Introduction

DSCI 551 Wensheng Wu

Logistics

- Instructor email: wenshenw@usc.edu
 - please indicate section: 551 MW or 551 Tue in the subject line.
- Class meeting times (3 sections):
 - MW & DEN 4-5:50pm
 - Tuesday 3:30-6:50pm
- Office hours:
 - MW 3-3:45pm (Zoom, link to be posted)
 - After class
 - Please email for appointment

Logistics

- TAs & office hours
 - Watch the announcements

- Class materials
 - Posted on Blackboard

Piazza

- Discussion forums
 - You may post general and homework questions
 - Do not post solutions
 - Please actively participate in helping others!
 - Do not abuse forum (an academic misconduct!)
- Check frequently for updates
- Follow instruction on Blackboard to sign up

Prerequisites

- Programming skills:
 - Python (homework, Spark), Java (e.g., for Hadoop only)
- Unix-like environment & shell commands (ls?)
 - E.g., Amazon EC2
- Basic knowledge of algorithms and data structures
 - Sorting, hashing, etc. (CS 570)
- Basic probability and statistics

Textbooks

- Remzi H. Arpaci-Dusseau and Andrea C. Arpaci-Dusseau. Operating Systems: Three Easy Pieces, 2015 (selected chapters only). Available free at: http://pages.cs.wisc.edu/~remzi/OSTEP/
- Hector Garcia-Molina, Jeffrey D. Ullman, and Jennifer Widom. Database Systems: The Complete Book (Second Edition), Prentice Hall, 2009. (selected chapters only)
 - http://infolab.stanford.edu/~ullman/dscb.html

Additional readings

- Links can be found in Syllabus
 - Check out the schedule

Grading structure

 Homework 	20%
Midterm 1	15%
• Midterm 2	15%
• Final	25%
 Lab sessions 	5%
 Group project 	20%

Grading scale

- [93, 100] = A
- [90, 93) = A-
- [87, 90) = B+
- [83, 87) = B
- [80, 83) = B-
- [77, 80) = C+
- [73, 77) = C
- ... (see Syllabus for complete breakdown)

Lab sessions

- Flipped:
 - Task and details posted before class
 - Bring questions to class

Typically utilize last 15-30 mins of class

Exams

Closed-notes & book

- Midterms:
 - See syllabus for time

- Final:
 - See syllabus for time

Calculator

Bring one to the tests

• If calculator is needed, we will either announce or state it on the tests

Otherwise, no electronic devices are allowed

Course project

Details coming up

- Done in phases
 - Proposal
 - Midterm report
 - Final report

Late Policy

- Homework/coursework will be submitted online
 - 10% for every 24 hours late
 - No credit after 3 days
- Make up for tests are permitted only when
 - You have an emergency, typically medical
 - Let me know at least two weeks in advance
 - You may be required to contact Campus Support and Intervention office for verification of emergency

Grading Corrections

- All coursework's grades are final one week after grades are posted
- Final exam grades (& all grades) are final after final exam grading review time (to be announced right before/after final)
- Please submit reasonable regrading requests
 - Irrational requests (e.g., simply asking for more points or special treatments) may result in reduction of your grades

Academic Integrity

Cheating will NOT be tolerated

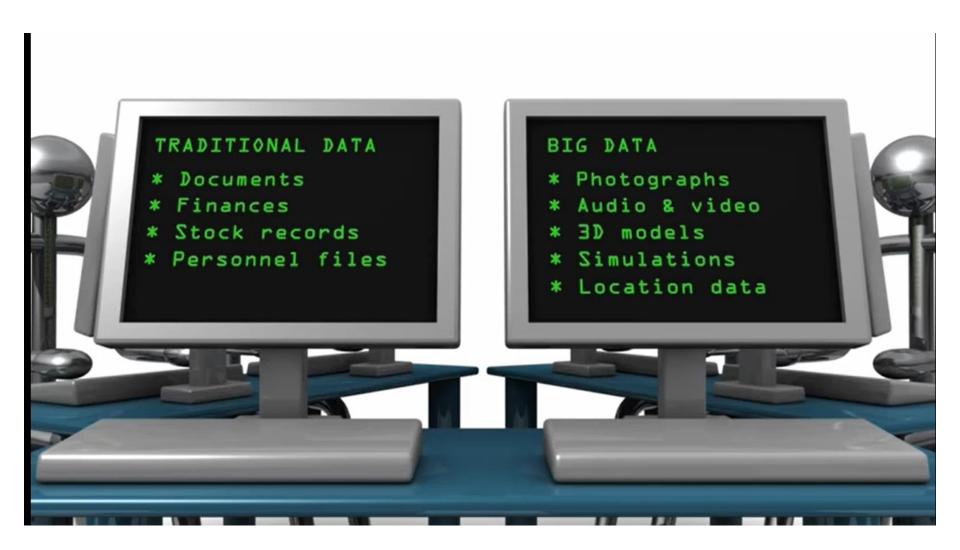
- All parties involved will receive a grade of F for the course and be reported to SJACS WITHOUT EXCEPTION
 - USC Student Judicial Affairs and Community
 Standards

Now, movie time ©

- Explain big data:
 - https://www.youtube.com/watch?v=7D1CQ LOizA

- Questions:
 - Where does big data come from?
 - What characteristics doe it have? 3Vs?
 - What big data technologies were mentioned?

