# [544] Hadoop Ecosystem

Tyler Caraza-Harter



# Learning Objectives

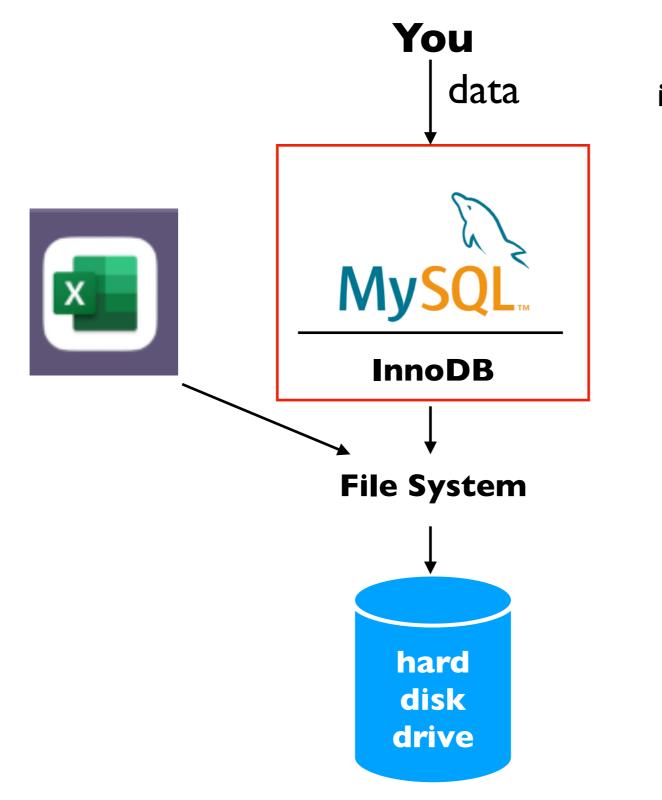
- describe the purpose of GFS, MapReduce, and BigTable (at a high level), and similar Hadoop systems (HDFS, Spark, and Cassandra)
- describe partitioning and replication and the motivation for each technique
- identify the role that clients, NameNodes, and DataNodes play for HDFS reads and writes

# **Outline: Hadoop Ecosystem**

Motivation, Hadoop Ecosystem

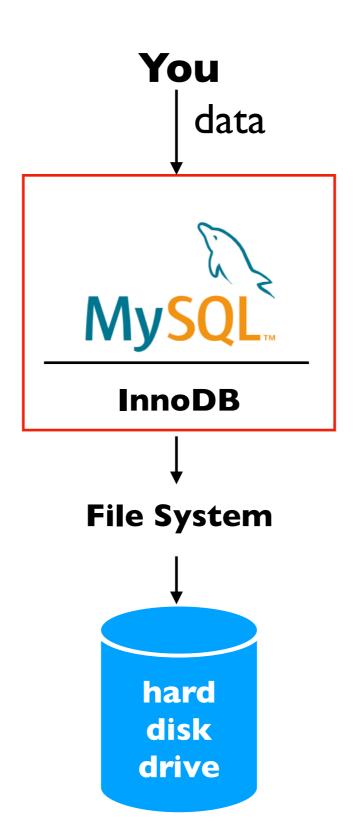
Hadoop File System (HDFS)

# Design: storage systems are generally built as a composition of layered subsystems

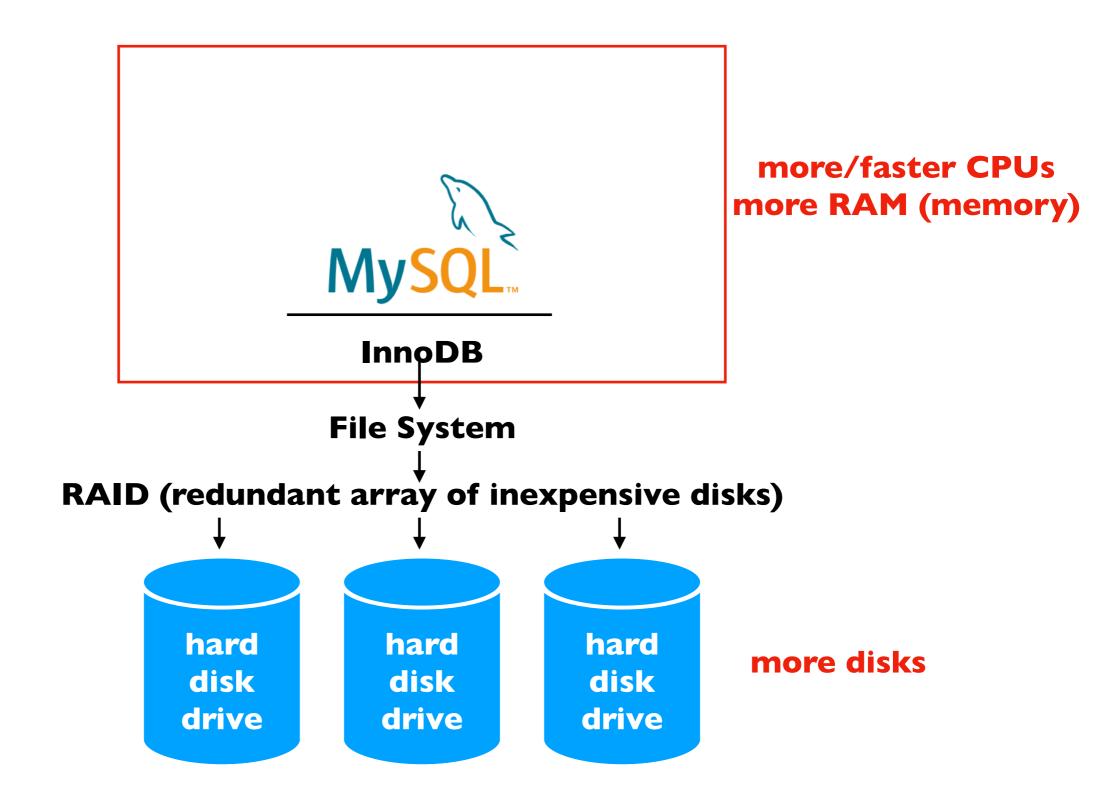


**Today: 3 layered systems** in the Hadoop Ecosystem

#### What if your data is too big for your server?

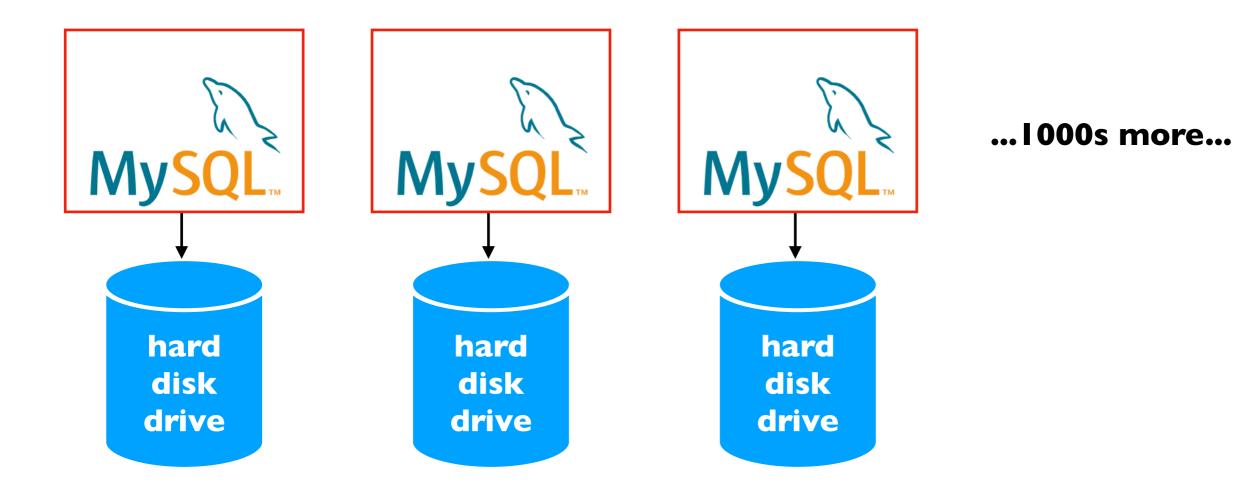


What if your data is too big for your server? Option I: scale up (buy better hardware)

