

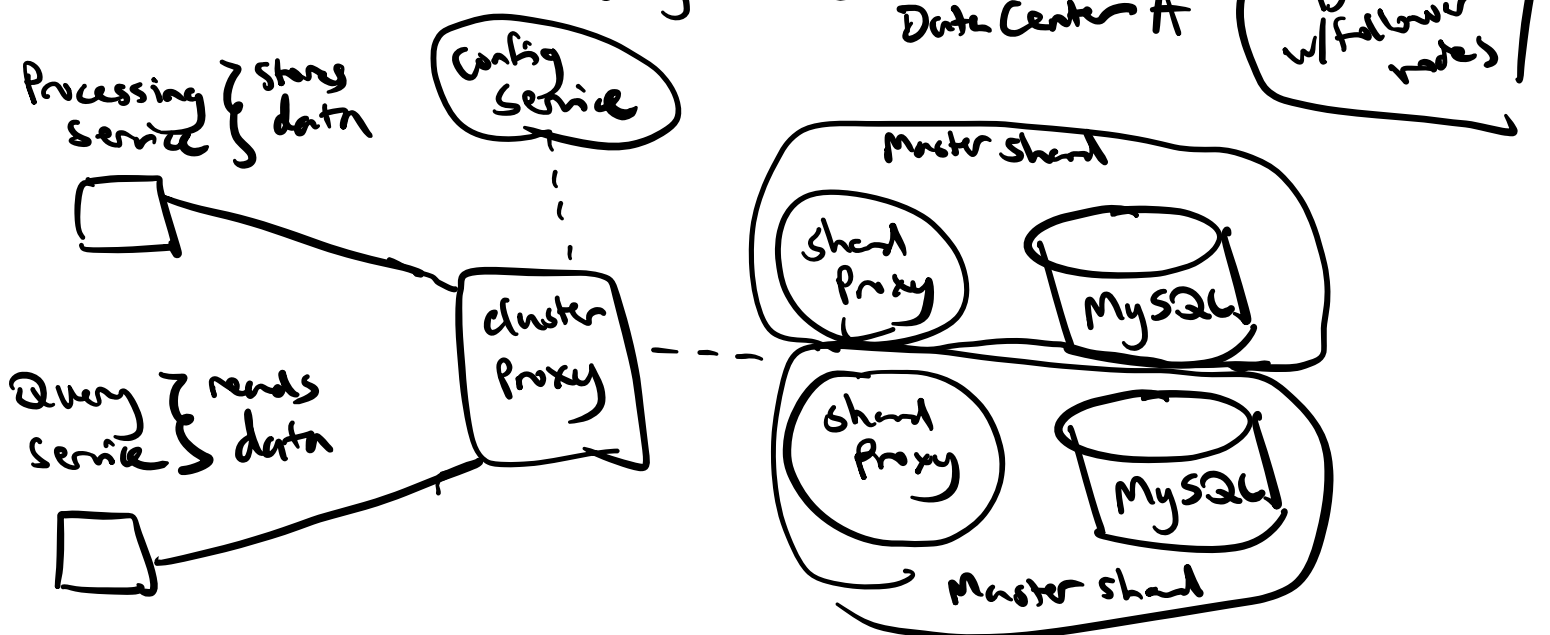
# System Design Interview Guide (Part II)

- Let's think about where we store data.