Variety



Internet Traffic in 2012

- 4.8 zettabyte = 4.8 billion terabytes
- Zettabyte (1000 exabytes)
- Exabyte
- Petabyte
- Terabyte (storage)
- Gigabyte = 2^30 (memory)
- Megabyte (128MB, HDFS)
- Kilobyte = 2^10 (4KB)

Major topics

Storage systems



File systems & file formats

- Database management systems (RDBMS)
 - -R = relational

Big data solution stack

Storage Systems

- Hard disk
- SSD (Solid state drive)

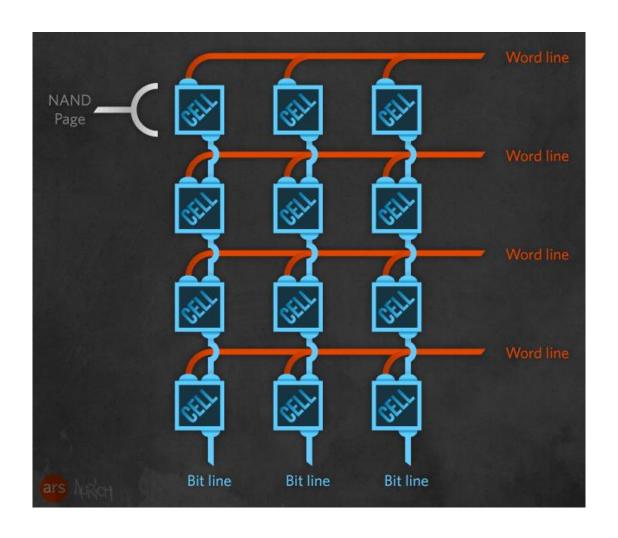




Internal of hard disk



NAND flash



Latencies: read, write, and erase

	SLC	MLC	TLC	HDD	RAM
P/E cycles	100k	10k	5k	*	*
Bits per cell	1	2	3	*	*
Seek latency (µs)	*	*	*	9000	*
Read latency (µs)	25	50	100	2000-7000	0.04-0.1
Write latency (µs)	250	900	1500	2000-7000	0.04-0.1
Erase latency (μs)	1500	3000	5000	*	*
Notes	* metric is not applicable for that type of memory				
Sources	P/E cycles [20] SLC/MLC latencies [1] TLC latencies [23] Hard disk drive latencies [18, 19, 25] RAM latencies [30, 52] L1 and L2 cache latencies [52]				

Major topics

Storage systems

File systems & file formats



Database management systems

Big data solution stack

File Systems

- Standalone
 - Single machine

- Distributed (e.g., Hadoop)
 - A number of data servers

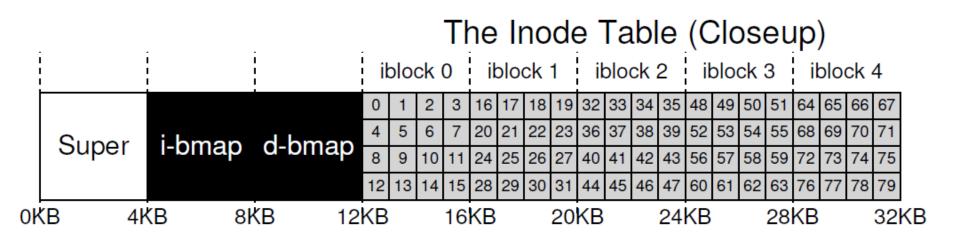
Standalone file systems

- Data structures
 - Data blocks
 - Metadata blocks (Inodes)
 - Bitmap blocks (for space allocation)

- Access paths
 - Read a file
 - Write a file

Inode (index node)

- Each is identified by a number
 - Low-level number of file name: inumber
- Can figure out location of inode from inumber

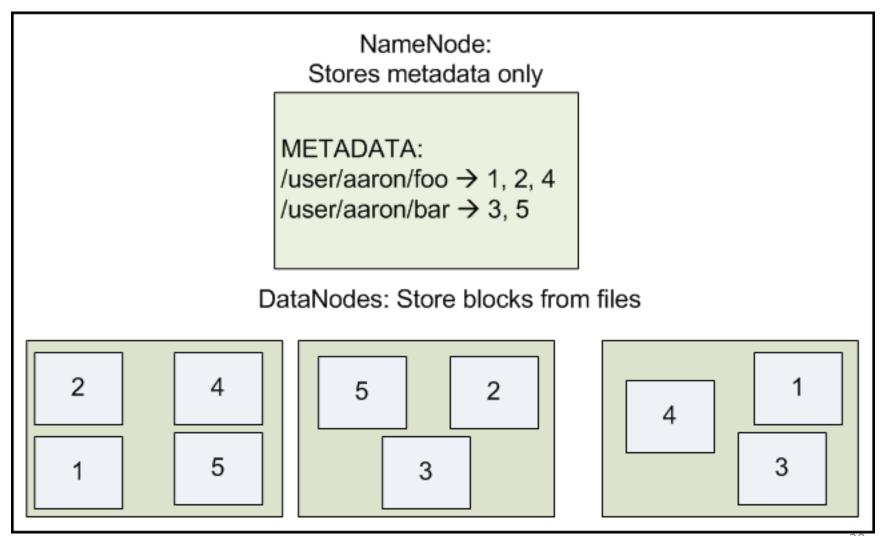


Distributed file systems

- Hadoop HDFS (after GFS)
 - Data are distributed among data nodes
- Replication
 - Automatic creation of replica (typically 2 or 3 copies/replica of data)

- Fault-tolerant
 - Automatic recovery from node failure

HDFS architecture



File system image in namenode

```
V/ LI dout ecourtigoectron/
▼<INodeSection>
   <lastInodeId>16422/lastInodeId>
   <numInodes>38</numInodes>
 ▼<inode>
     <id>16385</id>
     <type>DIRECTORY</type>
     <name/>
     <mtime>1581231015982
     <permission>ec2-user:supergroup:0755</permission>
     <nsquota>9223372036854775807/nsquota>
     <dsquota>-1</dsquota>
   </inode>
 ▼<inode>
     <id>16386</id>
     <type>DIRECTORY</type>
     <name>user</name>
     <mtime>1581231034866
     <permission>ec2-user:supergroup:0755</permission>
     <nsquota>-1</nsquota>
     <dsquota>-1</dsquota>
   </inode>
```