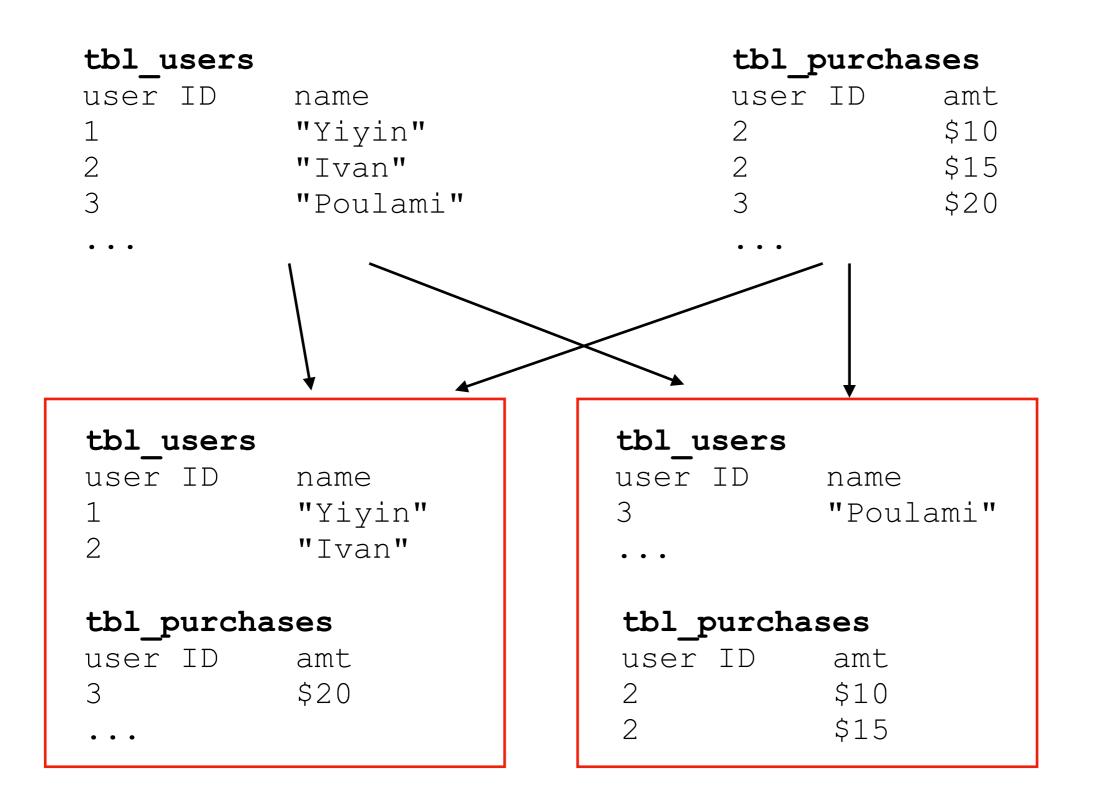


What if your data is too big for your server? Option 2: scale out (more machines)

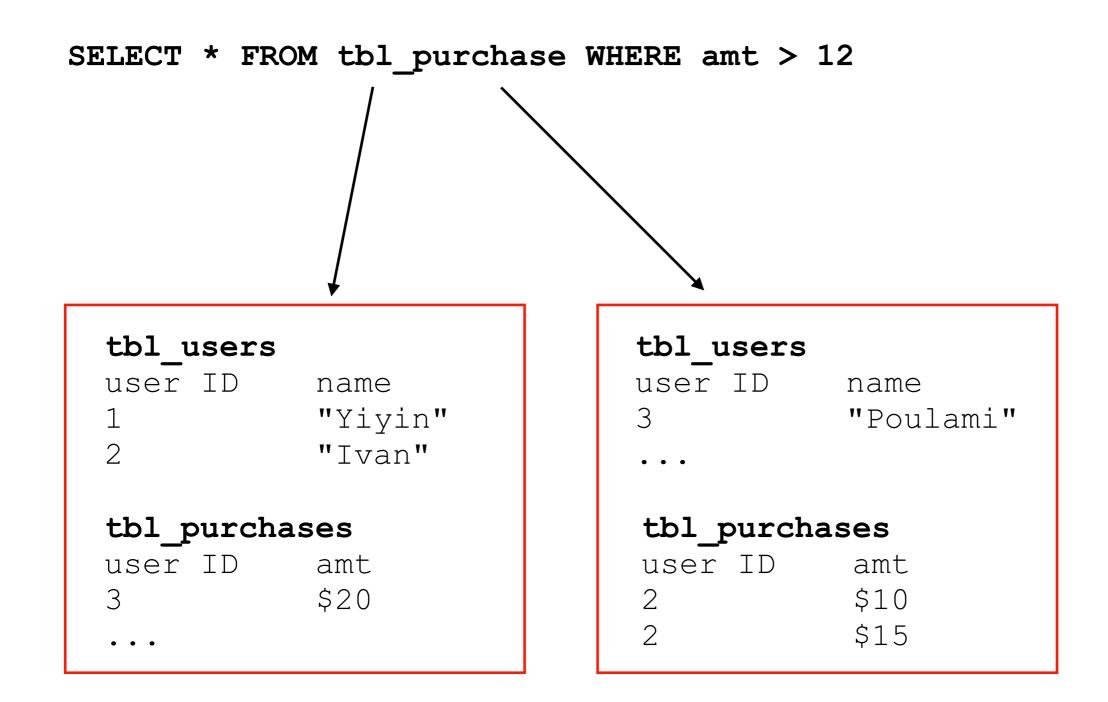
#### where does the data actually go?



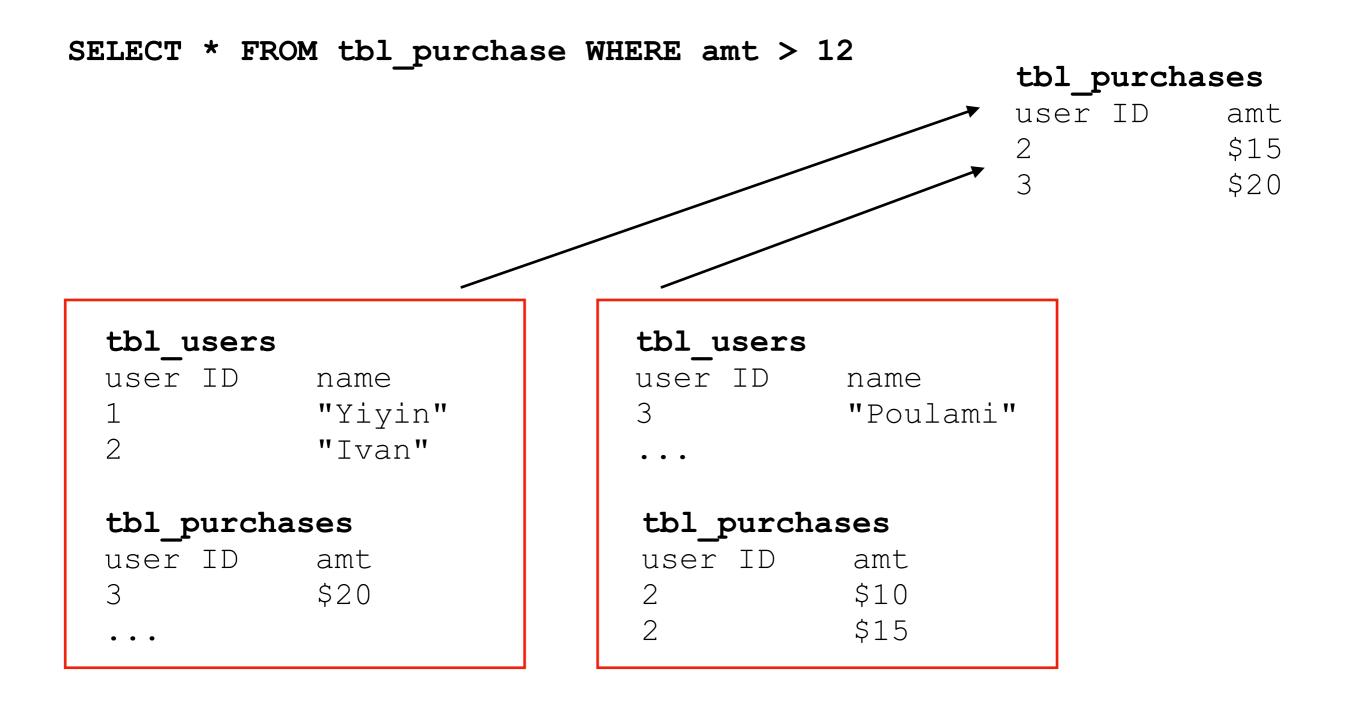
#### Approach: partition the tables



#### Approach: send queries to multiple DBs...



#### ...combine results



#### What is a query that would break things?

SELECT ...

tbl_users	
user ID	name
1	"Yiyin"
2	"Ivan"
tbl purcha	ases
<b>tbl_purch</b> a user ID	ases amt
user ID	amt

<b>tbl_users</b> user ID 3	name "Poulami"			
<b>tbl_purchases</b> user ID amt				
2 2	\$10 \$15			

#### What is a query that would break things?

SELECT \* FROM tbl\_users
INNER JOIN tbl\_purchases
ON tbl\_users.user\_id = tbl\_purchases.user\_id

<b>tbl_users</b> user ID 1 2	name "Yiyin" "Ivan"	<b>tbl_users</b> user ID 3	name "Poulami"
<b>tbl_purch</b> auser ID 3		<b>tbl_purch</b> user ID 2 2	

