# Introduction to Database Systems IDBS – Fall 2024

- Week 9:
- Storage Models
- Architecture of a DBMS
- Main-Memory DBMSs

**Martin Hentschel** 

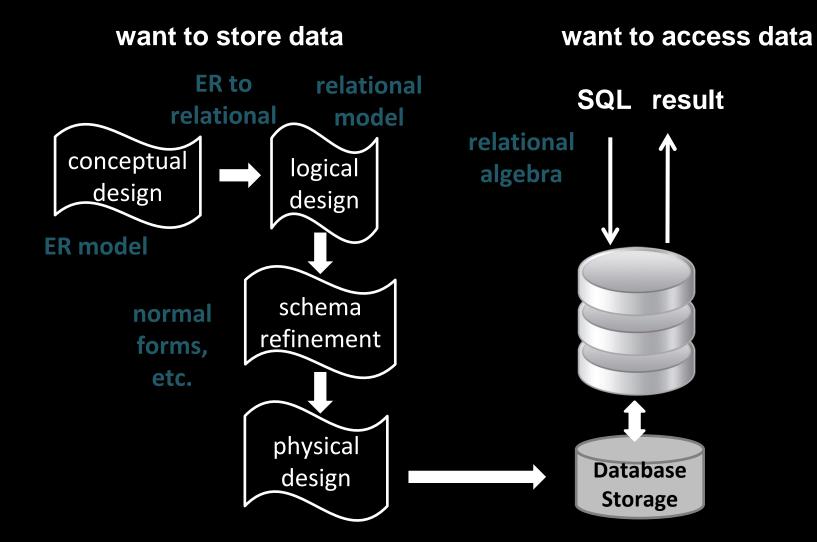
Slide credits: Eleni Tzirita Zacharatou

### Readings:

PDBM 2 + more material on LearnIT



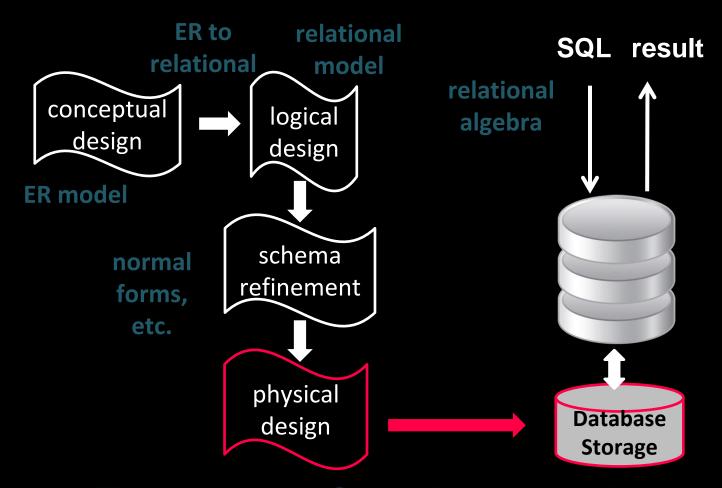
# The Bigger Picture



# The Bigger Picture

want to store data

want to access data



Today we will talk about the lowest physical representation of data in a database



### **Database Workloads**

### **On-Line Transaction Processing (OLTP)**

→ Fast operations that only read/update a small amount of data each time

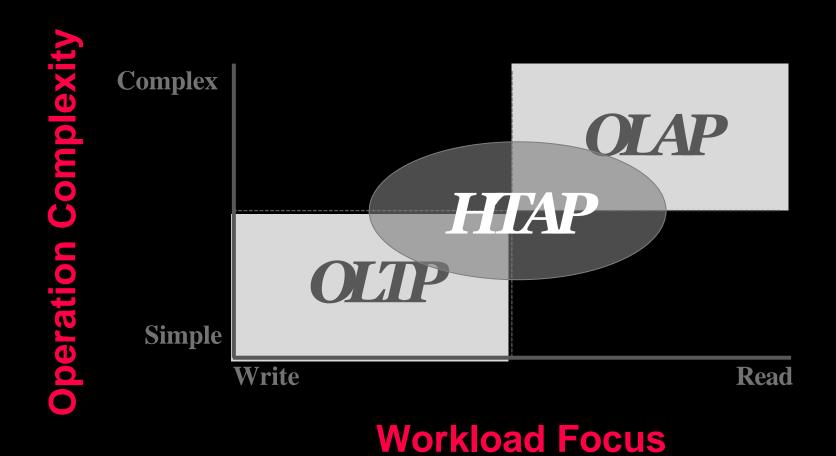
### On-Line Analytical Processing (OLAP)