Directory section

```
</snapshorsection>
▼ <INodeDirectorySection>
 ▼<directory>
     <parent>16385</parent>
     <child>16386</child>
   </directory>
 ▼<directory>
     <parent>16386</parent>
     <child>16387</child>
   </directory>
 ▼ <directory>
     <parent>16387</parent>
     <child>16390</child>
     <child>16412</child>
     <child>16401</child>
     <child>16391</child>
     <child>16388</child>
   </directory>
 ▼ <directory>
     <parent>16388</parent>
     <child>16389</child>
   </directorys
```

Major topics

Storage systems

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Database management systems

Big data solution stack

File Formats

JSON

```
"firstName": "John",
"lastName": "Smith",
"isAlive": true,
"age": 25,
"address": {
  "streetAddress": "21 2nd Street",
  "city": "New York",
 "state": "NY",
  "postalCode": "10021-3100"
},
"phoneNumbers": [
    "type": "home",
    "number": "212 555-1234"
  },
    "type": "office",
    "number": "646 555-4567"
"children": [],
"spouse": null
```

HTML

```
<h1> Bibliography </h1>
<i> Foundations of Databases </i>
     Abiteboul, Hull, Vianu
     <br/>
<br/>
dison Wesley, 1995
<i> Data on the Web </i>
     Abiteoul, Buneman, Suciu
     <br/>
<br/>
dr> Morgan Kaufmann, 1999
```

XML

```
<br/>
<br/>
dibliography>
    <book> <title> Foundations... </title>
             <author> Abiteboul </author>
             <author> Hull </author>
             <author> Vianu </author>
             <publisher> Addison Wesley </publisher>
             <year> 1995 
    </book>
</bibliography>
```

XML usages

- Software configurations files
 - E.g., HDFS

- Android app development
 - Layout resource files, e.g., activity_main.xml

- Java archive (.jar file)
 - Manifest.xml

Android app resource file

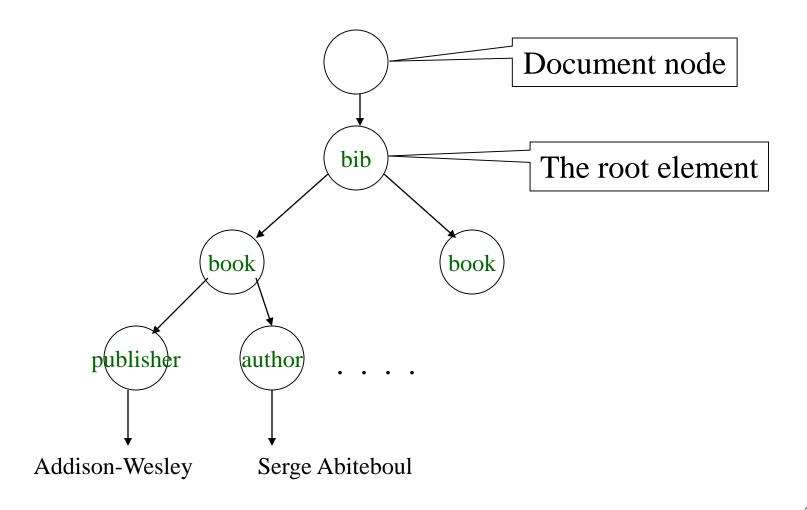
```
<?xml version="1.0" encoding="utf-8"?>
><RelativeLayout xmlns:android="http://schemas.android.com/apk/res/android"</pre>
    xmlns:tools="http://schemas.android.com/tools"
     android:layout width="match parent"
    android:layout height="match parent"
     tools:context=".MainActivity">
     <android.support.design.widget.TabLayout</pre>
         android:id="@+id/tabs"
         android:layout width="match parent"
         android:layout height="wrap content" />
     <android.support.v4.view.ViewPager</pre>
         android:id="@+id/container"
         android:layout width="match parent"
         android:layout height="match parent"
         android:layout below="@id/tabs" />
```

Manifest.xml

```
<?xml version="1.0" encoding="utf-8"?>
<manifest xmlns:android="http://schemas.android.com/apk/res/android"</pre>
   package="com.google.firebase.guickstart.database">
    <uses-permission android:name="android.permission.INTERNET" />
    <application
        android:allowBackup="true"
        android:icon="@mipmap/ic launcher"
        android:label="Firebase Database"
        android:supportsRtl="true"
        android:theme="@style/AppTheme">
        <activity
            android:name=".MainActivity"
            android: label="Firebase Database"
            android:theme="@style/AppTheme" />
        <activity android:name=".NewPostActivity" />
        <activity android:name=".SignInActivity">
            <intent-filter>
                <action android:name="android.intent.action.MAIN" />
                <category android:name="android.intent.category.LAUNCHER" />
            </intent-filter>
```

```
<bib>
<book price="35">
   <publisher>Addison-Wesley</publisher>
    <author>Serge Abiteboul</author>
    <author><first-name>Rick</first-name><last-name>Hull</last-name></author>
    <author age="20">Victor Vianu</author>
    <title>Foundations of Databases</title>
    <year>1995</year>
   <price>38.8</price>
</book>
<book price="55">
    <publisher>Freeman</publisher>
    <author>Jeffrey D. Ullman</author>
    <title>Principles of Database and Knowledge Base Systems</title>
    <year>1998</year>
</book>
</bib>
```

Data Model for XPath



XPath: Simple Expressions

```
/bib/book/year
```

```
Result: <year> 1995 </year> <year> 1998 </year>
```

/bib/paper/year

Result: empty (there were no papers)