# Why use a traditional/relational DB if basic things like JOIN don't work right at scale?

example: Cassandra documentation

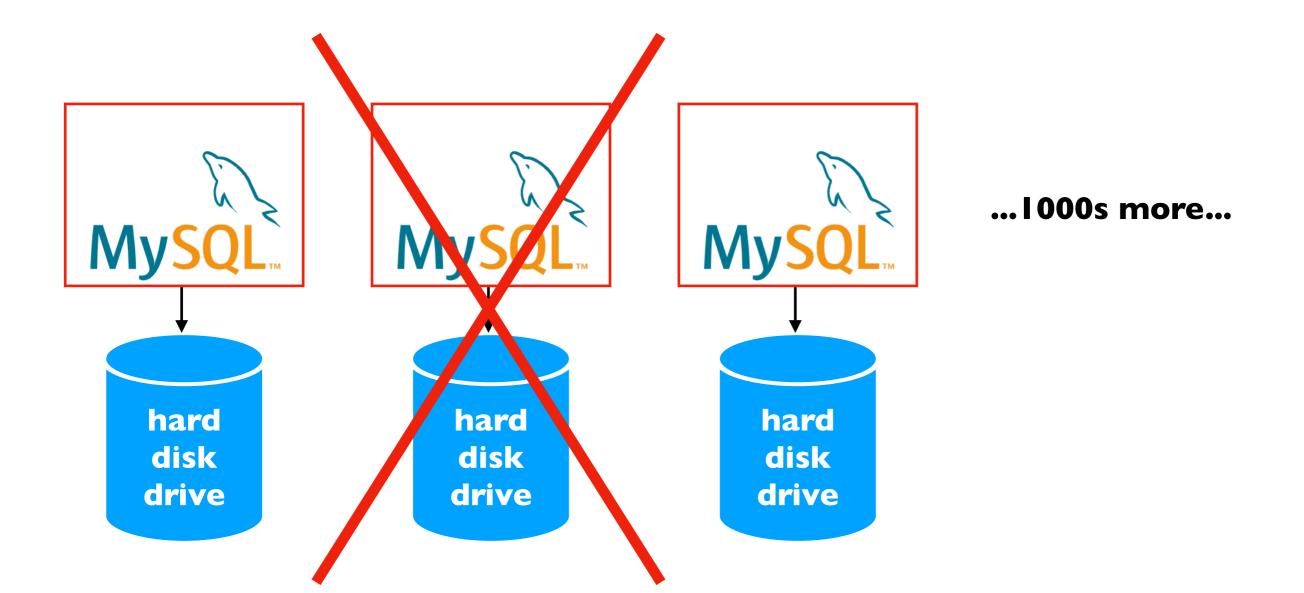
#### **STEP 3: CREATE FILES**

The Cassandra Query Language (CQL) is very similar to SQL but suited for the JOINless structure of Cassandra.

https://cassandra.apache.org/\_/quickstart.html

What if a server dies?

happens all the time when you have 1000s of machines

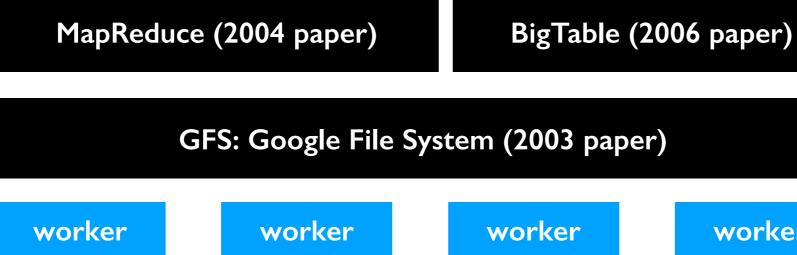


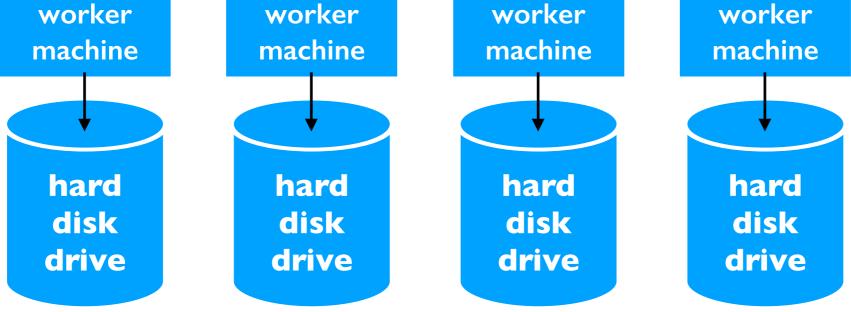
#### Motivation for System Redesign

#### Features

- some classic features (like JOINS and transactions) may not be essential
- scaling to many machines is essential
- fault tolerance is essential

#### **Google Architecture**





radical idea: base everything on lots of cheap, commodity hardware

#### Hadoop Ecosystem

Yahoo, Facebook, Cloudera, and others developed opensource Hadoop ecosystem, mirroring Google's systems

	Google (paper only)	Hadoop, 1st gen (open source)	Modern Hadoop
<b>Distributed File System</b>	GFS	HDFS	
<b>Distributed Analytics</b>	MapReduce	Hadoop MapReduce	Spark
Distributed Database	BigTable	HBase	Cassandra

Ecosystem: Ambari, Avro, Cassandra, Chukwa, HBase, Hive, Mahout, Ozone, Pig, Spark, Submarine, Tez, ZooKeeper

https://hadoop.apache.org/

# Outline: Hadoop Ecosystem

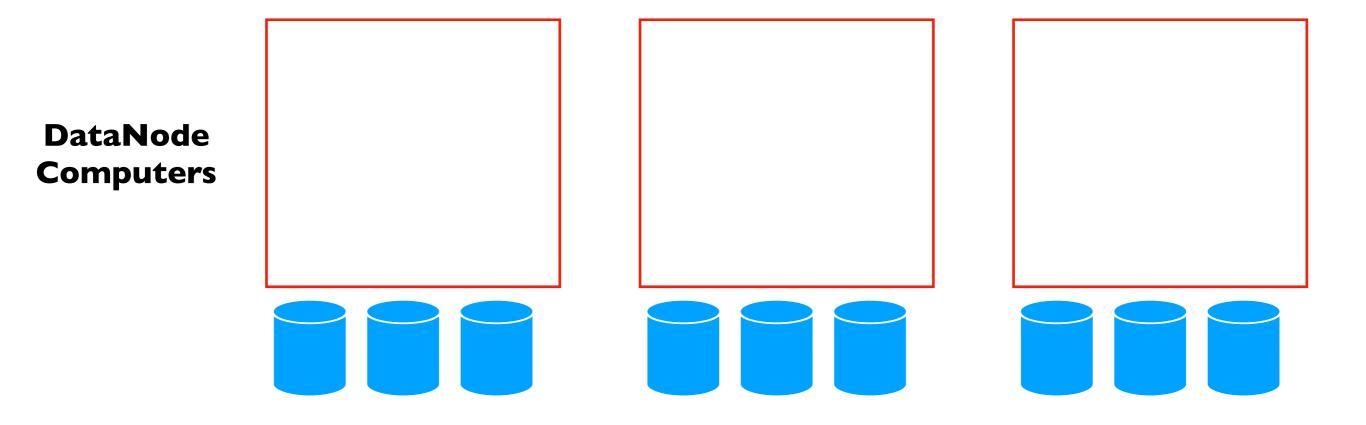
Motivation, Hadoop Ecosystem

Hadoop File System (HDFS)

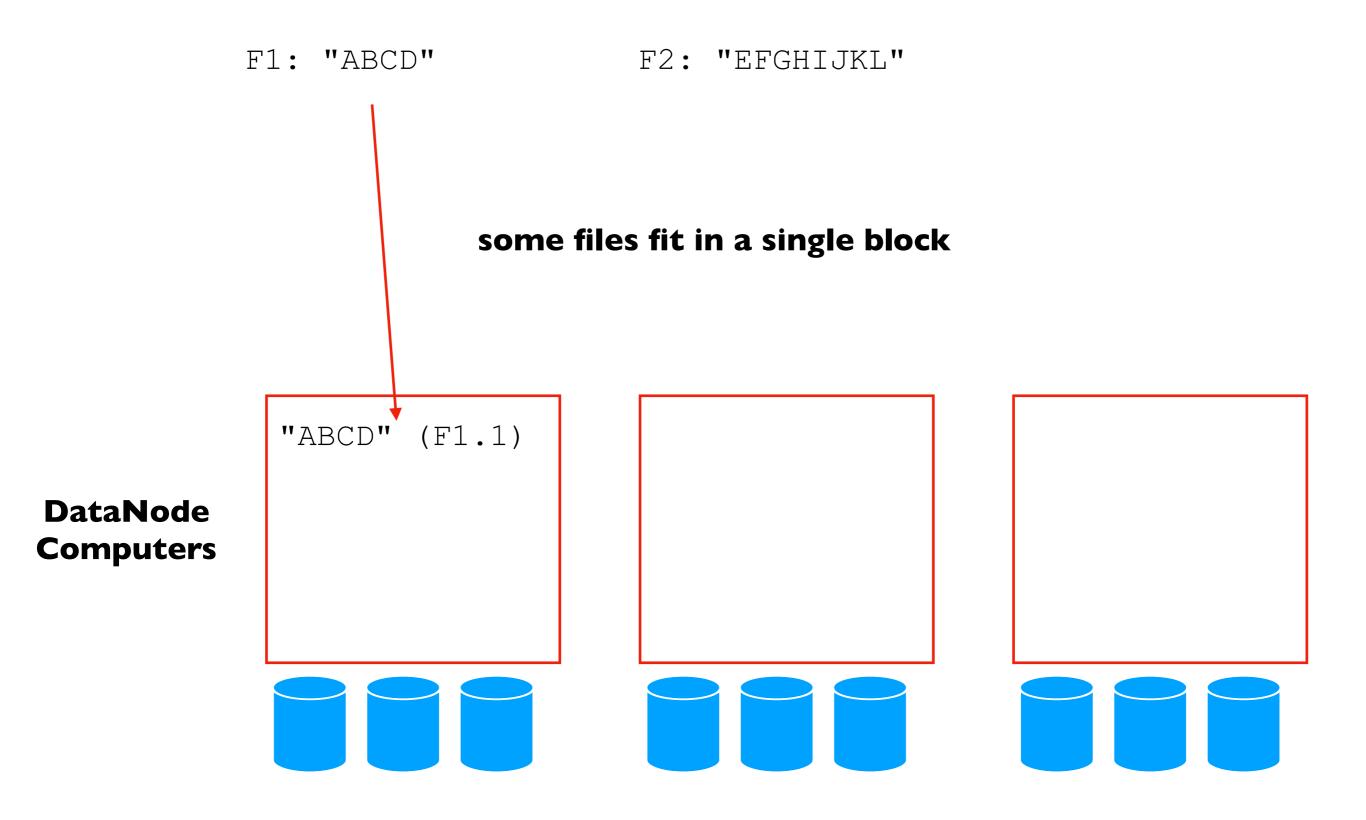
### HDFS: DataNodes store File Blocks

F1: "ABCD" F2: '

