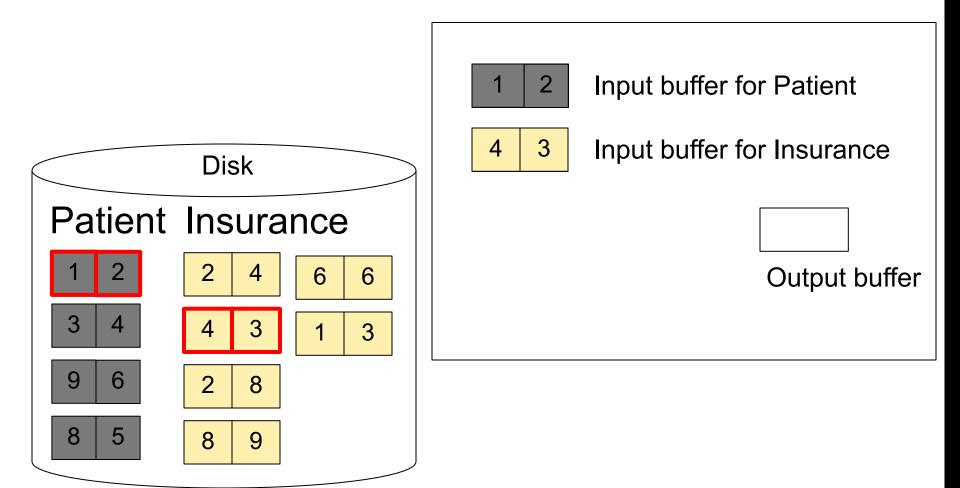
Cost: B(R) + T(R) B(S)

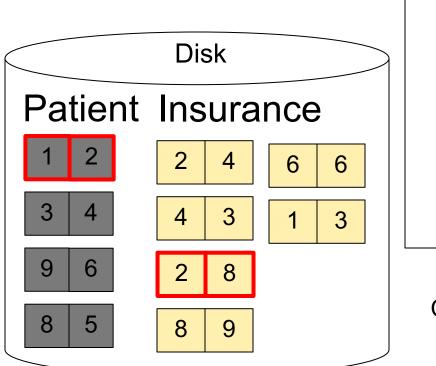
**Multiple-pass since S is read many times** 

 $\begin{array}{l} \label{eq:constraint} \begin{array}{l} \mbox{for each page of tuples r in R } \mbox{do} \\ \mbox{for each page of tuples s in S } \mbox{do} \\ \mbox{for all pairs of tuples } t_1 \mbox{ in r, } t_2 \mbox{ in s} \\ \mbox{if } t_1 \mbox{ and } t_2 \mbox{ join } \mbox{then} \mbox{ output } (t_1,t_2) \end{array}$ 

Cost: B(R) + B(R)B(S)









Input buffer for Patient



Input buffer for Insurance

Keep going until read all of Insurance



Output buffer

Then repeat for next page of Patient... until end of Patient

Cost: B(R) + B(R)B(S)

### **BLOCK-NESTED-LOOP REFINEMENT**

 $\begin{array}{l} \label{eq:starsest} \begin{array}{l} \mbox{for each group of M-1 pages r in R } \mbox{do} \\ \mbox{for each page of tuples s in S } \mbox{do} \\ \mbox{for all pairs of tuples } t_1 \mbox{ in r, } t_2 \mbox{ in s} \\ \mbox{if } t_1 \mbox{ and } t_2 \mbox{ join } \mbox{then} \mbox{ output } (t_1,t_2) \end{array}$ 

Cost: B(R) + B(R)B(S)/(M-1)

#### **SORT-MERGE JOIN**

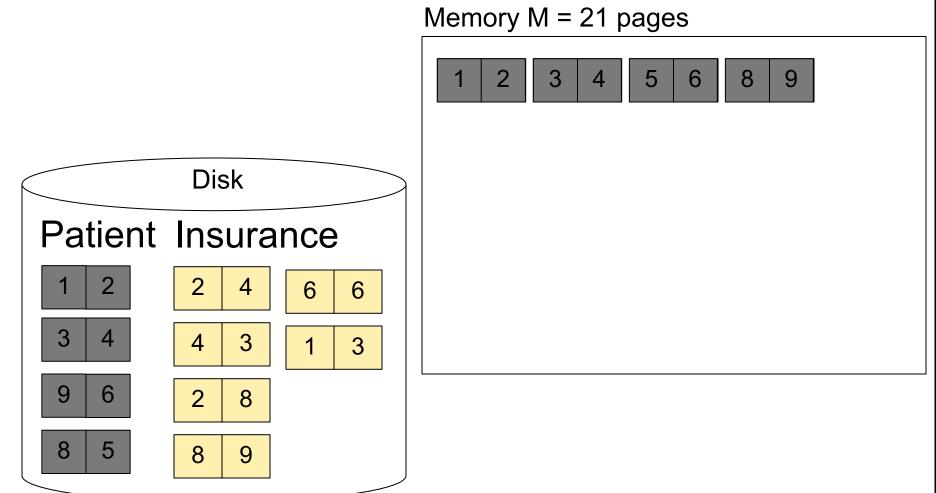
Sort-merge join: R ⋈ S Scan R and sort in main memory Scan S and sort in main memory Merge R and S