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Storage systems

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Database management systems



Big data solution stack

Relational DBMS

- Data models
 - E (entity set) R
 - Relational (redundancy => update anomaly)

Schema

- Normal forms, e.g., 3NF, BCNF (not required, unless we talk about data warehousing)
- Multidimensional model
- Fact table + dimension (product, customer, time) tables
- Star/snowflake
- ETL: insert into ... (warehouse)
 - select ... From ... (data sources)

- Query languages
 - Relational algebra
 - SQL, constraints, views

- Data organization
 - Records and blocks
 - Index structure: B+-tree (external data structure)

- Query execution algorithms
 - External sorting
 - One-pass algorithms
 - Nested-loop join, sorting, hashing-based
 - Multiple-pass algorithms

Rigid schema

- Strong consistency is the key design goal
 - Never read old data
 - Suitable for mission-critical applications, e.g., banking
- But may suffer from low availability
 - ACID vs CAP

- Hard to scale out
 - Horizontal partitioning/sharding possible
 - But would need distributed storage & computing support like Hadoop & MapReduce

RDBMS Examples

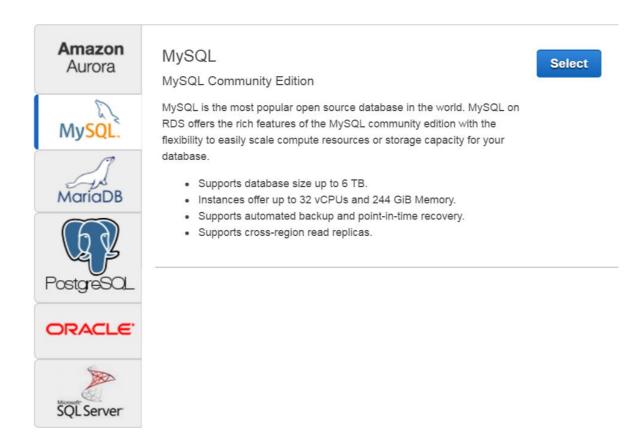
MySQL (can be installed in Amazon AWS EC2)

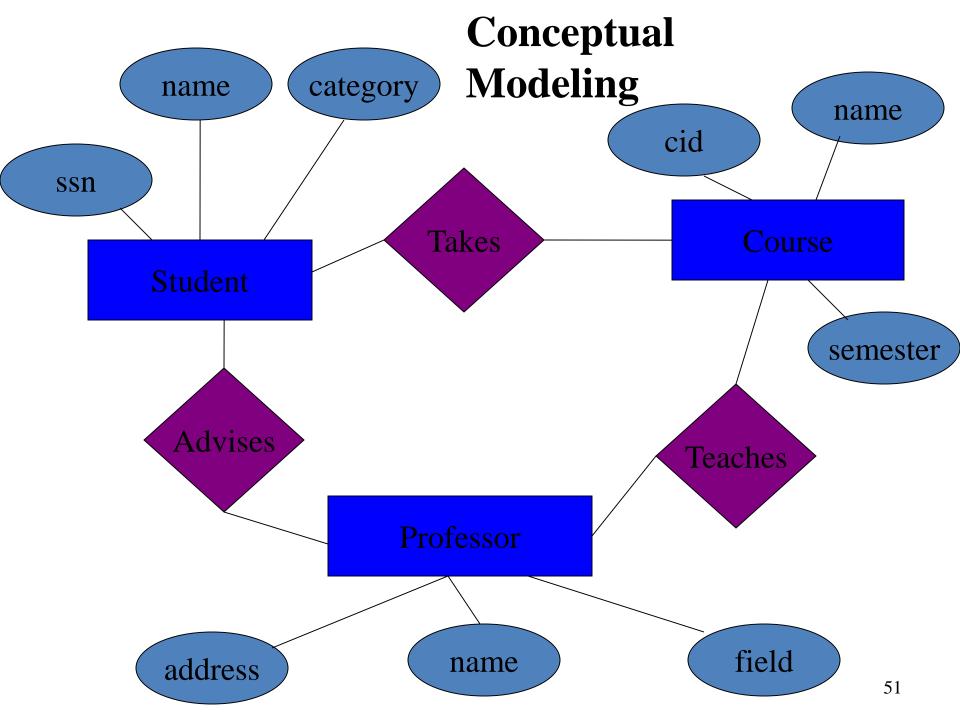
- Amazon RDS (Relational database as a service)
 - DBMS in the cloud
 - Database as a service

- Data warehouse on RDBMS
 - OLAP

Amazon RDS: Database-as-a-service

MySQL, PostgreSQL, Oracle, SQL Server, etc.





Schema Design and Implementation

Tables (relations):

Students:

SSN	Name	Category
123-45-6789	Charles	undergrad
234-56-7890	Dan	grad

Takes:

SSN	CID
123-45-6789	CSE444
123-45-6789	CSE444
234-56-7890	CSE142

Courses:

CID	Name	Semster
CSE444	Databases	fall
CSE541	Operating systems	spring

Separates the logical view from the physical view of the data.