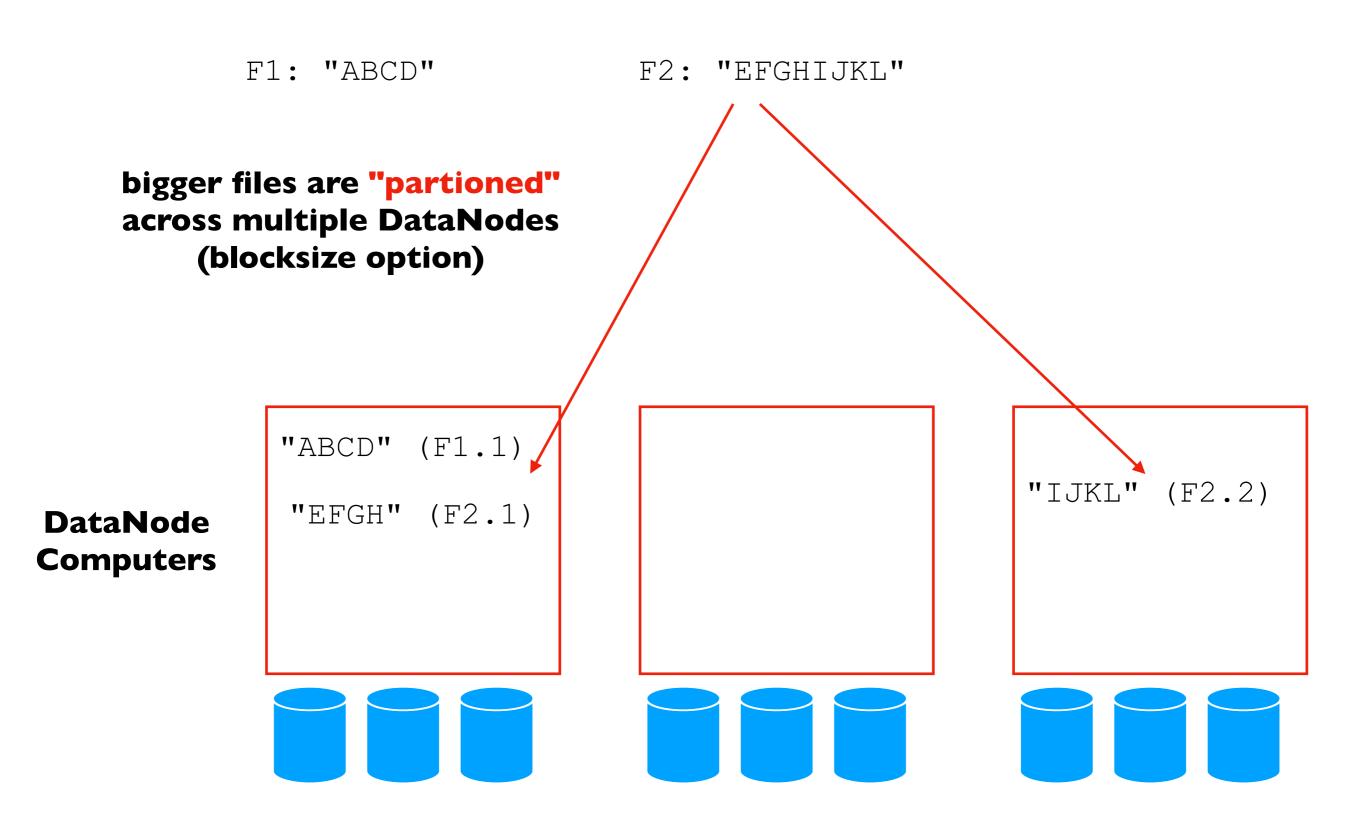
F2: "EFGHIJKL"



