Distributed Systems

The second half of *Concurrent and Distributed Systems* https://www.cl.cam.ac.uk/teaching/current/ConcDisSys

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Computer Science Tripos, Part IB

Slides and notes by Dr. Martin Kleppmann



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A distributed system is...

 "...a system in which the failure of a computer you didn't even know existed can render your own computer unusable." — Leslie Lamport



A distributed system is...

- "...a system in which the failure of a computer you didn't even know existed can render your own computer unusable." — Leslie Lamport
- ... multiple computers communicating via a network...
- ...trying to achieve some task together
- Consists of "nodes" (computer, phone, car, robot, ...)

Recommended reading

- van Steen & Tanenbaum.
 "Distributed Systems"
 (any ed), free ebook available
- ► Cachin, Guerraoui & Rodrigues. "Introduction to Reliable and Secure Distributed Programming" (2nd ed), Springer 2011
- Kleppmann. "Designing Data-Intensive Applications", O'Reilly 2017
- Bacon & Harris.
 "Operating Systems: Concurrent and Distributed Software Design", Addison-Wesley 2003

Relationships with other courses

- Concurrent Systems Part IB (every distributed system is also concurrent)
- Operating Systems Part IA (inter-process communication, scheduling)
- Databases Part IA (many modern databases are distributed)
- Computer Networking Part IB Lent term (distributed systems involve network communication)
- ► Further Java Part IB Michaelmas (distributed programming practical exercises)
- Cybersecurity Part IB Easter term (network protocols with encryption & authentication)
- Cloud Computing Part II
 (distributed systems for processing large amounts of data)



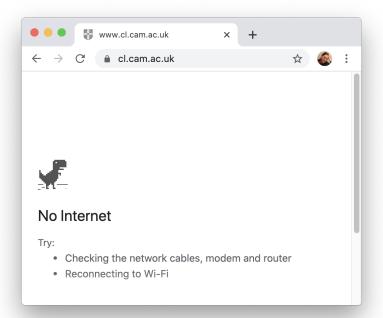
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- For better reliability: even if one node fails, the system as a whole keeps functioning
- ► For better performance: get data from a nearby node rather than one halfway round the world
- ► To solve bigger problems: e.g. huge amounts of data, can't fit on one machine



The trouble with distributed systems:

- Communication may fail (and we might not even know it has failed).
- Processes may crash (and we might not know).
- All of this may happen nondeterministically.

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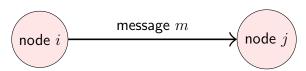
Fault tolerance: we want the system as a whole to continue working, even when some parts are faulty.

This is hard.

Writing a program to run on a single computer is comparatively easy?!

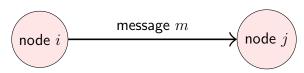
Distributed Systems and Computer Networking

We use a simple abstraction of communication:



Distributed Systems and Computer Networking

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Reality is much more complex:

- Various network operators: eduroam, home DSL, cellular data, coffee shop wifi, submarine cable, satellite...
- ► Physical communication: electric current, radio waves, laser, hard drives in a van...

Hard drives in a van?!



https://docs.aws.amazon.com/snowball/latest/ug/using-device.html

High latency, high bandwidth!

Latency and bandwidth

Latency: time until message arrives

- ▶ In the same building/datacenter: ≈ 1 ms
- ▶ One continent to another: ≈ 100 ms
- ▶ Hard drives in a van: ≈ 1 day

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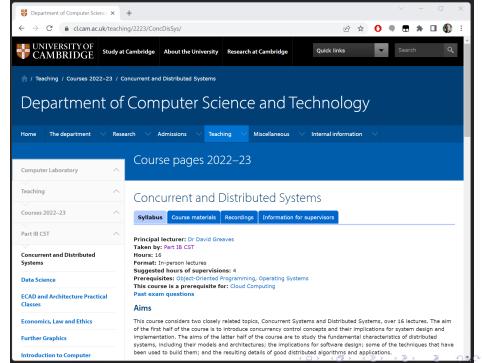
- ▶ In the same building/datacenter: ≈ 1 ms
- ▶ One continent to another: $\approx 100 \text{ ms}$
- ► Hard drives in a van: ≈ 1 day

Bandwidth: data volume per unit time

- ▶ 3G cellular data: ≈ 1 Mbit/s
- ▶ Home broadband: $\approx 10 \text{ Mbit/s}$
- ► Hard drives in a van: 50 TB/box ≈ 1 Gbit/s

(Very rough numbers, vary hugely in practice!)





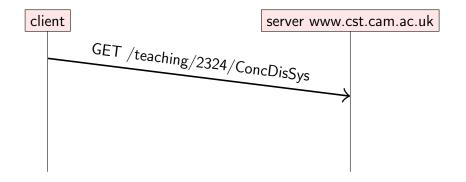
Client-server example: the web

Time flows from top to bottom.