Querying a Database

- Find all courses that "Mary" takes
- S(tructured) Q(uery) L(anguage)

Select A's ,agg
From R's
Where C's
Group by A's
Having
Order by
Limit ?
Offset ?
(pagination)

===

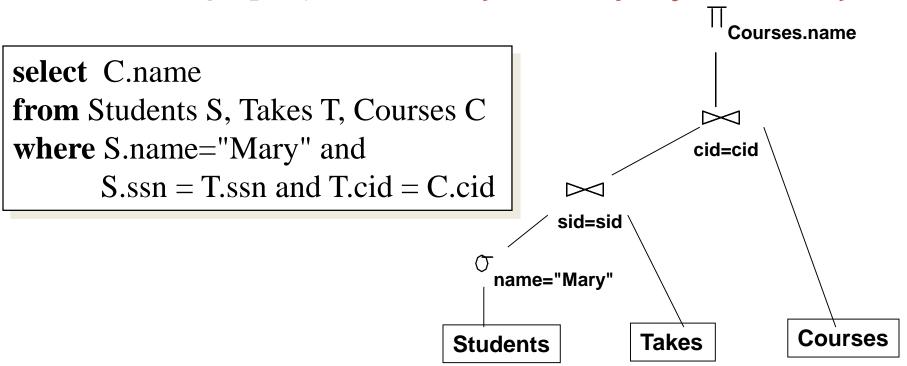
Insert Update Delete

Declarative (what)

 Query processor figures out how to answer the query efficiently.

Query Optimization

Goal:



<u>Plan:</u> tree of Relational Algebra operators, choice of algorithms at each operator

Major topics

Storage systems

File systems & file formats

Database management systems

Big data solution stack



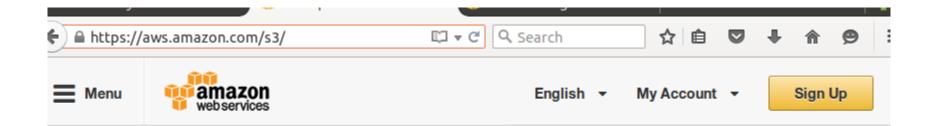
Topics

- Big data management & analytics
 - Cloud data storage (Amazon S3)
 - NoSQL
 - Google Firebase (real-time database, ...)
 - MongoDB (shell, mongo)
 - Amazon DynamoDB (row store, key-value)
 - Cassandra (not required)
 - Apache Hadoop & MapReduce
 - Apache Spark

Cloud data storage

- Amazon S3 (simple storage service)
 - Ideal for storing large binary files
 - E.g., audio, video, image
 - Simple RESTful web service

Eventual consistency for high availability



PRODUCTS & SERVICES > Amazon S3 **Product Details** > Storage Classes > Pricing > Getting Started > FAQs > Resources > Amazon S3 SLA > RELATED LINKS AWS Management Console Documentation

Release Notes

Amazon S3

Amazon Simple Storage Service (Amazon S3), provides developers and IT teams with secure, durable, highly-scalable object storage. Amazon S3 is easy to use, with a simple web service interface to store and retrieve any amount of data from anywhere on the web. With Amazon S3, you pay only for the storage you actually use. There is no minimum fee and no setup cost.

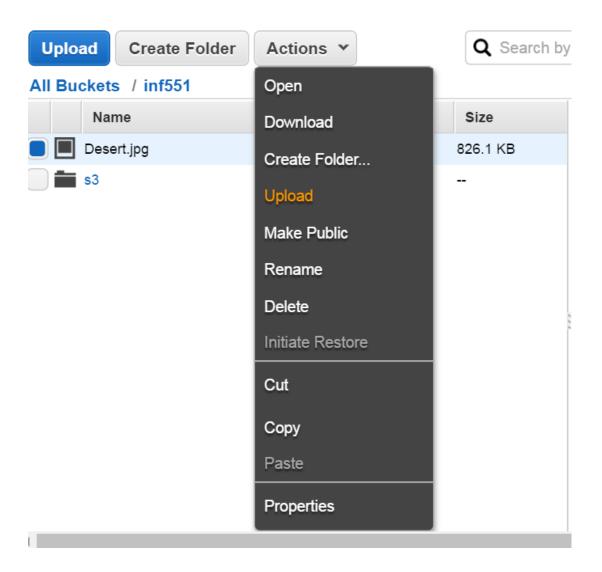
Amazon S3 offers a range of storage classes designed for different use cases including Amazon S3 Standard for generalpurpose storage of frequently accessed data, Amazon S3 Standard - Infrequent



In Recent News

New: Amazon VPC

Upload a file







← → X 🔓 https://s3.amazonaws.com/inf551/Desert.jpg



NoSQL

- Not only SQL
- Flexible schemas
 - e.g., JSON documents or key-value pairs
 - Ideal for managing a mix of structured, semistructured, and unstructured data
- High availability (CAP)
- Weaker (e.g., eventual) consistency model

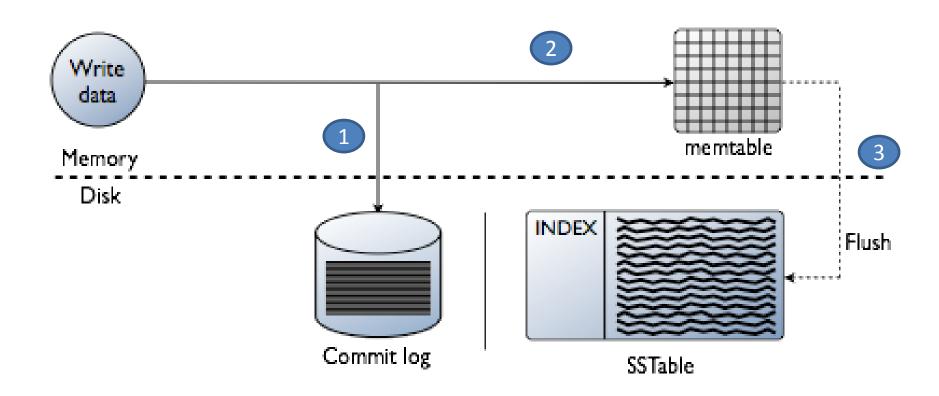
Example NoSQL databases

- MongoDB, Firebase, etc.
 - Manage JSON documents
- Amazon DynamoDB
 - Row store
 - row = item = a collection of key-value pairs
- Apache Cassandra (not required)
 - Wide column store
 - Google's Bigtable clone
- Neo4J...