### HDFS: DataNodes store File Blocks

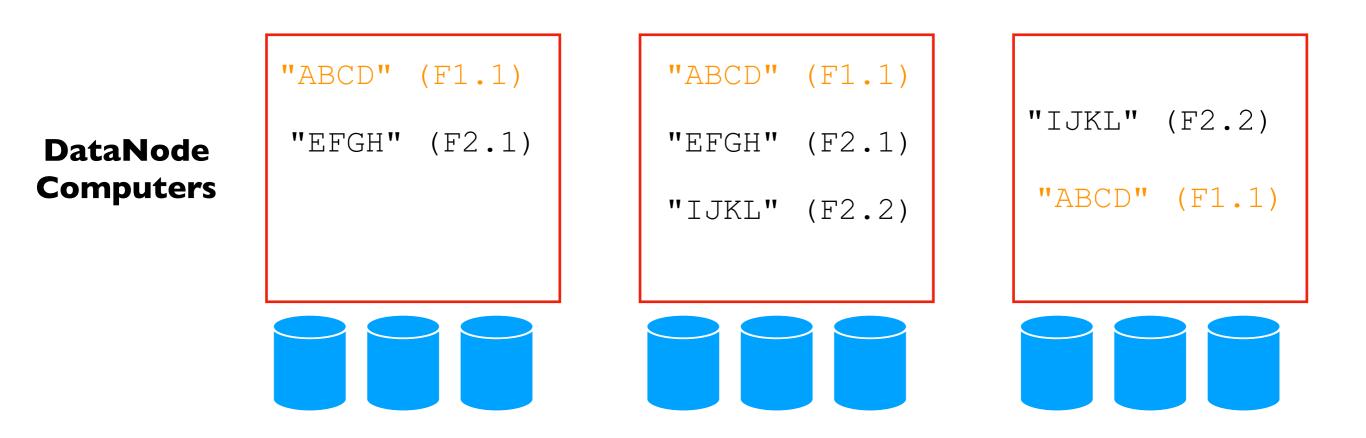


### Partitioning Across DataNodes

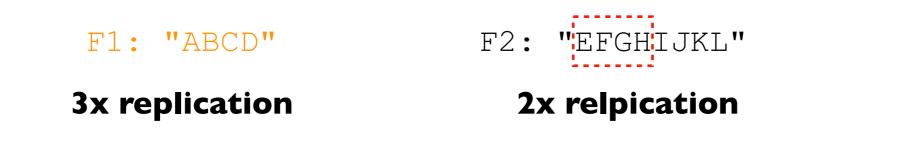


### **Replication Across DataNodes**

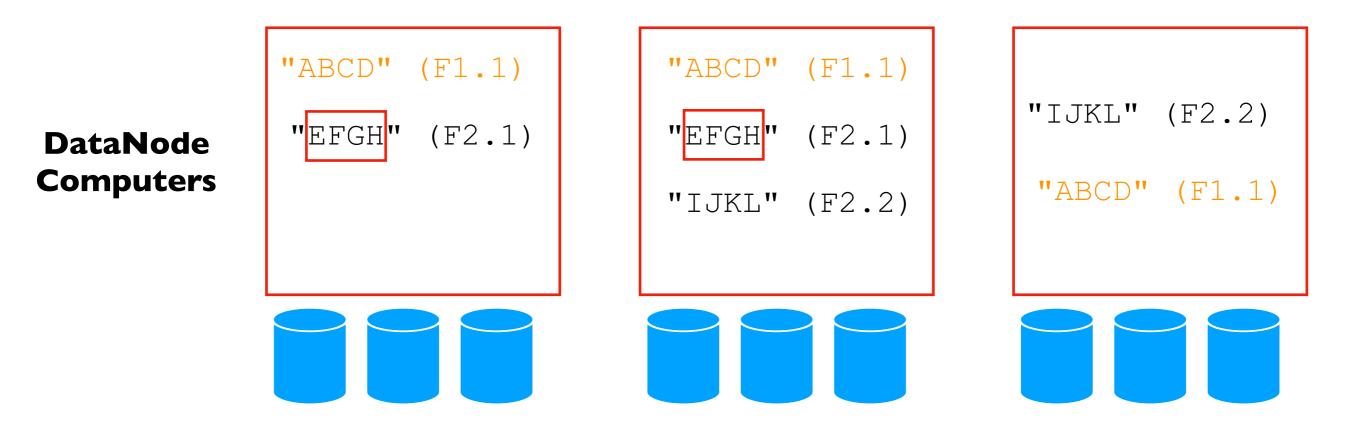




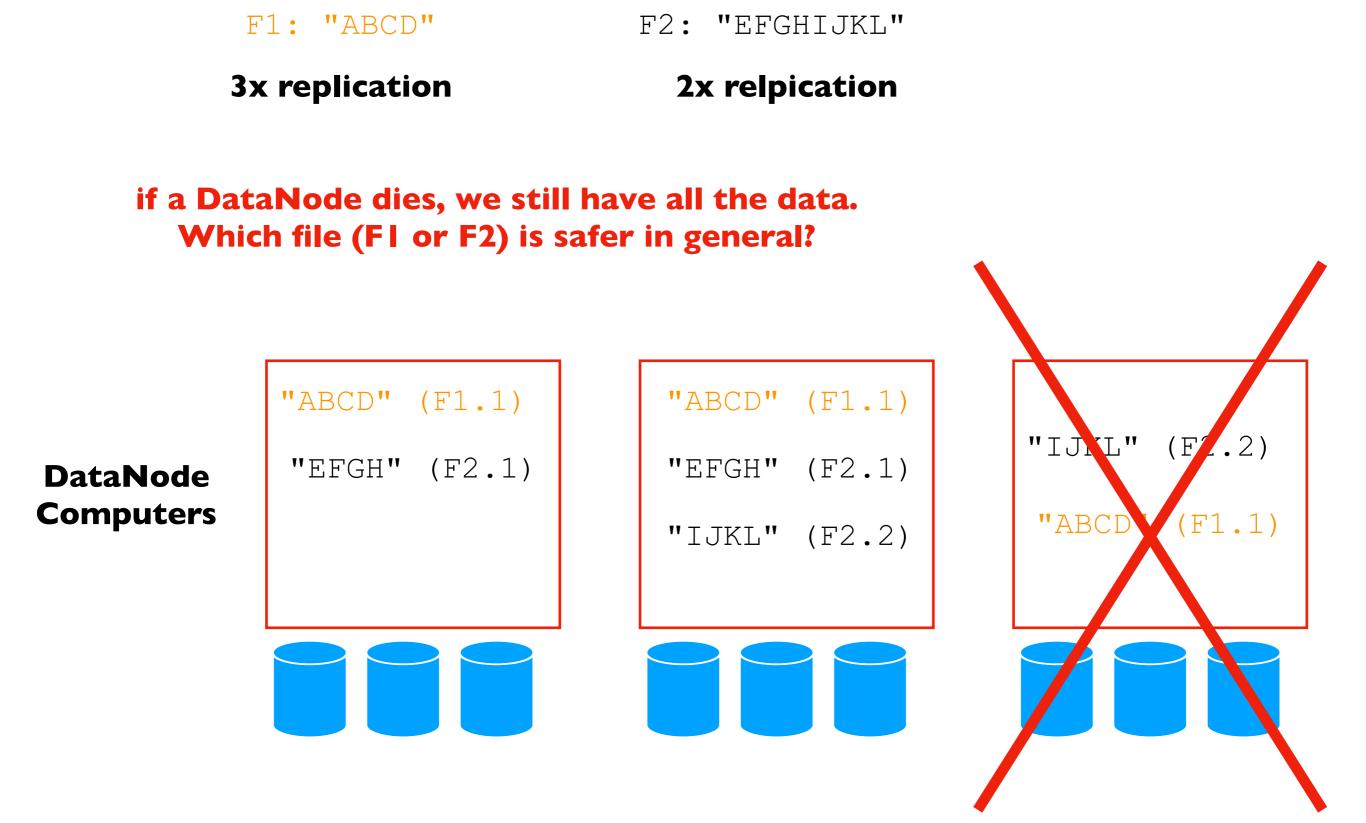
### **Replication Across DataNodes**



#### logical vs. physical blocks



### **Replication Across DataNodes**



# Aside: Replication vs. Erasure Encoding

HDFS Strategies for handling node failure

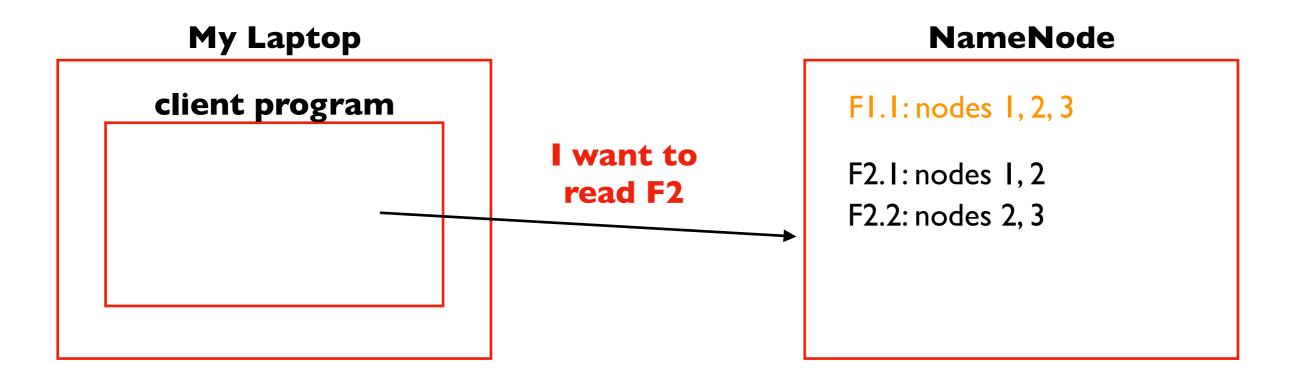
#### Replication

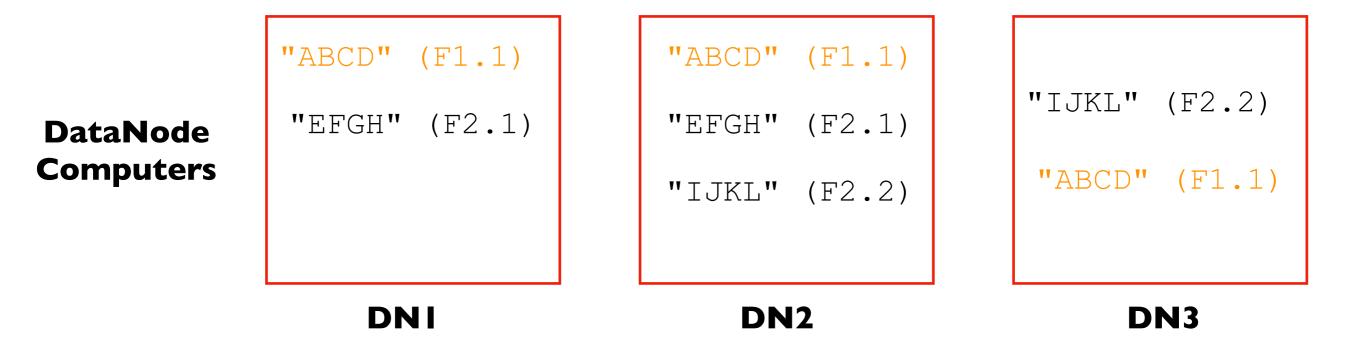
- original strategy, used for new/hot data
- covered in CS 544

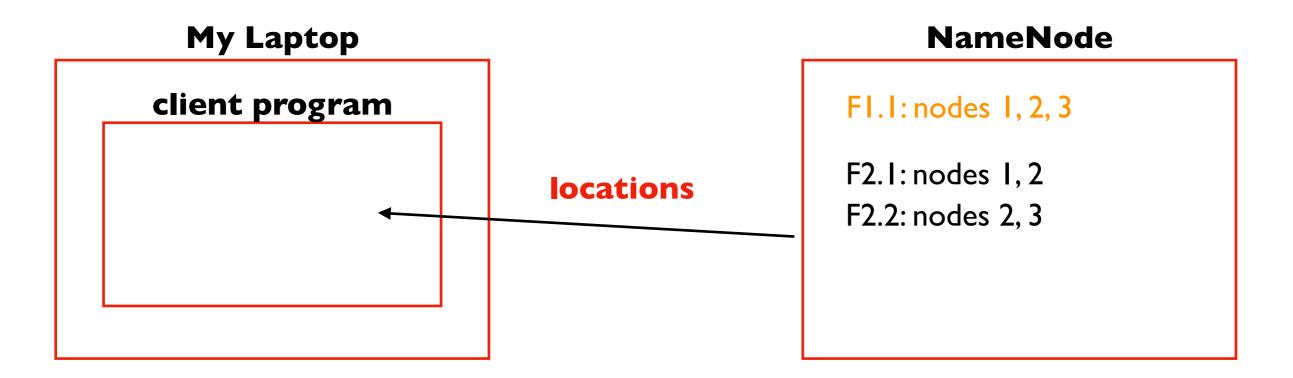
#### **Erasure Encoding**

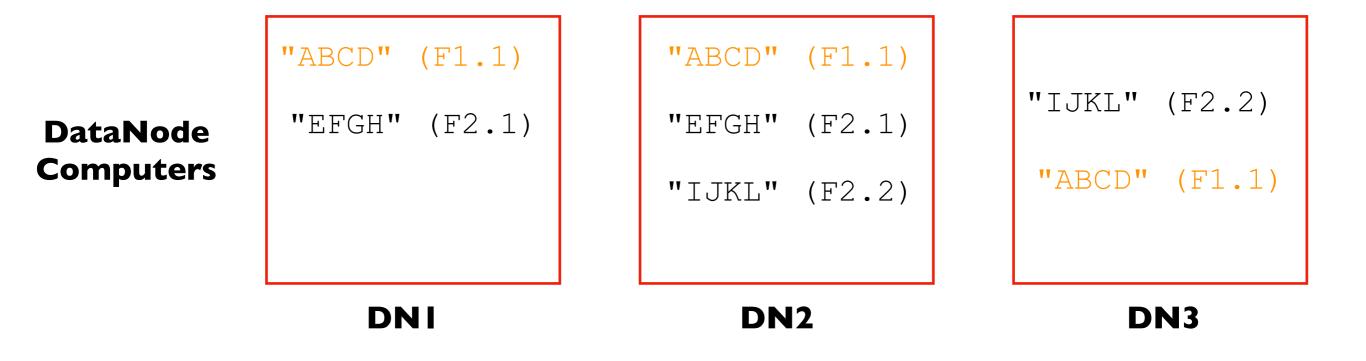
- more space efficient, less I/O efficient
- recent HDFS feature used for cold data (NOT covered in CS 544)