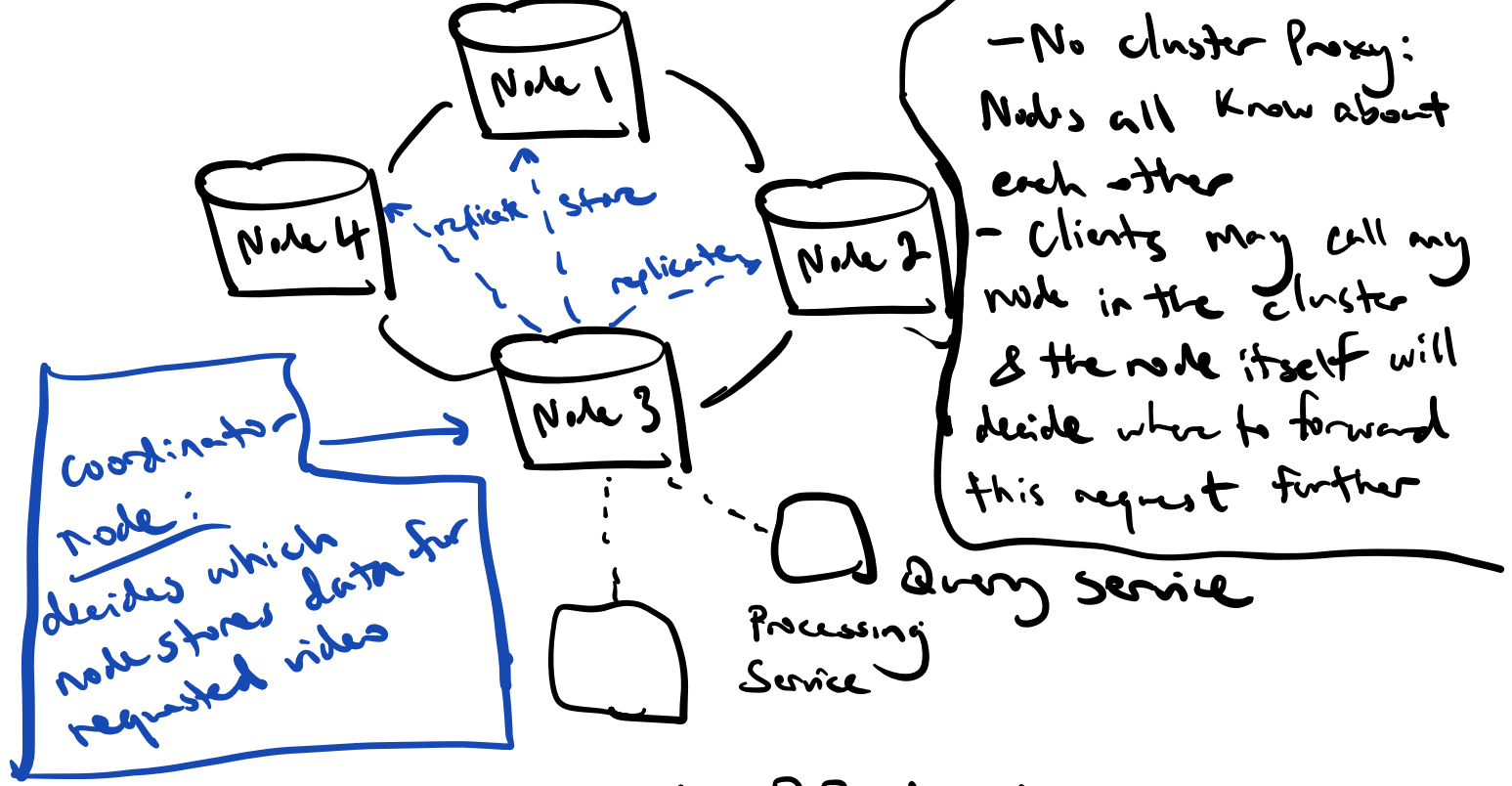
We're using the same example as Part I: "Count YT vid views"  
Evaluate both SQL & NoSQL databases, since both can scale & perform well.

- ↳ How to scale writes?
- ↳ How to scale reads?
- ↳ How to make both reads (r) & writes (w) fast?
- ↳ How to not lose data in case of hardware faults & network partitions?
- ↳ How to achieve strong consistency? What are the tradeoffs?
- ↳ How to recover data in case of an outage?
- ↳ How to ensure data security?
- ↳ How to make it extensible for data model changes in the future?

## SQL Database (MySQL)



## NoSQL Database (Cassandra)



- we can use a simple RR algo to choose the initial node, or we may be smarter & choose a "closest to client" node.
  - Node 3 writes to Node 1. Can also copy to other nodes simultaneously.
    - ↳ waiting for responses may be slow, so use a quorum: coordinating node returns successful if a certain number of nodes return successfully
  - Similar concept for reads: Some nodes may have stale data, a read quorum defines a minimum # of nodes that have to agree upon the response
  - Cassandra uses version num to determine staleness of data
- use consistent hashing to decide which node handles the operation.**

Similar to SQL DBs, we copy data to other data centers for high availability. So this setup could be one of many.

- In Part I, we chose **availability over consistency**:  
State data better than no data.
  - If we replicate data **synchronously** it will be slow but consistent. We usually replicate it **asynchronously**.

### Example: Leader-Follower Replication

↳ some read replicas may be behind their master

↳ some users will see different view counts

- eventually the nodes they're reading from will have correct info: **eventual consistency**

### • How do we store Data?

- Define Needs

↳ Build a report that shows: info about the video, # of total views per hour for last several hours & info about the channel this video belongs to

↳ can store all of the data in either relational or non relational DBs

- While we don't need to know the details of how every DB works, it's important to know the tradeoffs

↳ example: Cassandra is a wide-column DB that supports asynchronous masterless replication.

↳ example: MongoDB is a document-oriented DB & uses leader-based replication

↳ example: HBase is column based like Cassandra but supports leader-based replication

### • Processing Service

↳ How to scale?