Key techniques

- Consistent hashing (Cassandra, Dynamo)
 - Avoid moving too much data when adding new machines (scaling out)
- Efficient writes (for update-heavy apps)
 - Append-only
 - No overwrites
 - Avoid random seek
 - But compaction needed later

Write path in Cassandra



Key techniques

- Compaction
 - Introduced in Google "Bigtable" paper
 - Merge multiple versions of data
 - Remove expired or deleted data

DynamoDB

 https://console.aws.amazon.com/dynamodb/ home?region=us-east-1#gettingStarted:

Amazon DynamoDB

Amazon DynamoDB is a fast and flexible NoSQL database service for all applications that need consistent, single-digit millisecond latency at any scale. Its flexible data model and reliable performance make it a great fit for mobile, web, gaming, ad-tech, IoT, and many other applications.

Create table

Create DynamoDB table

Tutorial



DynamoDB is a schema-less database that only requires a table name and primary key. The table's primary key is made up of one or two attributes that uniquely identify items, partition the data, and sort data within each partition.

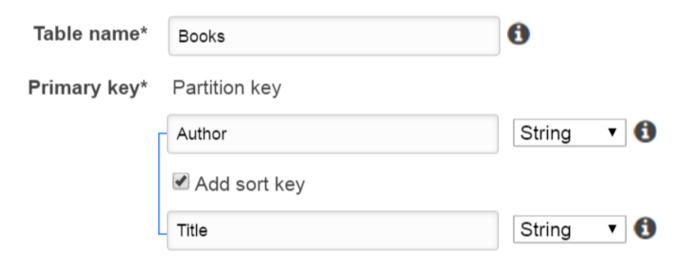
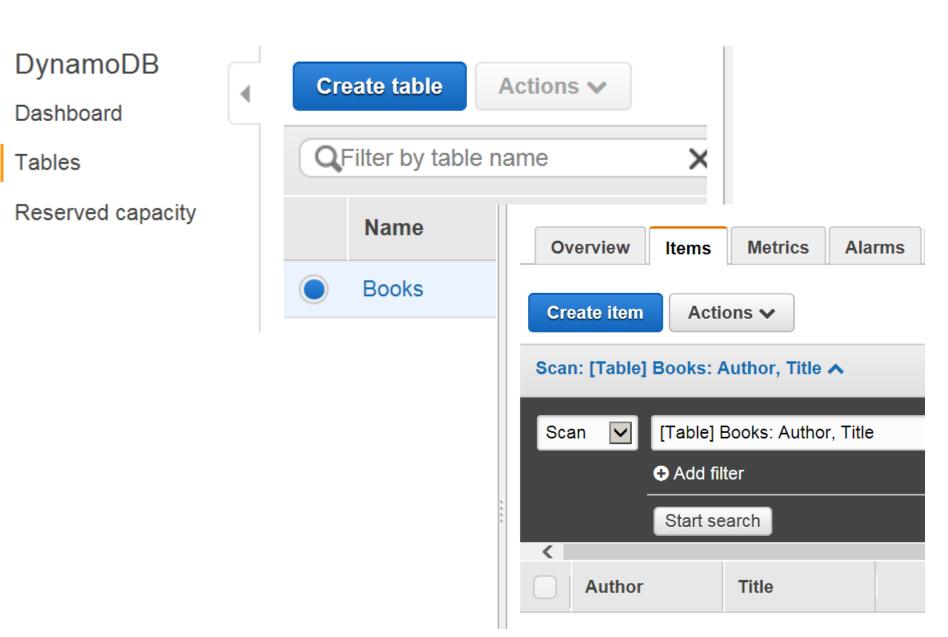


Table settings

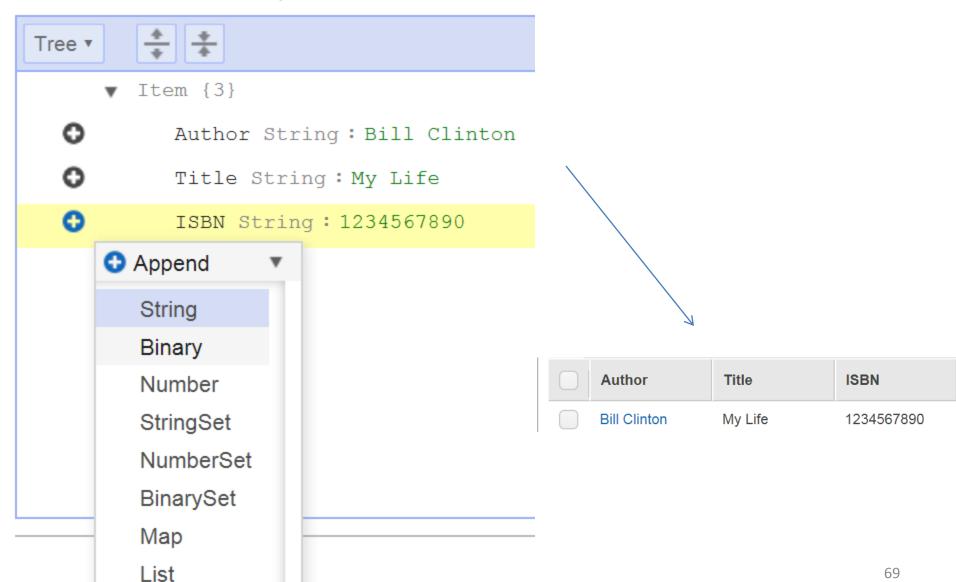
Default settings provide the fastest way to get started with your table. You can modify these default settings now or after your table has been created.