→ Complex queries that read a lot of data to compute aggregates

### **Hybrid Transaction + Analytical Processing (HTAP)**

→ OLTP + OLAP together on the same database instance

# **Database Workloads**



### **OLTP**

### **On-Line Transaction Processing:**

→ Simple queries that read/update a small amount of data that is related to a single entity in the database.

This is usually the kind of application that people build first.

```
SELECT P.*, R.*
  FROM pages AS P
  INNER JOIN revisions AS R
  ON P.latest = R.revID
  WHERE P.pageID = ?
```

```
UPDATE useracct
  SET lastLogin = NOW(),
    hostname = ?
WHERE userID = ?
```

```
INSERT INTO revisions
VALUES(?,?...,?)
```

### **OLAP**

### **On-Line Analytical Processing:**

→ Complex queries that read large portions of the database spanning multiple entities.

You execute these workloads on the data you have collected from your OLTP application(s).

```
SELECT COUNT(U.lastLogin), EXTRACT(month FROM
    U.lastLogin) AS month
FROM useracct AS U
WHERE U.hostname LIKE '%.gov'
GROUP BY EXTRACT(month FROM U.lastLogin)
```

### Exercise

### Consider the following relation:

• Sales (ProductID, ProductName, Date, Quantity, Revenue)

### **Query A**

# SELECT ProductID, SUM(Quantity) AS TotalQuantitySold FROM Sales WHERE Date BETWEEN '2023-01-01' AND '2023-12-31' GROUP BY ProductID;

### **Query B**

```
UPDATE Sales
SET Revenue = Revenue * 1.1
WHERE ProductID = 12345 AND
Date = '2023-03-10';
```

### Observation

The relational model *does not* specify that the DBMS must store all a tuple's attributes together on a single page.

This *may not* be the best layout for some workloads.

# Storage Models

A DBMS's <u>storage model</u> specifies how it physically organizes tuples on disk and in memory.

- → Can have different performance characteristics based on the target workload (OLTP vs. OLAP).
- → Influences the design choices of the rest of the DBMS.

**Choice #1: N-ary Storage Model (NSM)** 

**Choice #2: Decomposition Storage Model (DSM)** 

Choice #3: Hybrid Storage Model (PAX)

# N-ary Storage Model (NSM)

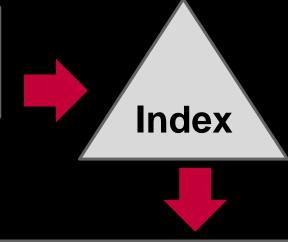
The DBMS stores (almost) all attributes for a single tuple contiguously on a single page.

→ Also known as a "row store"

Ideal for OLTP workloads where queries are more likely to access individual entities and execute write-heavy workloads.

# NSM: OLTP Example

```
SELECT * FROM useracct
WHERE userName = ?
AND userPass = ?
```

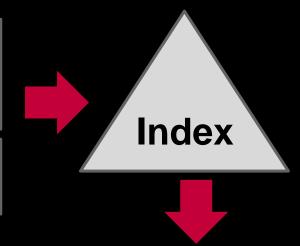




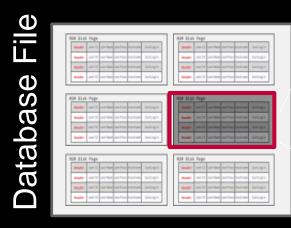
NSM Disk Page								
header	userID	userName	userPass	hostname	lastLogin			
header	userID	userName	userPass	hostname	lastLogin			
header	userID	userName	userPass	hostname	lastLogin			
header	-	-	-	-	-			

