WELCOME!!!

CS 6650 *Scalable* Distributed Systems

(while maintaining some sanity?!) ©

What are we doing?

- Weekly class time
 - Yvonne talking fast but recorded!
 - Industry mentors visiting
 - Working together to get the crazy code to RUN on AWS (or where you want to?)!!!
- Weekly mock interviews (sign up!) to walk through your code with the teaching team!

Please read the syllabus and ask any questions!

How are we going to do it?!

- Weekly homeworks and Masteries are reviewed in mock interviews
 - 1:1 time with your TA!
- Homework will cover
 - Basics of concurrency and distributed system architectures
 - Server processing
 - Data, Data and more Data
 - Scalability issues throughout
- We will spend class time on homeworks together!
- Canvas, Teams, ... talk to us!

The Homeworks build into larger Assignments!

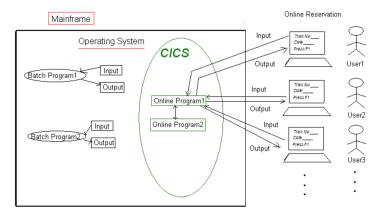
- Client
 - Multithreaded exerting request load on dumb server to start ©
 - Infrastructure to measure is on client-side
- Server evolves!
 - Database introduced along with "business logic"
 - Data storage management
 - Write-oriented workloads (making it harder to scale)
 - Threading on server
 - State management, horizontal scaling

Server is Still evolving!

- Now scaling across more servers
 - Queuing for better responsiveness
 - Caching for performance
 - Monitoring
- And finally! ☺
 - Distributed database issues
 - More design patterns for data intensive computing
- Project ideas?!?

Quick intro, and history lesson!

- CICS IBM in the 60s! (still around)
 - Customer Information Control System
 - Mainframe and users
 - Timeshared server, dumb terminal
 - MTS (Michigan Terminal System)
- Along comes the Internet in the 90s
 - LANs, WANs, modems, and browsers (slow and simple!)
 - Eventually needed to start thinking about scale!
 - How many users/clients are hitting the banking server?
 - Amazon, Google, ebay?
 - Now we have MASSIVE data centers all over the world!
 - If we can process it!!!
 - Check out petabyte scale data sets
 - https://www.lifewire.com/terabytes-gigabytes-amp-petabytes-how-big-are-they-4125169



http://www.mainframegurukul.com/tuto rials/programming/cics/introduction-tocics.html



Software fabric that make distributed systems!

- Remote Procedure Calls
 - Java RMI
- Sockets (80s!)
 - Websockets
- CORBA (remote objects)
- HTTP
- More evolution to come...
 - Languages like Go and Rust!
 - Dask?!
 - Projects! ©

Applications today

- These are just a few...
- Numbers a little out of date
- BUT YOU GET THE IDEA!

- More applications coming
 - Your project?!
 - The Earth Data Store

<u>Dropbox</u> – <u>exascale storage system</u>, millions of new files per hour, from 2012-16 seen over 12x growth.

Gmail – 1.2 billion users, blocks 10M spam messages every minute

Netflix – runs on AWS, took 7 years to migrate, uses 15% of global downstream internet traffic

Facebook - More than 300 million photos uploaded per day and 510,000 comments posted and 293,000 statuses updated per minute

Youtube - Users watch <u>4,146,600 YouTube videos</u> every minute

What do these systems all have in common?

- Transactions, user interactions, analysis
- Many more have sensors, IoT
- Many more now have rich UIs
 - AR/VR!
- This means we have large code bases along with these massive data sets!
- And large data sets of performance monitoring results!
- They operate on multiple "nodes", and are distributed systems

Nonfuctional Requirements (quality attributes)

- Performance, performance, performance
 - NOTE: collecting data on performance generates a lot of data!
 - Analysis of this is also challenging
- Scale
 - Performance is still good, even with more data, more users, more everything!
 - Note this can be very dynamic spikes
 - A Tweet from a well known person!
- Security
 - Data and access control
- Available
 - 24/7 (Leah knows about stress!)

What is SCALABILITY?

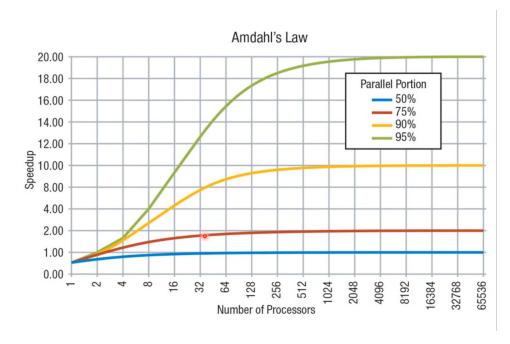
 The ability of a system to increase or decrease capacity in response to changing demands while continuing to satify service level agreements for latency and availability

"Elastic Compute"

 Hyperscalability is the same as the above, but while supporting exponential growth with linear costs

Can we just throw hardware at it?