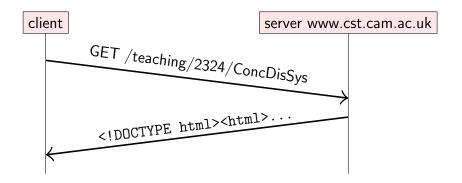
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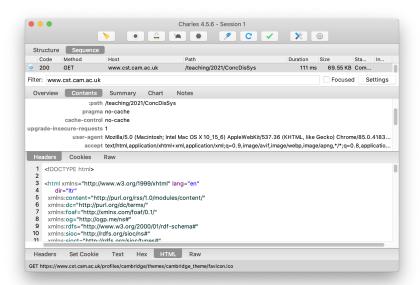
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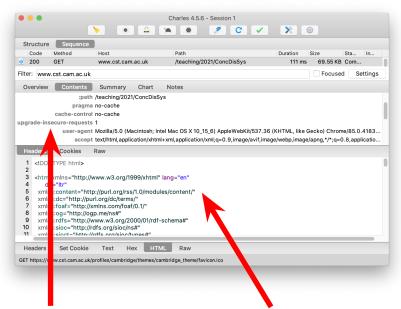


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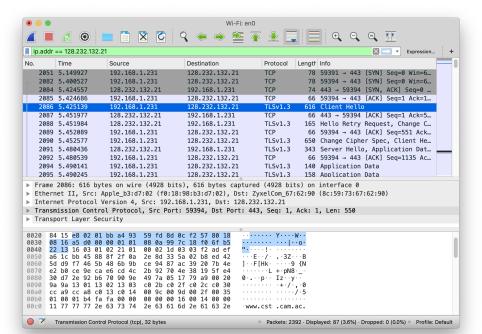






request message

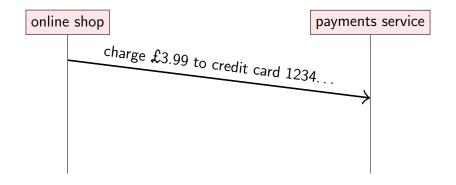
response message



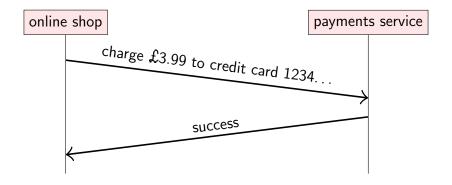
Client-server example: online payments



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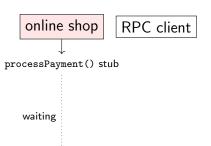
Remote Procedure Call (RPC) example

```
// Online shop handling customer's card details
Card card = new Card():
card.setCardNumber("1234 5678 8765 4321");
card.setExpiryDate("10/2024");
card.setCVC("123");
Result result = paymentsService.processPayment(card,
    3.99, Currency.GBP);
if (result.isSuccess()) {
    fulfilOrder();
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Implementation of this function is on another node!

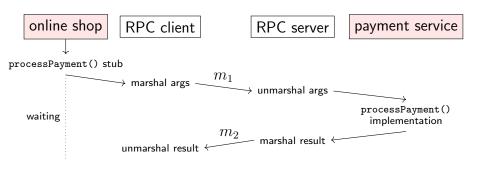


RPC server

payment service

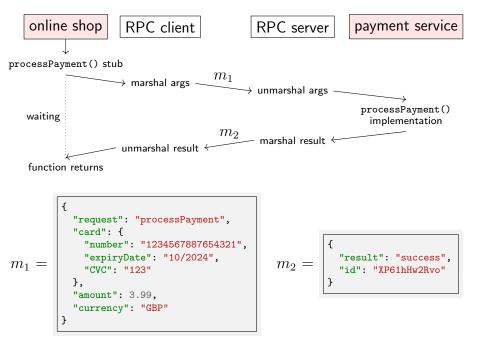
```
m_1 = \begin{cases} \{ & \text{"request": "processPayment",} \\ & \text{"card": } \{ \\ & \text{"number": "1234567887654321",} \\ & \text{"expiryDate": "10/2024",} \\ & \text{"CVC": "123"} \\ \}, \\ & \text{"amount": 3.99,} \\ & \text{"currency": "GBP"} \\ \} \end{cases}
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```

```
m_2 = \begin{bmatrix} \text{ "result": "success", } \\ \text{"id": "XP61hHw2Rvo"} \\ \text{ } \end{bmatrix}
```



Remote Procedure Call (RPC)

Ideally, RPC makes a call to a remote function look the same as a local function call.

"Location transparency":

system hides where a resource is located.

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In practice...

- what if the service crashes during the function call?
- what if a message is lost?
- what if a message is delayed?
- if something goes wrong, is it safe to retry?