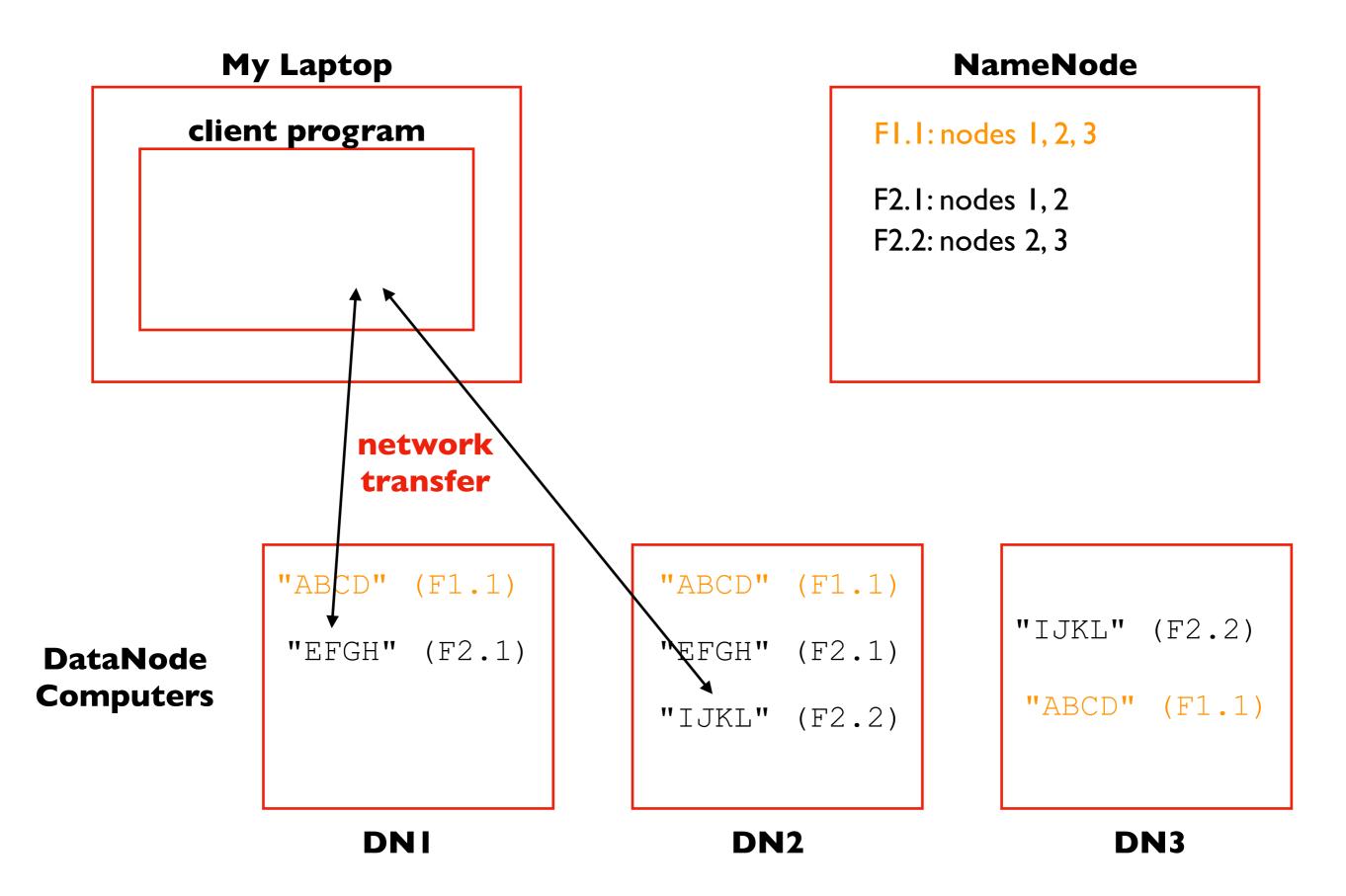
https://hadoop.apache.org/docs/stable/hadoop-project-dist/hadoop-hdfs/HDFSErasureCoding.html