# NSM: OLTP Example

```
SELECT * FROM useracct
WHERE userName = ?
  AND userPass = ?
INSERT INTO useracct
VALUES (?,?,...?)
```







N	NSM Disk Page									
	header	userID	userName	userPass	hostname	lastLogin				
	header	userID	userName	userPass	hostname	lastLogin				
	header	userID	userName	userPass	hostname	lastLogin				
	header	userID	userName	userPass	hostname	lastLogin				

# NSM: OLAP Example

```
SELECT COUNT(U.lastLogin),
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	NSM Disk Page								
	header	userID	userName	userPass	hostname	lastLogin			
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# NSM: OLAP Example

```
SELECT COUNT (U.lastLogin),

EXTRACT (month FROM U.lastLogin) AS month
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N	NSM Disk Page									
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	header	userID	userName	userPass	hostname	lastLogin				
	header	userID	userName	userPass	hostname	lastLogin				
<u> </u>										

### Reflection

- Do you see a problem in the previous OLAP example?
- Is NSM a good choice for OLAP queries?
- Can you think of some drawbacks of NSM?

# NSM: OLAP Example

```
SELECT COUNT (U.lastLogin),

EXTRACT (month FROM U.lastLogin) AS month

FROM useracct AS U

WHERE U.hostname LIKE '%.gov'

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NSM Disk Page									
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header	userID	userName	userPass	hostname	lastLogin				
header	userID	userName	userPass	hostname	lastLogin				
header	userID	userName	userPass	hostname	lastLogin				

**Useless Data** 

# **NSM:** Summary

### **Advantages**

- → Fast inserts, updates, and deletes.
- → Good for queries that need the entire tuple (OLTP).
- → Can use index-oriented physical storage for clustering.

### **Disadvantages**

- → Not good for scanning large portions of the table and/or a subset of the attributes.
- → Terrible memory locality in access patterns.
- → Not ideal for compression because of multiple value domains within a single page.

# Decomposition Storage Model (DSM)

The DBMS stores a single attribute for all tuples contiguously in a block of data.

→ Also known as a "column store"

Ideal for OLAP workloads where read-only queries perform large scans over a subset of the table's attributes.

DBMS is responsible for combining/splitting a tuple's attributes when reading/writing.

# DSM: Database Example

The DBMS stores the values of a single attribute across multiple tuples contiguously in a page.

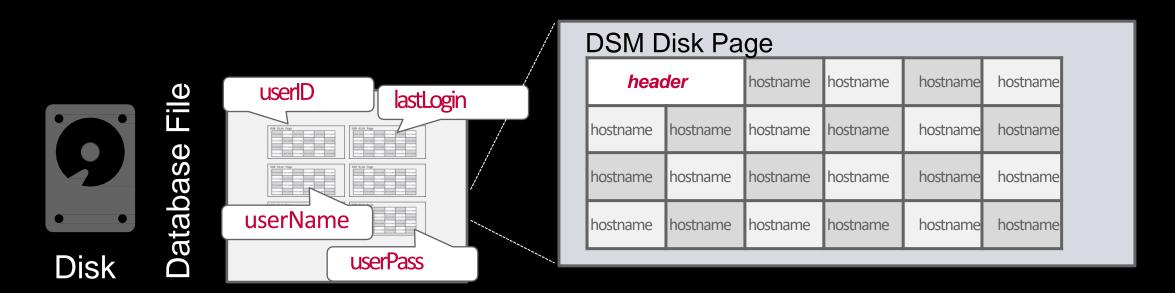
→ Also known as a "column store".

header	userID	userName	userPass	hostname	lastLogin
header	userID	userName	userPass	hostname	lastLogin
header	userID	userName	userPass	hostname	lastLogin
header	userID	userName	userPass	hostname	lastLogin

# DSM: Database Example

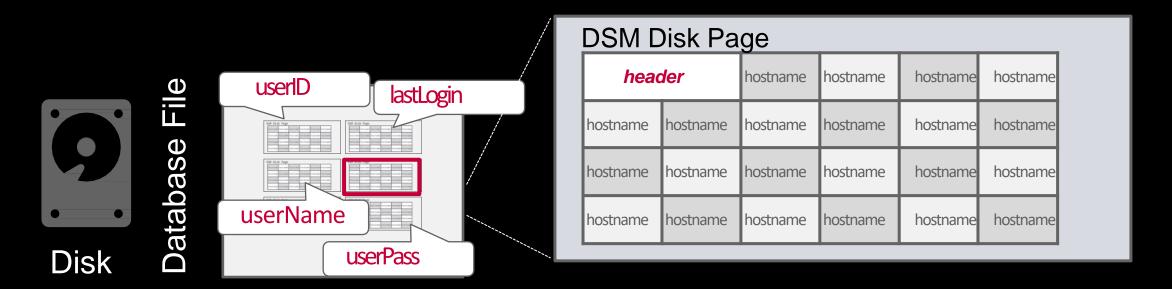
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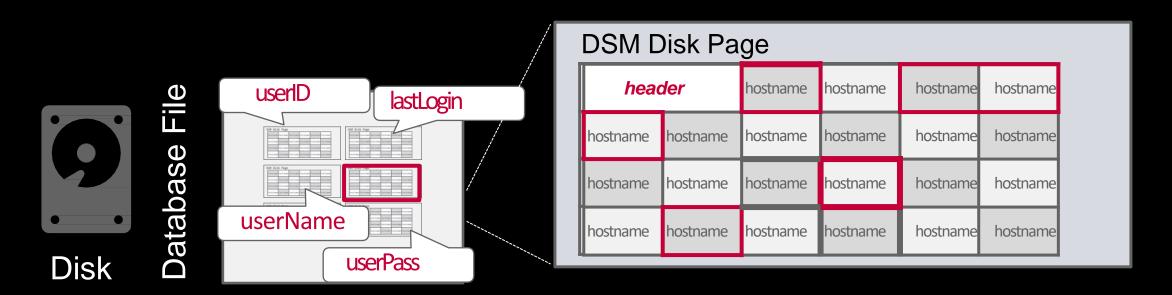
# DSM: OLAP Example