↳ How to achieve high throughput?

↳ How to not lose data when processing node crashes?

↳ What to do when DB is unavailable or slow?

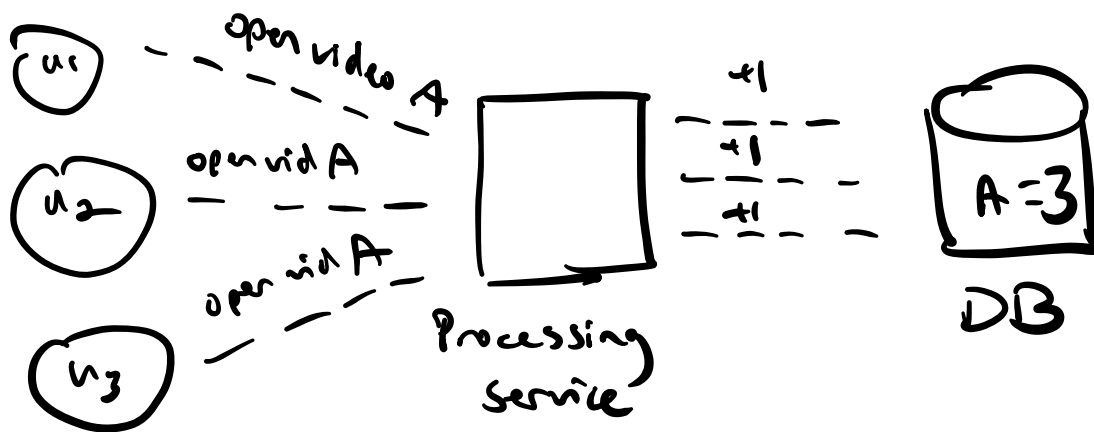
How to make data processing scalable, reliable, & fast?

- ↳ scalable = Partitioning
- ↳ Reliable = Replication & checkpointing
- ↳ Fast = In-Memory

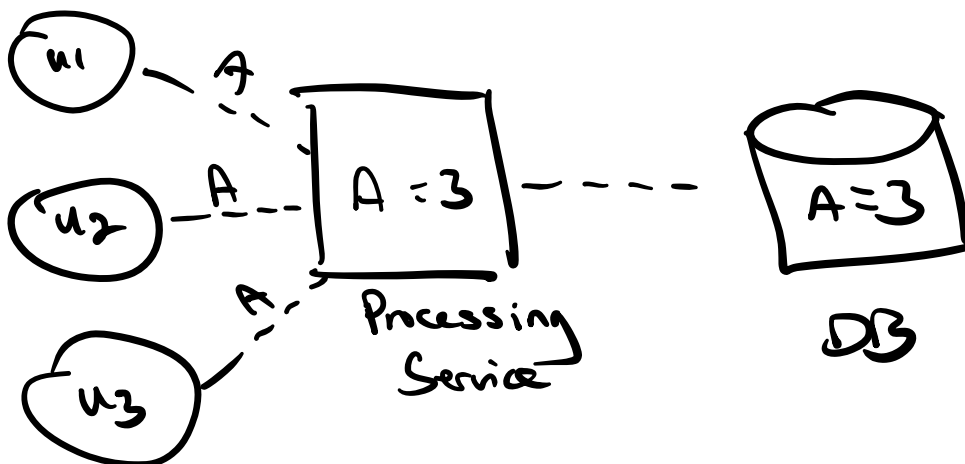
## . Data Aggregation Basics

- Should we pre-aggregate data in the processing service?

What does that look like: (Push events synchronously to processing service)

