

### Fallacies of Distributed computing

and their effect on Cloud Solutions



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#### What is this session about?

- What to think about when building data pipelines in Cloud platforms
- Managed is a big deal but can we take it for granted?
- Best practices for architecting Cloud Data platforms (and how some of the veteran cloud customers deal with it)

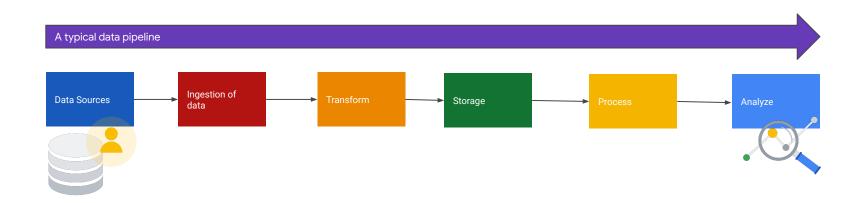
# This is the opposite of a cloud sales pitch: But don't forget the following..

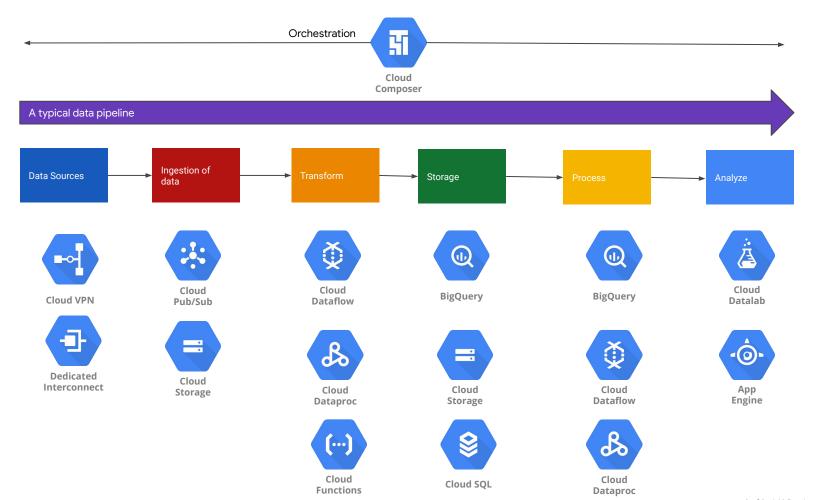


"just as flying is still the safest form of travel in terms of number of fatalities per miles travelled, cloud computing services provide as high—or higher—levels of availability and reliability as enterprise data centers."

# Let's look at a typical data pipeline

How would you build it in the cloud?



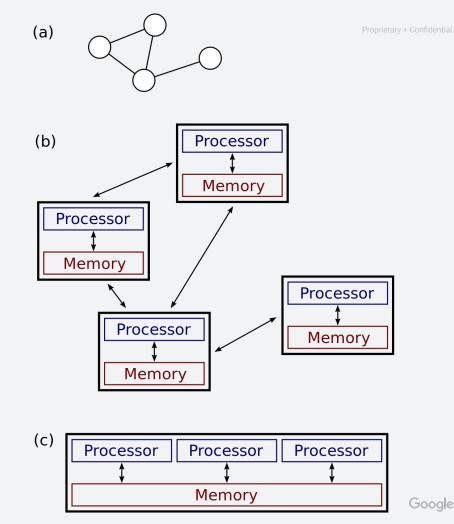


# = Network component / Node The Reality is.. Ingestion / Staging Data Sources Storage (m)

# What is a Distributed System?

## What is a Distributed System?

A distributed system is a system whose components are located on different networked computers, which communicate and coordinate their actions by passing messages to one another.



### 

## = Network component / Node The Reality is.. Ingestion / Data Sources Storage Staging Not enough bandwidth (m)

#### = Network component / Node The Reality is.. Ingestion / Data Sources Storage Staging High latency API Throttle response time for a Concurrency limit small query exceeded.. (m)Too many requests z)

#### = Network component / Node The Reality is.... Things will always fail, at cloud scale it fails more frequently.. Ingestion / Data Sources Storage Staging Too many worker nodes failures. Dynamic rebalancing (m)causing latency

z)

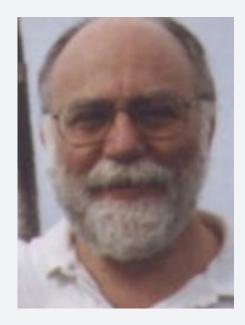
# What do we take for granted here?

"The fallacies of Distributed Computing"



# The fallacies of Distributed Computing

- 1. The network is reliable.
- 2. Latency is zero.
- 3. Bandwidth is infinite.
- 4. The network is secure.
- 5. Topology doesn't change.
- 6. There is one administrator.
- 7. Transport cost is zero.
- 8. The network is homogeneous.