```
SELECT COUNT (U.lastLogin),
EXTRACT (month FROM U.lastLogin) AS month
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WHERE U.hostname LIKE '%.gov'
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```



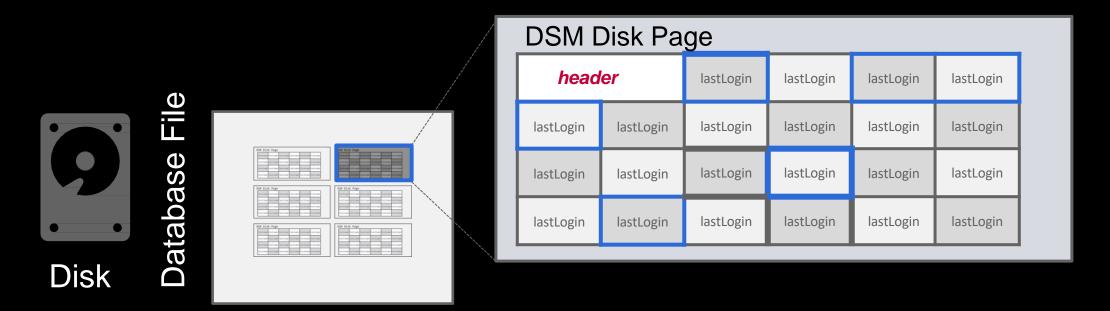
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SELECT COUNT (U.lastLogin),
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# DSM: OLAP Example

```
SELECT COUNT (U.lastLogin),
EXTRACT (month FROM U.lastLogin) AS month
FROM useracct AS U
WHERE U.hostname LIKE '%.gov'
GROUP BY EXTRACT (month FROM U.lastLogin)
```



### Reflection

Is DSM a good choice for OLTP queries?

```
SELECT * FROM useracct
WHERE userName = ?
AND userPass = ?
```

Can you think of some drawbacks of DSM?

