

# Lesson 9 -> Partitioning vs Sharding

## 1 Partitioning vs Sharding (Clear Distinction)

### Partitioning

- **Logical division of data**
- Usually **within the same database system**
- May or may not be on different machines

Think: “How is data split inside a DB?”

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### Sharding

- **Physical distribution of data across multiple databases / nodes**
- Always implies **horizontal partitioning**
- Used when a single DB cannot handle scale

Think: “Which database node holds this data?”

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### Simple Analogy

- Partitioning → rooms inside one building
  - Sharding → multiple buildings
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## 2 Database Partitioning (Techniques)

### 2.1 Horizontal Partitioning (Row-based)

#### What it is

- Rows are split across partitions
- Same schema in each partition

Example:

```
orders_2024 orders_2025
```

Each partition contains different rows.

## Pros

- ✓ Scales reads & writes
- ✓ Queries scan less data

## Cons

- ✗ Cross-partition joins expensive
  - ✗ Partition management complexity
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## 2.2 Vertical Partitioning (Column-based)

### What it is

- Columns split into separate tables
- Based on access patterns

Example:

```
users_basic (id, name) users_profile (id, address, preferences)
```

## Pros

- ✓ Faster reads for common queries
- ✓ Smaller row size

## Cons

- ✗ Joins needed for full data
  - ✗ Schema complexity
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## 2.3 Range-Based Partitioning

### What it is

- Data partitioned based on value ranges

Example:

```
orders: Jan-Mar → P1 Apr-Jun → P2
```

## Pros

- ✓ Very efficient range queries
- ✓ Good for time-series data

## Cons

- ✗ Hot partitions
- ✗ Manual rebalancing

Used heavily in:

- Retail orders
  - Logs
  - ML features by time
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## 2.4 Hash-Based Partitioning

### What it is

- Hash function decides partition

Example:

```
hash(user_id) % 4 → partition
```

### Pros

- ✓ Even data distribution
- ✓ Avoids hot partitions

### Cons

- ✗ Poor range queries
  - ✗ Rehashing pain when adding partitions
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## 2.5 Key-Based Partitioning

### What it is

- Partitioning based on a natural key

Example:

```
user_id → partition
```

Often implemented via hashing.

### Pros

- ✓ Predictable routing
- ✓ Simple logic

## Cons

- ✗ Key skew causes imbalance
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## 2.6 Composite Partitioning

### What it is

- Combine multiple strategies

Example:

- Range by date
- Hash by user\_id inside range

Used in **high-scale retail & ML systems**.

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## 3 Database Sharding (Techniques)

Sharding is **horizontal partitioning + distribution**.

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### 3.1 Range-Based Sharding

#### How it works

Shard 1 → user\_id 1–1M    Shard 2 → user\_id 1M–2M

#### Pros

- ✓ Simple routing
- ✓ Range queries efficient

#### Cons

- ✗ Hot shards
  - ✗ Rebalancing difficult
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## 3.2 Hash-Based Sharding

### How it works

```
hash(user_id) % N → shard
```

### Pros

- ✓ Uniform load
- ✓ Easy scale (initially)

### Cons

- ✗ Resharding is painful
  - ✗ Range queries inefficient
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## 3.3 Consistent Hashing (Important)

### How it works

- Hash ring
- Minimal key movement when adding/removing shards

### Pros

- ✓ Scales well
- ✓ Minimal data reshuffle

### Cons

- ✗ Complex to implement
- ✗ Debugging harder

Used in:

- Cassandra
  - DynamoDB
  - Redis Cluster
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## 3.4 Directory-Based Sharding

### How it works

- Lookup service maps key → shard

Example:

```
user_id → shard_3
```

## Pros

- ✓ Flexible
- ✓ Easy rebalancing

## Cons

- ✗ Directory is SPOF
  - ✗ Extra network hop
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## 3.5 Geo-Based Sharding

### How it works

- Shards based on geography

Example:

- India → shard IN
- US → shard US

## Pros

- ✓ Low latency
- ✓ Compliance friendly

## Cons

- ✗ Cross-region queries hard

Used by:

- Global retail platforms
  - Recommendation systems
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## 4 Partitioning vs Sharding (Side-by-Side)

Aspect	Partitioning	Sharding
Scope	Logical	Physical

Aspect	Partitioning	Sharding
Machines	Same or multiple	Multiple
Scaling	Limited	High
Complexity	Medium	High
Failure impact	Local	Larger

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## 5 ML & Retail-Specific Examples

### Feature Store

- Partition by date
- Shard by entity\_id

### User Embeddings

- Hash-based sharding
- Consistent hashing

### Online Inference Cache

- Redis cluster with hash slots

### Training Data

- Range partitioned by time
- Sharded by region

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## 6 Failure & Edge Cases (Architect Thinking)

### Hot Partitions

- Skewed keys
- Celebrity users

Solution:

- Salting keys
  - Adaptive sharding
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## Cross-Shard Queries

- Aggregations across shards

Solution:

- Fan-out queries
  - Pre-aggregations
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## Rebalancing

- Adding/removing nodes

Solution:

- Consistent hashing
  - Background migration
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## 7 Interview-Ready One-Liners

- Partitioning organizes data inside a database; sharding distributes it across databases.
  - Sharding always implies horizontal partitioning.
  - Hash-based sharding favors uniform load; range-based favors query efficiency.
  - Consistent hashing minimizes data movement during scaling.
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## How This Fits Your Growth Path

Understanding this means:

- You can design **feature stores**
- You can scale **online inference**
- You can discuss **data-heavy system tradeoffs**

This is **ML Architect / Head of DS-level knowledge**.