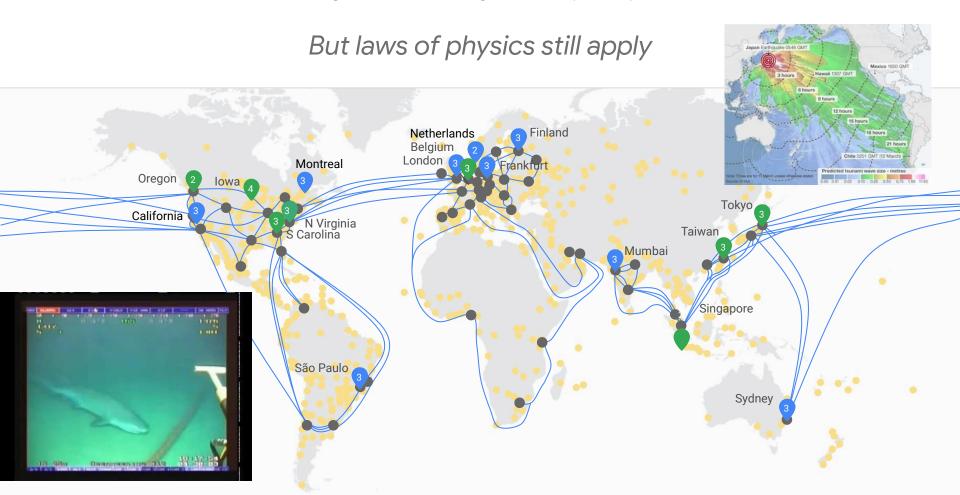
Peter Deutsch - Sun Microsystems, 1994

### How much time you would spend uploading data?

Close Far 100 PB 34,048 years 124 days 340 years 3,404 years 3 years 34 years 10 PB 12 days 3 years 340 years 124 days 34 years 3,404 years 1 PB 12 days 124 days 340 years 30 hours 3 years 34 years 12 days 124 days 100 TB 3 hours 30 hours 3 years 34 years 10 TB 18 minutes 3 hours 30 hours 12 days 124 days 3 years 1 TB 2 minutes 18 minutes 3 hours 30 hours 12 days 124 days 100 GB 11 seconds 2 minutes 18 minutes 3 hours 30 hours 12 days 10 GB 1 second 11 seconds 2 minutes 18 minutes 3 hours 30 hours 1 GB 1 second 2 minutes 18 minutes 0.1 seconds 11 seconds 3 hours 100 Gbps 10 Gbps 1 Gbps 100 Mbps 10 Mbps 1 Mbps

Data Size

Hundreds of thousands of miles of fiber optic cable connecting all of our datacenter regions and 100+ points of presence.



So, what should Design for failure:)
we do?

Start your solution design with the end user in mind...

# The fallacies of Distributed Computing

Meets

## Design principles for Cloud Solutions

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- 1. Design for failure
- 2. Loose coupling
- 3. Implement "Elasticity"
- 4. Build Security in every layer
- 5. Don't fear constraints
- 6. Think Parallel
- 7. Leverage different storage options

This doesn't matter anymore, we have achieved a good level of abstraction

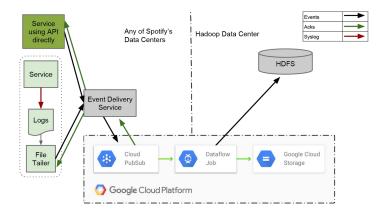
<sup>\*</sup> These best practices for Architecting in Cloud was introduced by AWS in 2011. Google https://aws.amazon.com/whitepapers/architecting-for-the-aws-cloud-best-practices/

## Design principles for Cloud Solutions

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- Avoid single points of failure
- Use queues to decouple: Job A passes the output to a queue for Job B to pull from
- Idempotent and Metadata driven scaling new jobs/actors should be bootstrapped to ask "who am I, what am I supposed to do?"
- In a network, location/perimeter security is useless - so wrap all the components with ownership, audit.
- Memory, IOPS distribution by sharding,
   Caching for performance
- **Divide, Parallelize and conquer.** Serial/Sequential days are gone.
- One database does not fit all anymore. Think different stores for the outcome (Graph, Key-value, OLAP)

## Learn from the industry leaders..



"To achieve good end-to-end latency, we wrote our ETL as a streaming job. By having it constantly running, we're able to incrementally fill discrete hourly buckets as the data arrives. This gives us better latency compared to a batch job that exported data once at the end of every hour."

- Igor Maravić, Spotify



Spotify has a great published story about their migration to cloud.

BigQuery	10 million queries per month on scanning over 500PBs
Pub/Sub	Over 1 Trillion messages/day - under 400ms latency
Dataflow	5000+ dataflow jobs a day
Dataproc	Lots of dynamically spun up job scoped clusters

## "The best way to avoid failure is to fail constantly"



Simian Army

Kill/inspect running instances



### Thank You

### END