# **DSM: Summary**

### **Advantages**

- → Reduces the amount wasted I/O per query because the DBMS only reads the data that it needs.
- → Faster query processing because of increased locality and cached data reuse.
- → Better data compression.

### **Disadvantages**

→ Slow for point queries, inserts, updates, and deletes because of tuple splitting/stitching/reorganization.

# Optimal Layout

NSM

**DSM** 

easy to implement

good if you are accessing the whole tuple or most columns of a tuple have tuple reconstruction cost

good if you are accessing one or a few columns of a tuple

### Choice depends on the workload!

# Systems in the Wild

### row stores





### column stores







# Systems in the Wild

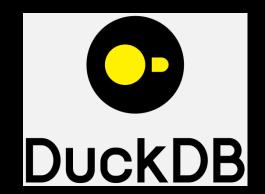
big vendors offering different products with both row & column stores

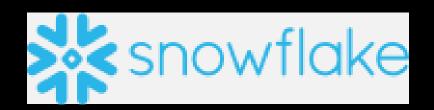






### mix of row & column stores







### **DBMS** Architecture

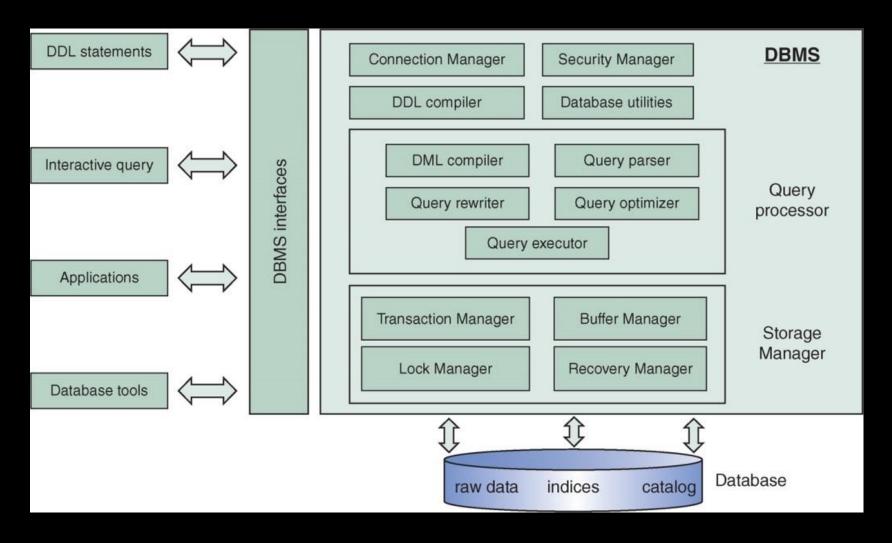
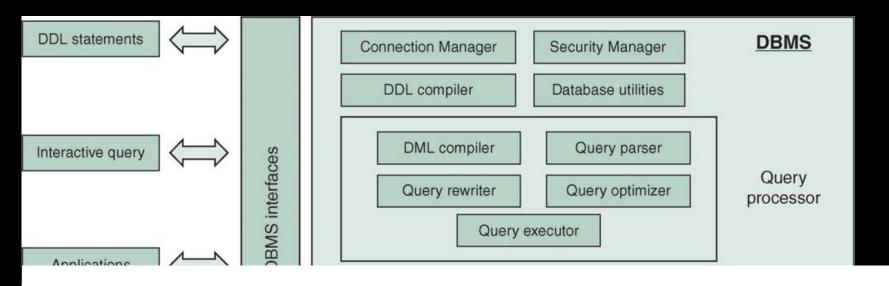