Cost: B(R) + B(S)

One pass algorithm when B(S) + B(R) <= M

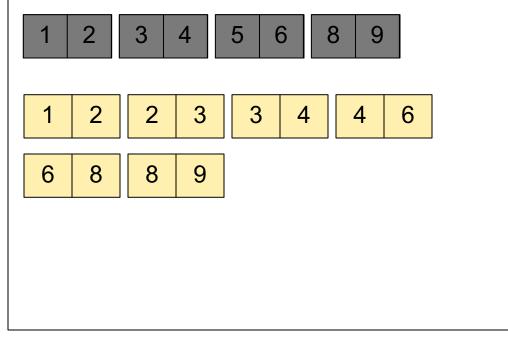
Typically, this is NOT a one pass algorithm

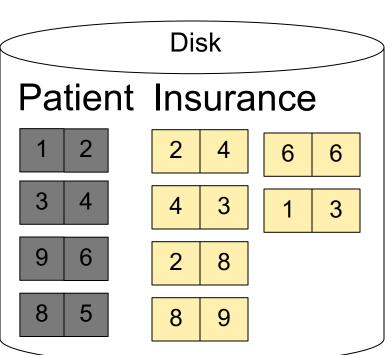
#### Step 1: Scan Patient and sort in memory



Step 2: Scan Insurance and sort in memory

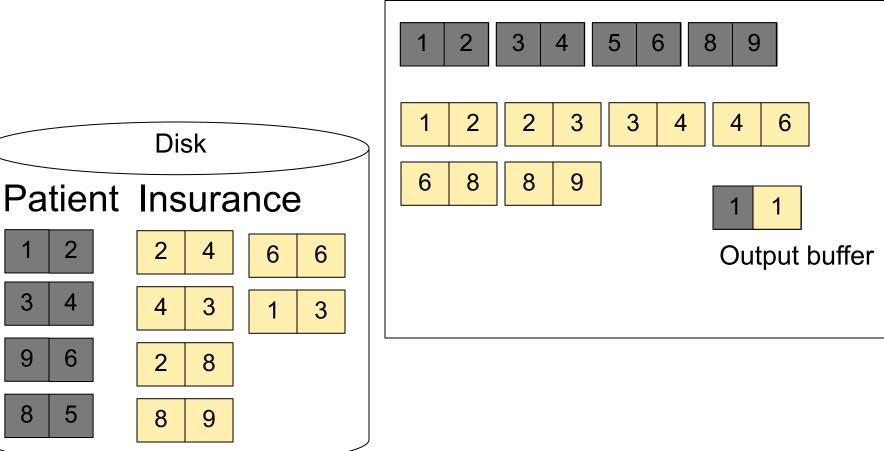
Memory M = 21 pages



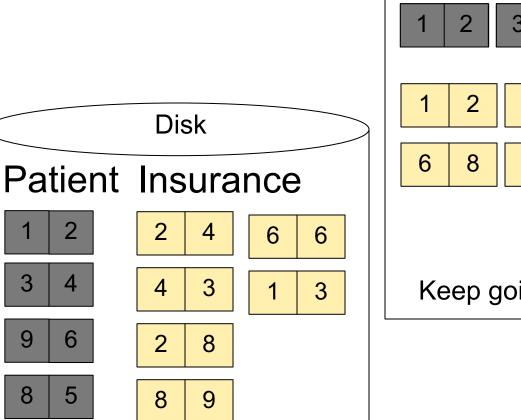


#### Step 3: Merge Patient and Insurance

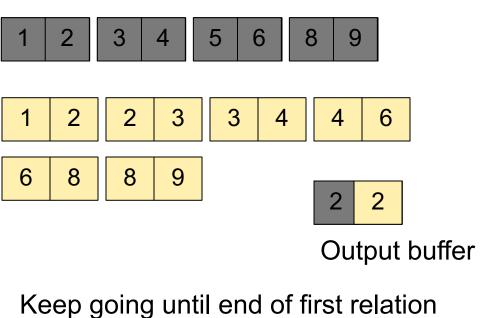
Memory M = 21 pages



#### Step 3: Merge Patient and Insurance



Memory M = 21 pages



## INDEX NESTED LOOP JOIN

 $\mathbf{R} \bowtie \mathbf{S}$ 

Assume S has an index on the join attribute

Iterate over R, for each tuple fetch corresponding tuple(s) from S

#### Cost:

If index on S is clustered:
 B(R) + T(R) \* (B(S) \* 1/V(S,a))

If index on S is unclustered:
 B(R) + T(R) \* (T(S) \* 1/V(S,a))