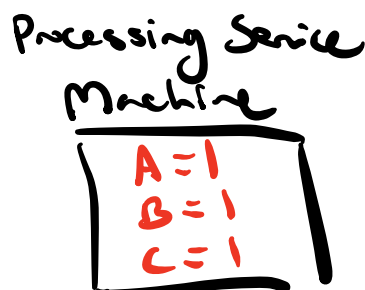
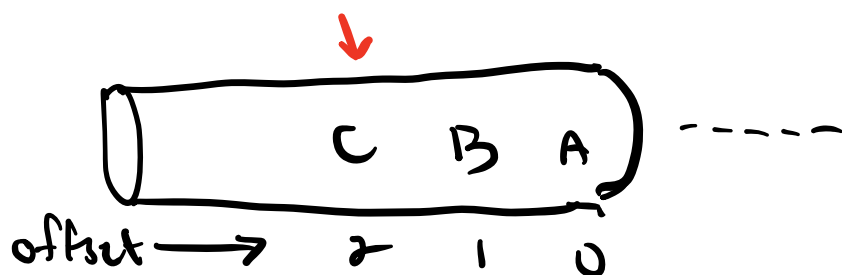
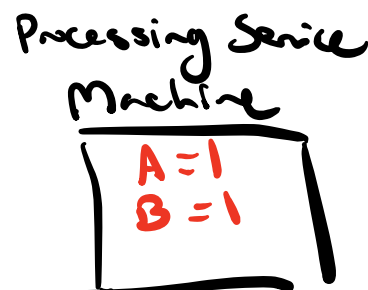
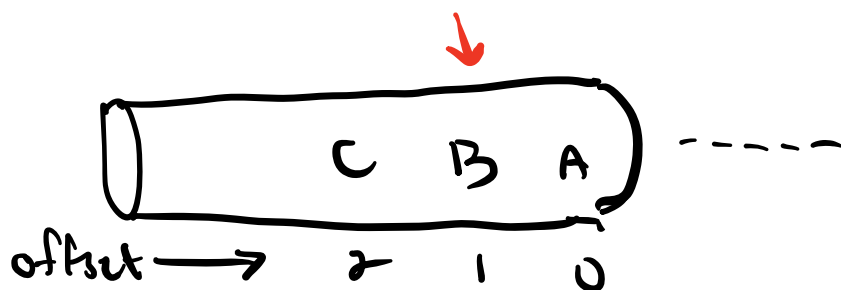
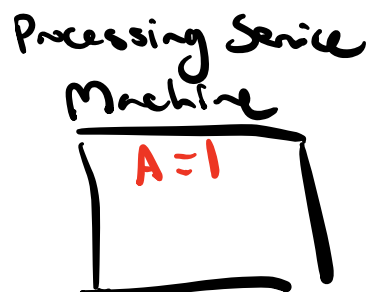
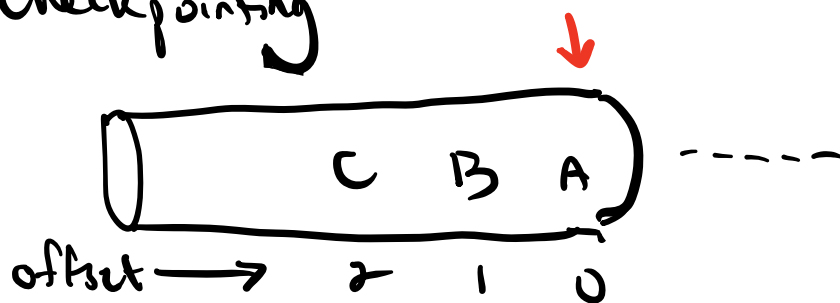


or better (for large systems): (Pull from temp storage)



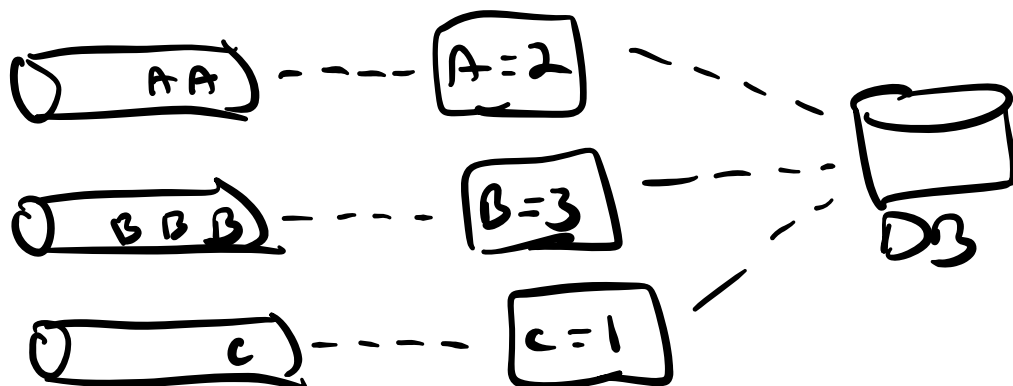
Both options are valid, but Pull has more fault tolerance because data is pulled from temporary storage by the processing service. If the processing service fails, the temporary storage will persist the data.

## • Checkpointing



- Every time an event is read from storage, current offset moves forward.
- After several events are processed & stored in the DB, we write the checkpoint to some persistent storage
- If processing machine for service fails, it will be replaced; resumes processing where prev left off

## • Partitioning



- Compute Hash & based on this hash we pick a queue
- As we can see, partitioning helps us parallelize events

processing.

- More events, more partitions are made

• Processing Service Deep Dive

Purpose of service:

↳ reads events from partition one by one

↳ counts events in memory

↳ flushes counted values to DB periodically