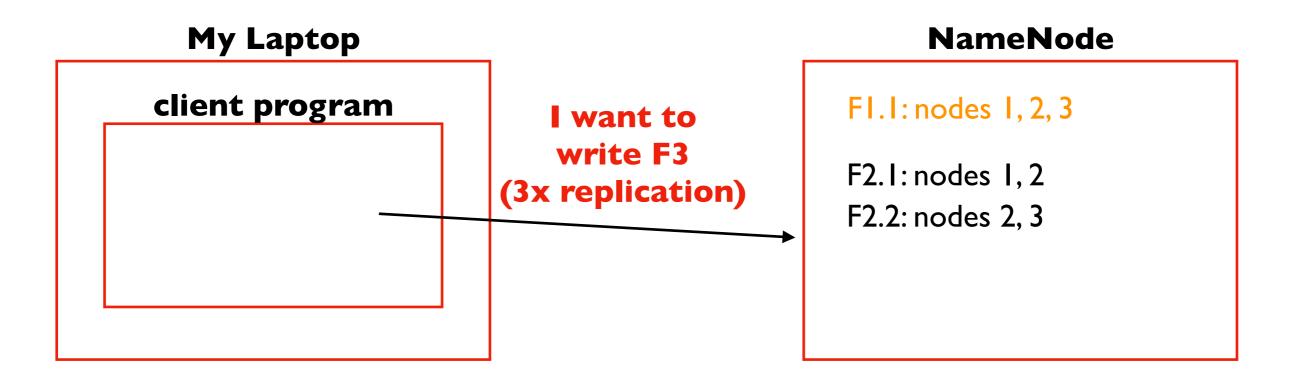


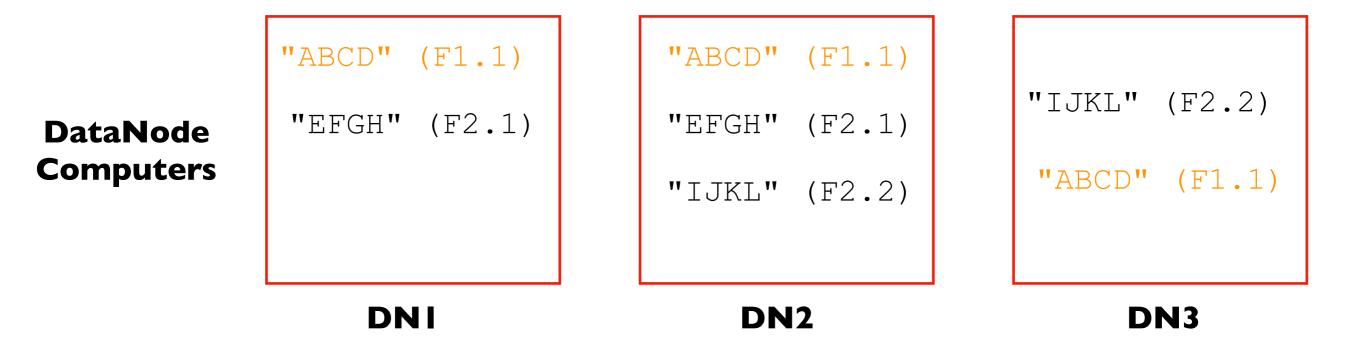


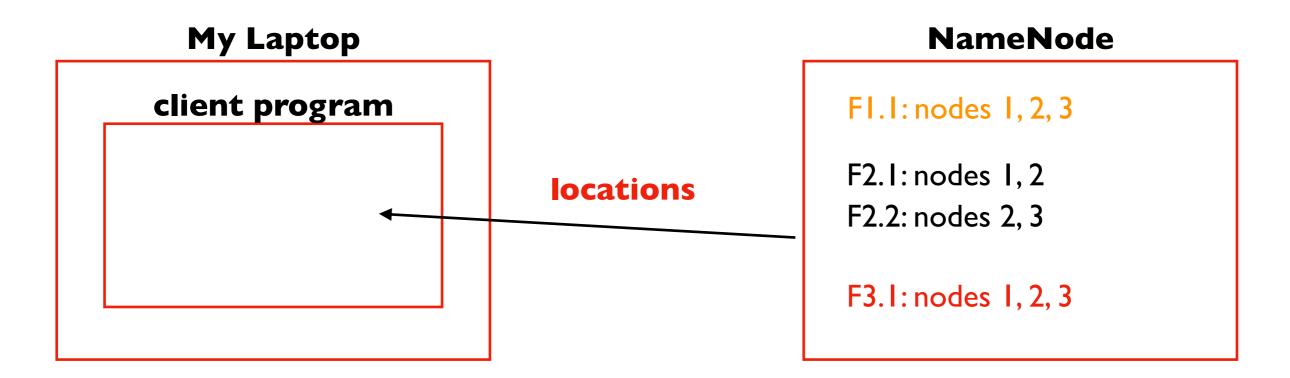


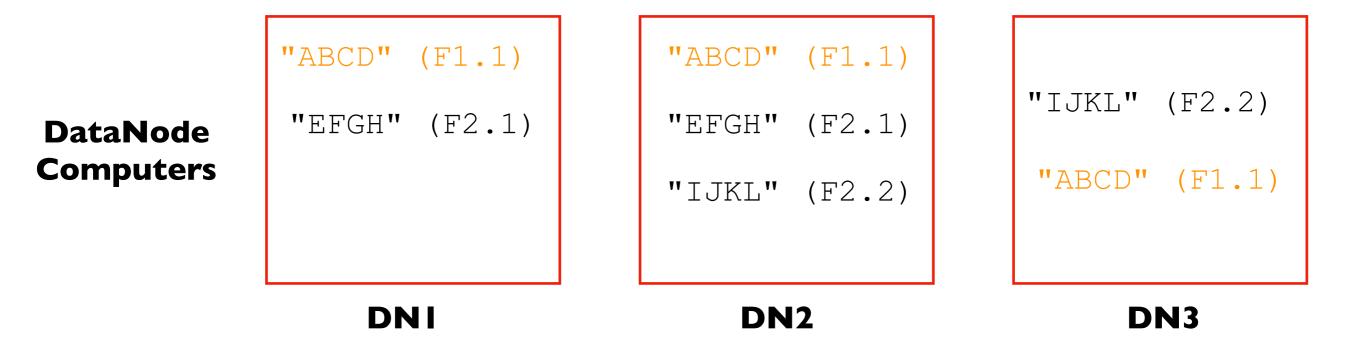


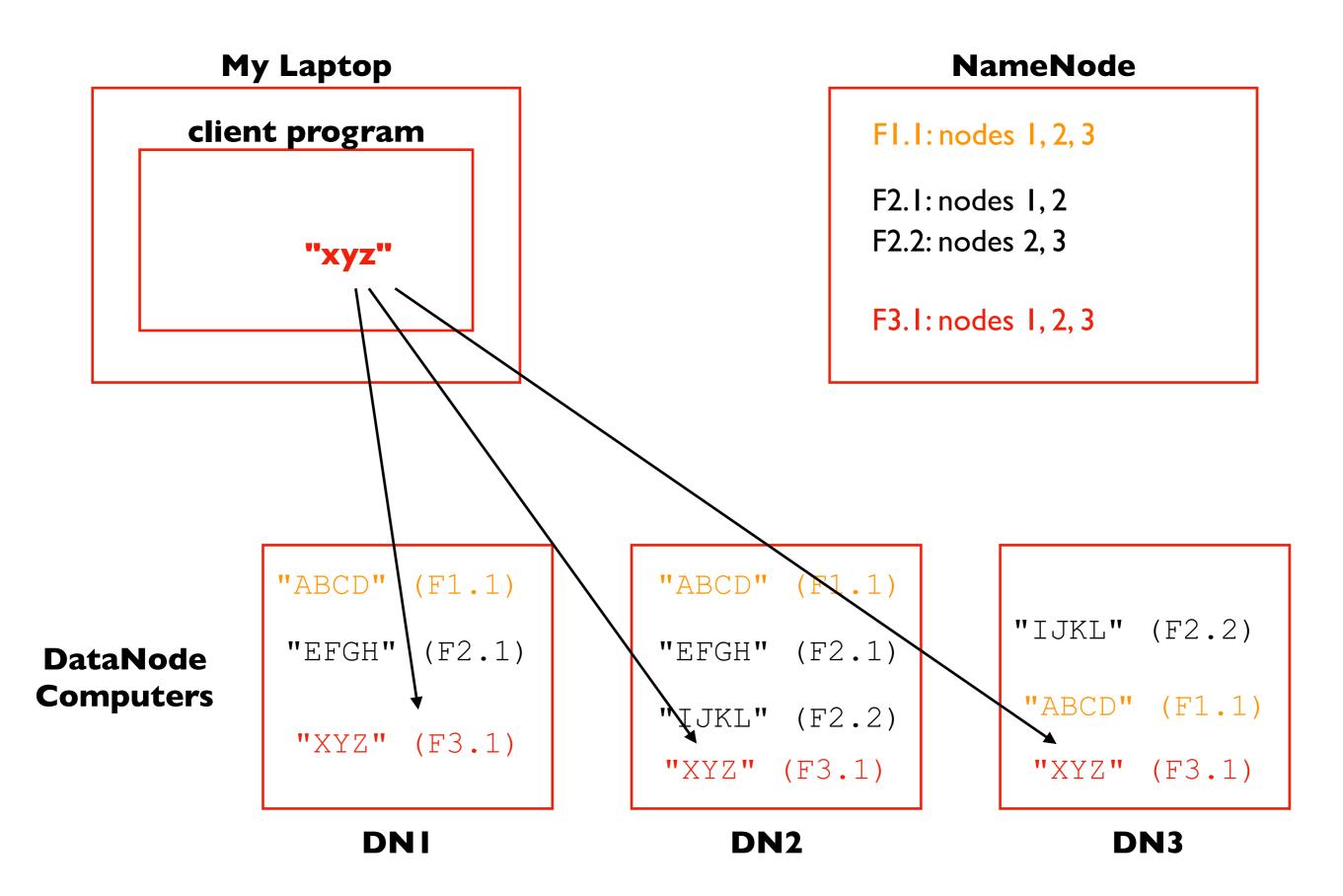


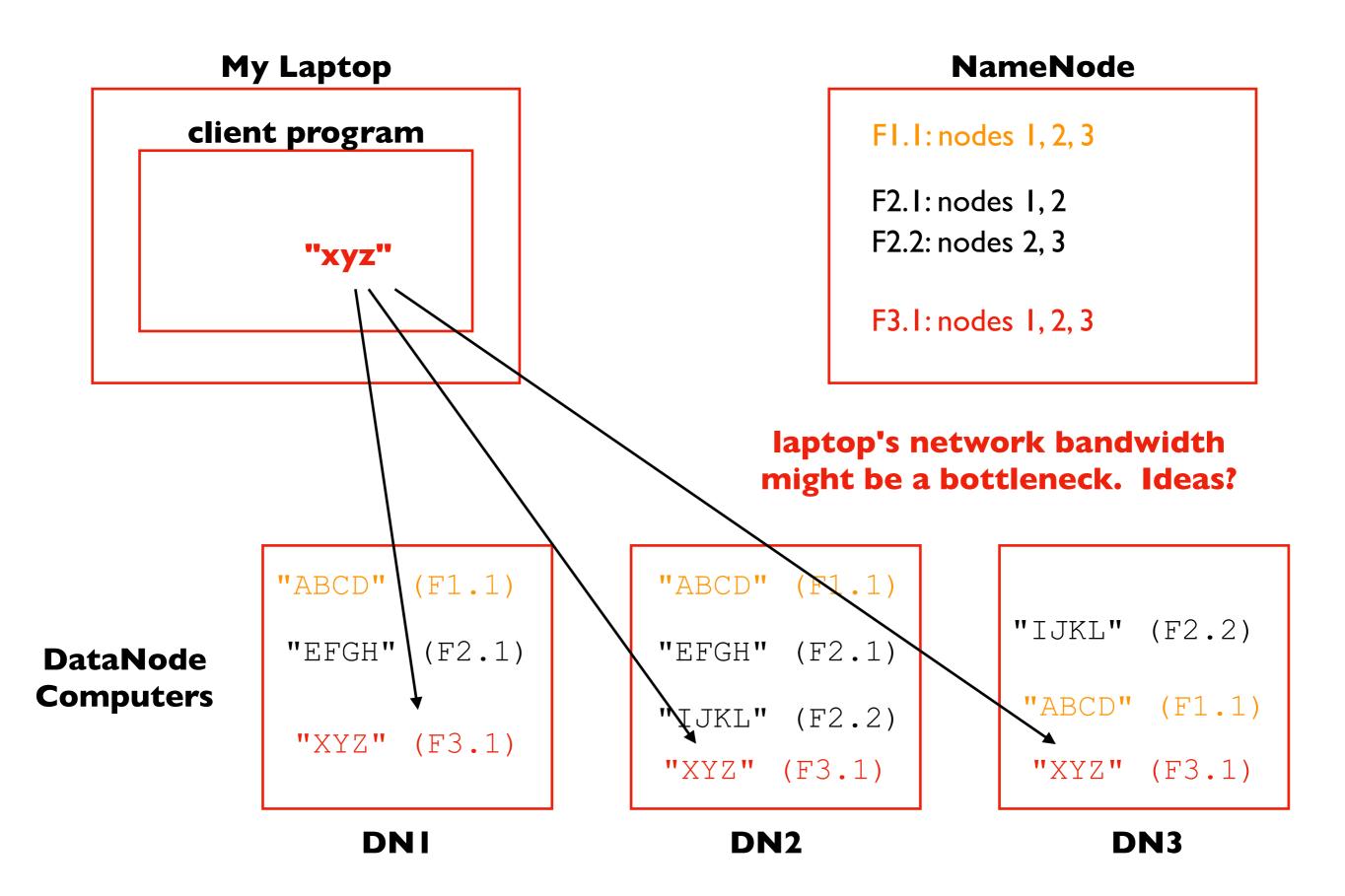




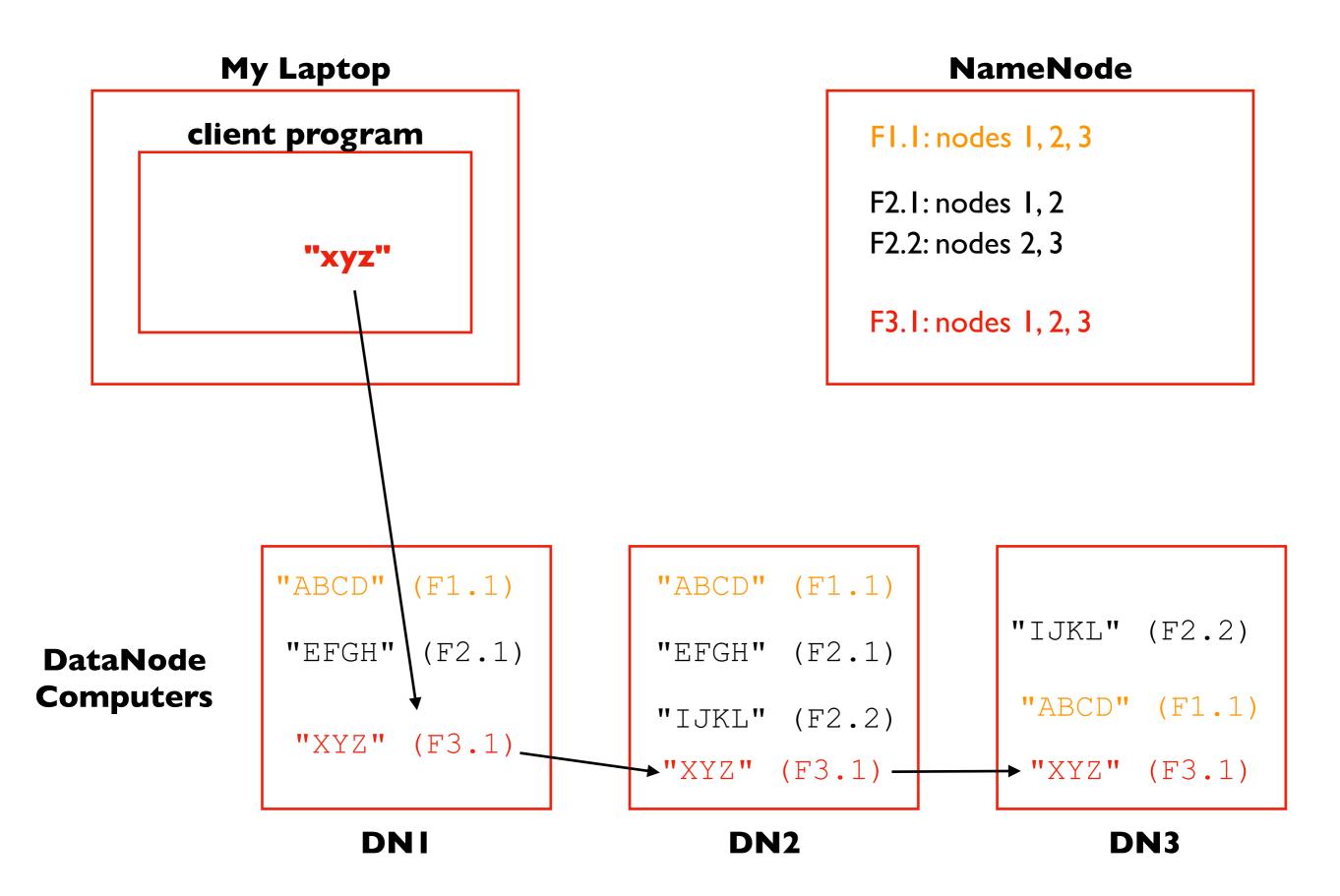








# **Pipelined Writes**

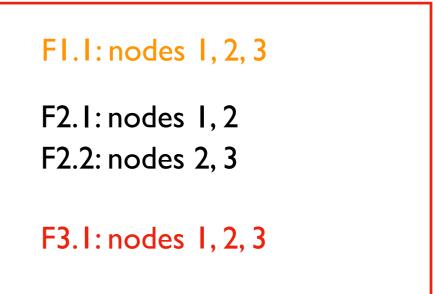


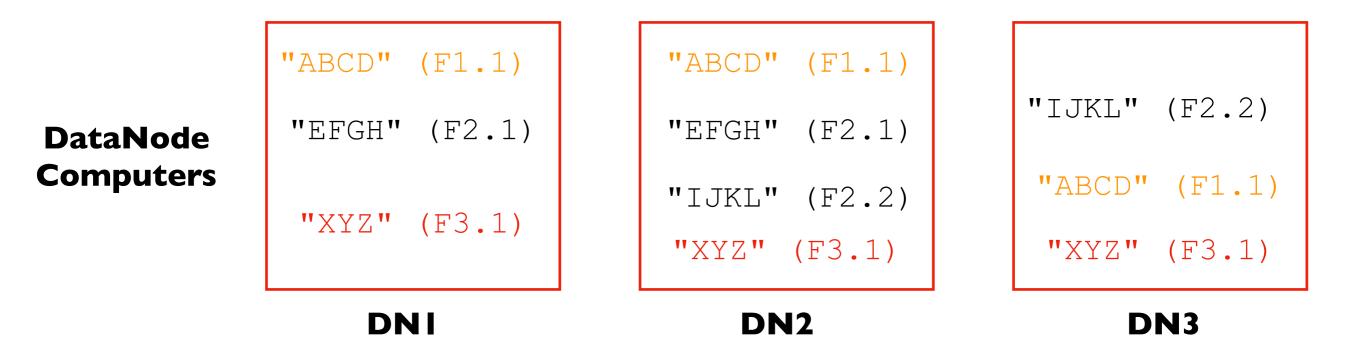
### How are reads/writes amplified at disk level?

if a client **writes** 4 MB to a 2x replicated file, how much data do we **write** to hard drives?

if a client **reads** 2 MB to a 3x replicated file, how much data do we **read** from hard drives?

#### NameNode





#### What are the tradeoffs of replication factor and block size?

benefits of high replication?

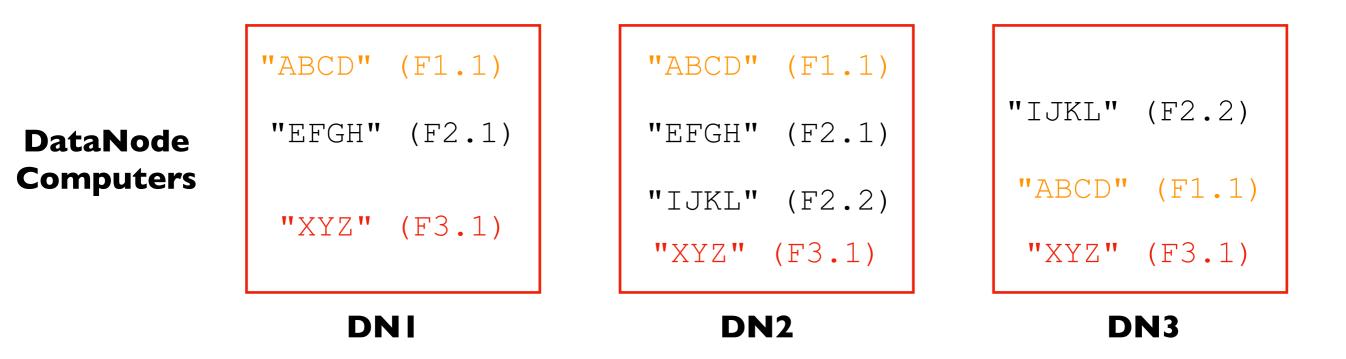
benefits of low replication?

benefits of large block size?

benefits of small block size?

NameNod	e
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F1.1: nodes 1, 2, 3 F2.1: nodes 1, 2 F2.2: nodes 2, 3 F3.1: nodes 1, 2, 3



# How do we know when a DataNode fails?

#### Heartbeat Message **NameNode** DataNode to NameNode FI.1: nodes 1, 2, 3 • Every N seconds (e.g., 3) • Thresholds for no messages F2.1: nodes 1, 2 F2.2: nodes 2, 3 stale (>M seconds) dead (>N seconds) F3.1: nodes 1, 2, 3 • When dead, blocks might be underreplicated and need new replicas live stale (eventually dead) live "ABCD" (F1.1) "ABCD" (F1.1) "IJKL" (F2.2) "EFGH" (F2.1) "EFGH" (F2.1) DataNode Computers "ABCD" (F1.1) "IJKL" (F2.2) "XYZ" (F3.1) "XYZ" (F3.1) "XYZ" (F3.1) DN DN2 DN3

### Summary: Some Key Ideas Applied to HDFS

#### To build complex systems...

• compose layers of subsystems

#### To scale out...

• partition your data

#### To handle faults...

• replicate your data

#### To detect faults...

• send heartbeats

#### To optimize I/O...

• pipeline writes