Figure 2.1

# Query Processor

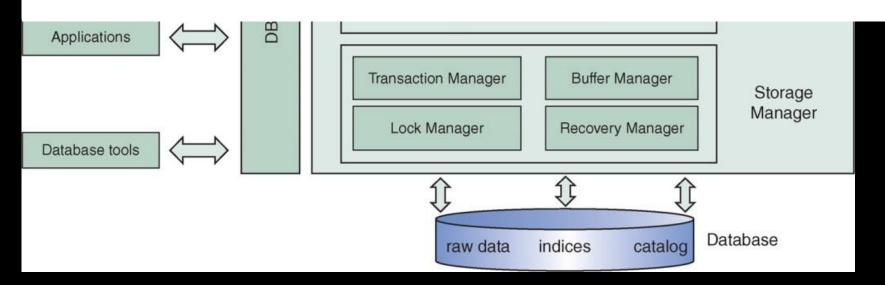


parsing an SQL query into a plan optimizing that plan executing the optimized plan using relational operators (e.g., join, group by, etc.)

# Storage Manager

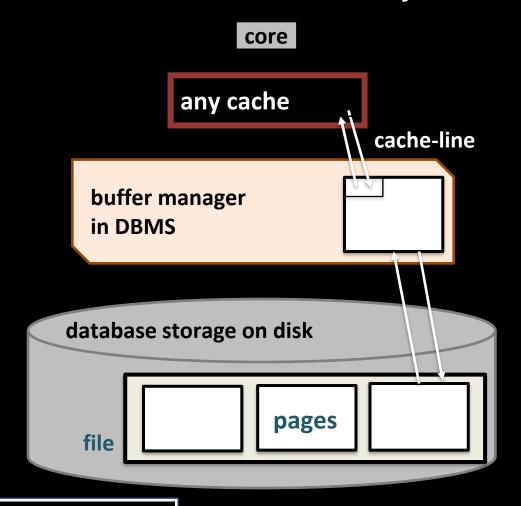
manages & keeps track of the data read from persistent storage into main memory:

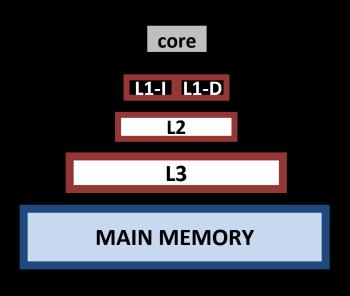
- better optimizations because of application knowledge, e.g., replacement policy, prefetching
- you may want to force-flush some data to disk (e.g., log)
- you may not want to ever flush some data (e.g., aborted requests)



# Data Movement through Storage Hierarchy

for a traditional database system







# What if data fits in-memory?

### Traditional System

multicore CPU

caches

buffer manager

main memory



### **Main-Memory System**

- query compilation that generates more efficient code
- no buffer manager
- data organized for better cache utilization/accesses
- no disk use during query execution
- lightweight logging for recovery

# Main-Memory DBMSs

- What if all data is in RAM?
  - No disk while executing tasks → never need to wait for disk
  - *Note:* There is still disk for persistent storage!
- Don't have to optimize data accesses for disk
- → no buffer manager
  - → no need for tuples in slotted-pages in-memory, have direct access to tuples
  - → indexes aren't kept in pages either & usually aren't persisted
- → aim better cache utilization/accesses
  - → row-store for OLTP, column-store for OLAP
  - → cache-conscious index structures
  - → query compilation that generates cache-conscious code

also valid for disk

### Main-Memory DBMSs: Downsides

- Startups/restarts are expensive
  - need to load everything in-memory
  - indexes has to be rebuilt
  - recovery takes longer if not done carefully

- There comes a day where your data doesn't fit in memory anymore
  - most main-memory systems added support for efficiently accessing cold/old data from disk later
    - → because customers ask for it

# Take Away

- It is important to choose the right storage model for the target workload:
  - OLTP = Row Store
  - OLAP = Column Store
- Goal: maximize sequential access, minimize unnecessary data reads.
- Main-memory DBMS
  - No blocking for disk I/O
  - No buffer management/buffer pool
  - Optimizing for cache accesses
  - Non-blocking concurrency, lightweight logging