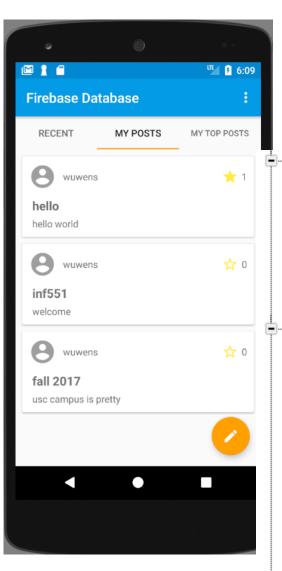
Use default settings

Insert items



May add new attributes





Firebase: a cloud database

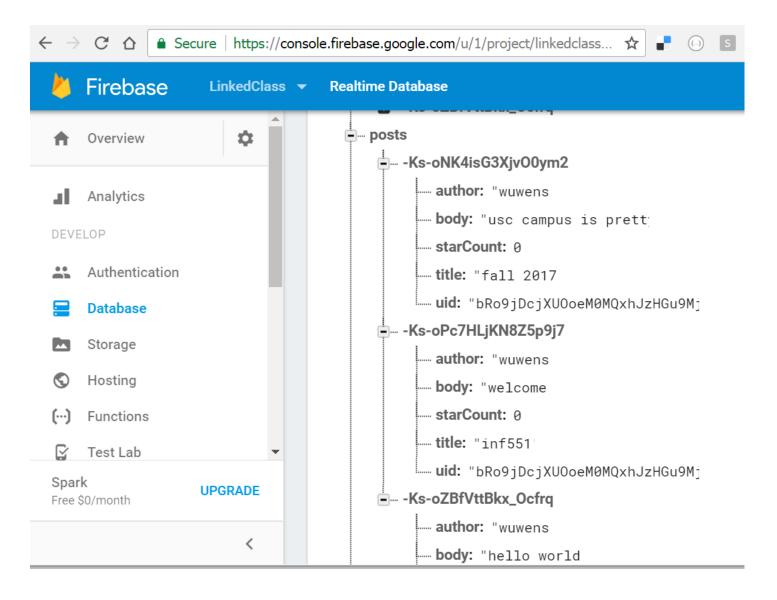
```
impost-comments
    -Ks-oZBfVttBkx_Ocfrq
        -Ks-otimnHiahFzpzqvY
               author: "wuwens
               text: "hello hello
              - uid: "bRo9jDcjXUOoeM0MQ

    posts

   -Ks-oNK4isG3Xjv00ym2 + ×
    -Ks-oPc7HLjKN8Z5p9j7
    -Ks-oZBfVttBkx_Ocfrq
           author: "wuwens
           body: "hello world
           starCount: 1
        🗖 ... stars
          --- title: "hello'
           uid: "bRo9jDcjXUOoeM0MQxhJz
```

```
"post-comments" : {
  "-Ks-oZBfVttBkx Ocfrq" : {
    "-Ks-otimnHiahFzpzqvY" : {
      "author" : "wuwens",
      "text" : "hello hello",
      "uid": "bRo9jDcjXUOoeM0MQxhJzHGu9Mj2"
"posts" : {
  "-Ks-oNK4isG3XjvO0ym2" : {
    "author" : "wuwens",
    "body" : "usc campus is pretty",
    "starCount" : 0,
    "title" : "fall 2017",
    "uid": "bRo9jDcjXUOoeM0MQxhJzHGu9Mj2"
  "-Ks-oPc7HLjKN8Z5p9j7" : {
    "author" : "wuwens",
    "body" : "welcome",
    "starCount" : 0,
    "title": "inf551",
    "uid": "bRo9jDcjXUOoeM0MQxhJzHGu9Mj2"
  "-Ks-oZBfVttBkx Ocfrq" : {
    "author" : "wuwens",
    "body" : "hello world",
    "starCount" : 1,
    "stars" : {
      "bRo9jDcjXUOoeM0MQxhJzHGu9Mj2": true
    },
    "title" : "hello",
    "uid": "bRo9jDcjXUOoeM0MQxhJzHGu9Mj2"
```

Firebase



Topics

- Big data management & analytics
 - Cloud data storage (Amazon S3)
 - NoSQL (Amazon DynamoDB, Cassandra, MongoDB)
 - MapReduce



- Apache Hadoop
- Apache Spark

Roots in functional programming

- Functional programming languages:
 - Python, Lisp (list processor), Scheme, Erlang, Haskell
- Two functions:
 - Map: mapping a list => list
 - Reduce: reducing a list => value
- map() and reduce() in Python
 - https://docs.python.org/2/library/functions.html#ma
 p

map() and reduce() in Python

- list = [1, 2, 3]
- def sqr(x): return x ** 2
- list1 = map(sqr, list)

What are the value of list1 and z?

- def add(x, y): return x + y
- z = reduce(add, list)

reduce() is in functools module of Python 3

Lambda function

Anonymous function (not bound to a name)

• list = [1, 2, 3]

- list1 = map(lambda x: x ** 2, list)
- z = reduce(lambda x, y: x + y, list)

How is reduce() in Python evaluated?

z = reduce(f, list) where f is add function

- Initially, z (an accumulator) is set to list[0]
- Next, repeat z = add(z, list[i]) for each i > 0
- Return final z