- Yes and no!
- We want linear "scale up"
 - With each additional machine, we want to be able to grow linearly
 - BUT there is Amdahl's Law (wikipedia here!)
 - At some point we level out
 - Not getting faster because some things have to be done in sequence, one at a time
 - Example, a write to a database
 - "Serial bottlenecks"!

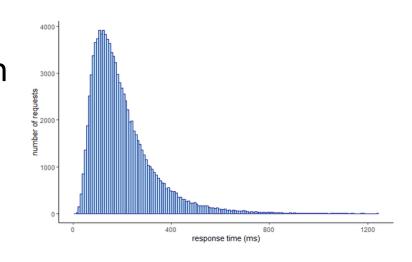


SOME problems are easily parallelizable

- Map/Reduce (Google paper, please share your thoughts on Piazza!)
- Some serialized behaviour but does well!
- Can be a little tricky to get right...
 - Data Skew
 - Breaking the data set down into "chunks"
 - Not every chunk takes the same amount of time to execute
- Please take a look at the paper (link provided on Canvas) and share your thoughts!

Response times are critical!

- Within 40 milliseconds
 - A user will LEAVE!
 - Impacts google rankings!
 - There are delays in data transfer (latencies per request)
 - You will love measuring these!
- Fun fact is that there is a VERY LONG TAIL graph
 - Slow responses DO happen
 - So how can we mitigate those?!



Throughput

- Typically reported as requests per second
 - Note: Latency is for a single request
 - Throughput is a bulk measure of many requests
 - You will love measuring this!
- When you have millions of requests per second and bursts
 - We need to split the load across multiple nodes!
- "Scale-out"
 - Add more nodes to increase throughput...
 - Does it always work?

Availability!

- 24/7
 - 99%, 99.999%, 99.9999%...

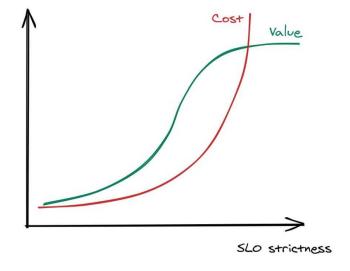
Availability %	Downtime per year ^[note 1]	Downtime per quarter	Downtime per month	Downtime per week	Downtime per day (24 hours)
90% ("one nine")	36.53 days	9.13 days	73.05 hours	16.80 hours	2.40 hours
95% ("one and a half nines")	18.26 days	4.56 days	36.53 hours	8.40 hours	1.20 hours
97%	10.96 days	2.74 days	21.92 hours	5.04 hours	43.20 minutes
98%	7.31 days	43.86 hours	14.61 hours	3.36 hours	28.80 minutes
99% ("two nines")	3.65 days	21.9 hours	7.31 hours	1.68 hours	14.40 minutes
99.5% ("two and a half nines")	1.83 days	10.98 hours	3.65 hours	50.40 minutes	7.20 minutes
99.8%	17.53 hours	4.38 hours	87.66 minutes	20.16 minutes	2.88 minutes
99.9% ("three nines")	8.77 hours	2.19 hours	43.83 minutes	10.08 minutes	1.44 minutes
99.95% ("three and a half nines")	4.38 hours	65.7 minutes	21.92 minutes	5.04 minutes	43.20 seconds
99.99% ("four nines")	52.60 minutes	13.15 minutes	4.38 minutes	1.01 minutes	8.64 seconds
99.995% ("four and a half nines")	26.30 minutes	6.57 minutes	2.19 minutes	30.24 seconds	4.32 seconds
99.999% ("five nines")	5.26 minutes	1.31 minutes	26.30 seconds	6.05 seconds	864.00 milliseconds
99.9999% ("six nines")	31.56 seconds	7.89 seconds	2.63 seconds	604.80 milliseconds	86.40 milliseconds
99.99999% ("seven nines")	3.16 seconds	0.79 seconds	262.98 milliseconds	60.48 milliseconds	8.64 milliseconds
99.999999% ("eight nines")	315.58 milliseconds	78.89 milliseconds	26.30 milliseconds	6.05 milliseconds	864.00 microseconds
99.999999% ("nine nines")	31.56 milliseconds	7.89 milliseconds	2.63 milliseconds	604.80 microseconds	86.40 microseconds

https://en.wikipedia.org/wiki/High_availability

- Don't want a single point of failure (SPoF)
 - So we need to make sure we have replicas/redundancy!
- When you have 1000s of components, some will fail!
 - Need to incorporate FAILURE into our designs

It is all about the *Tradeoffs*!

- Throughput versus cost
 - More hardware means more \$\$
 - When is it worth it? When isn't it?



https://robertovitillo.com/why-you-should-measure-tail-latencies/

- Performance versus availability
 - Fast but components may fail!
 - Remember the old mainframes, they were fast-ish ©
- Need to be able to measure and discuss the tradeoffs in the work you do in this course!
- This all depends on the context of the requirements of the system you are building, there is no ONE right answer!