Example: z = reduce(add, [1, 2, 3])
 - i = 0, z = 1; i = 1, z = 3; i = 2, z = 6

Hadoop MapReduce

Map

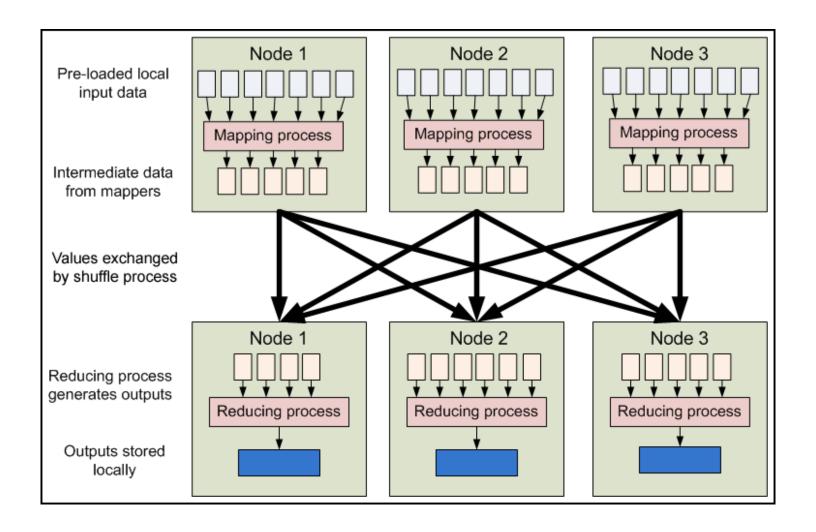
$$- \langle k, v \rangle =>$$
 list of $\langle k', v' \rangle$

• Reduce:

$$- \langle k', \text{ list of } v' \rangle => \text{ list of } \langle k'', v'' \rangle$$

- Write MapReduce programs on Hadoop
 - Using Java

MapReduce



WordCount: mapper

Object can be replaced with LongWritable

```
Data types of input key-value
public class WordCount {
                                             Data types of output key-value
  public static class Tokenizer Mapper
       extends Mapper object, Text, Text, IntWritable>{
    private final static IntWritable one = new/ IntWritable(1);
    private Text word = new Text();
    public void map(Object key, Text value,/Context context
                       throws IUException, interruptedException {
      StringTokenizer itr = new StringTokenizer(value.toString());
      while (itr.hasMoreTokens()) {
        word.set(itr.nextToken());
        context.write(word, one);
                           Key-value pairs with specified data types
```

WordCount: reducer

Data types of input key-value Data types of output key-value public static class IntSumReducer extends Reducer<Text,IntWritable,Text,IntWritable> private IntWritable result = new IntWritable(); public void reduce(Text key, Iterable<IntWritable> values, Context context) throws IOException, InterruptedException { int sum = 0; for (IntWritable val : values) { sum += val.qet(); A list of values result.set(sum); context.write(key, result);

Characteristics of Hadoop

- Acyclic data flow model
 - Data loaded from stable storage (e.g., HDFS)
 - Processed through a sequence of steps
 - Results written to disk

- Batch processing
 - No interactions permitted during processing

Problems

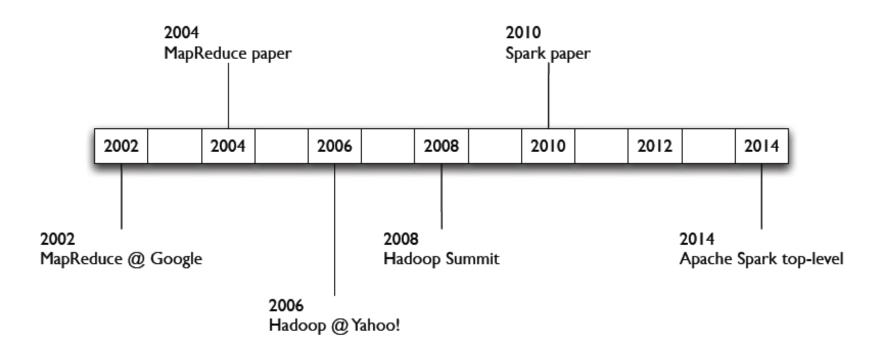
- Ill-suited for iterative algorithms that requires repeated reuse of data
 - E.g., machine learning and data mining algorithms such as k-means, PageRank, logistic regression

- Ill-suited for interactive exploration of data
 - E.g., OLAP on big data

In-memory MapReduce (Spark)

- Key concepts
 - RDD (resilient distributed dataset)
 - Transformations
 - Actions

Apache Spark: history



Spark

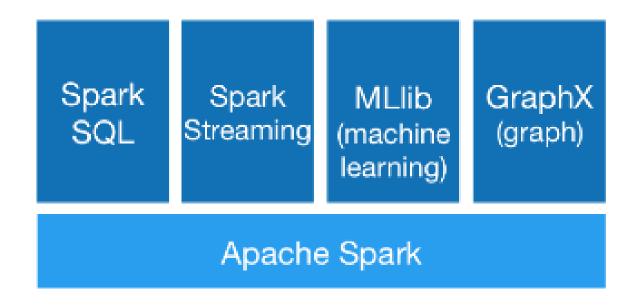
- Support working sets through RDD
 - Enabling reuse & fault-tolerance

10x faster than Hadoop in iterative jobs

Interactively explore 39GB with sub-second response time

Spark

- Combine SQL, streaming, and complex analytics
- We will see DataFrame in Spark too



Spark

Run on Hadoop, Cassandra, HBase, etc.











wc.py

```
from pyspark import SparkContext
from operator import add
sc = SparkContext(appName="dsci551")
lines = sc.textFile('hello.txt')
counts = lines.flatMap(lambda x: x.split(' ')) \
       .map(lambda x: (x, 1)) \
       .reduceByKey(add)
output = counts.collect()
for v in output:
  print(v[0], v[1])
```

Lab session

Task: Setting up an EC2 instance

• Details: see lab session slides to be posted...

Resources

Merge sort:

- https://www.interviewbit.com/tutorial/merge-sortalgorithm/
- https://www.youtube.com/watch?v=Nso25TkBsYl

Hashing

- https://www.tutorialspoint.com/python data structu re/python hash table.htm
- https://www.programiz.com/pythonprogramming/methods/built-in/hash