RPC history

- ► SunRPC/ONC RPC (1980s, basis for NFS)
- ► CORBA: object-oriented middleware, hot in the 1990s
- Microsoft's DCOM and Java RMI (similar to CORBA)
- ► SOAP/XML-RPC: RPC using XML and HTTP (1998)
- ► Thrift (Facebook, 2007)
- ▶ gRPC (Google, 2015)
- ► REST (often with JSON)
- Ajax in web browsers

RPC/REST in JavaScript

```
let args = {amount: 3.99, currency: 'GBP', /*...*/};
let request = {
  method: 'POST',
  body: JSON.stringify(args),
  headers: {'Content-Type': 'application/json'}
};
fetch('https://example.com/payments', request)
  .then((response) => {
    if (response.ok) success(response.json());
    else failure(response.status); // server error
  })
  .catch((error) => {
    failure(error); // network error
  }):
```

RPC in enterprise systems

"Service-oriented architecture" (SOA) / "microservices":

splitting a large software application into multiple services (on multiple nodes) that communicate via RPC.

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Different services implemented in different languages:

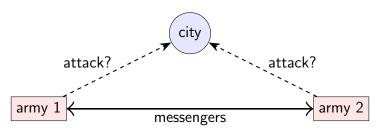
- interoperability: datatype conversions
- ► Interface Definition Language (IDL): language-independent API specification

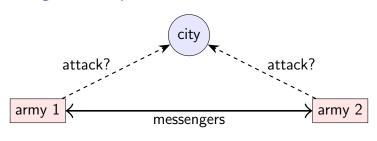
gRPC IDL example

```
message PaymentRequest {
 message Card {
   required string cardNumber = 1;
   optional int32 expiryMonth = 2;
   optional int32 expiryYear = 3;
   optional int32 CVC = 4;
 enum Currency { GBP = 1; USD = 2; }
 required Card card = 1;
 required int64 amount = 2;
 required Currency currency = 3;
message PaymentStatus {
 required bool success = 1;
 optional string errorMessage = 2;
service PaymentService {
 rpc ProcessPayment(PaymentRequest) returns (PaymentStatus) {}
}
```

Lecture 2

Models of distributed systems





army 1	army 2	outcome
does not attack	does not attack	nothing happens
attacks	does not attack	army 1 defeated
does not attack	attacks	army 2 defeated
attacks	attacks	city captured

Desired: army 1 attacks if and only if army 2 attacks







From general 1's point of view, this is indistinguishable from:



How should the generals decide?

- 1. General 1 always attacks, even if no response is received?
 - Send lots of messengers to increase probability that one will get through
 - ▶ If all are captured, general 2 does not know about the attack, so general 1 loses

How should the generals decide?

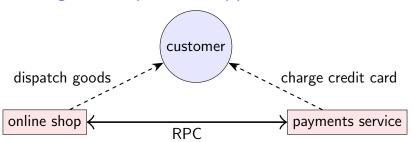
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 - ▶ But general 2 knows that general 1 will only attack if general 2's response gets through
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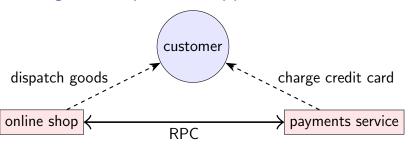
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No common knowledge: the only way of knowing something is to communicate it

The two generals problem applied



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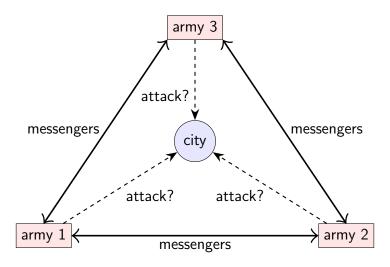


online shop	payments service	outcome
does not dispatch	does not charge	nothing happens
dispatches	does not charge	shop loses money
does not dispatch	charges	customer complaint
dispatches	charges	everyone happy

Desired: online shop dispatches if and only if payment made



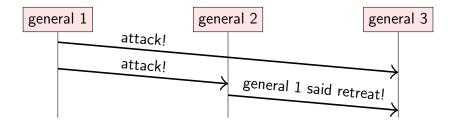
The Byzantine generals problem



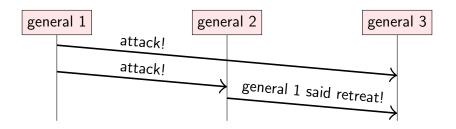
Problem: some of the generals might be traitors



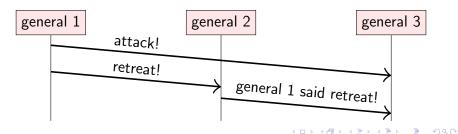
Generals who might lie



Generals who might lie



From general 3's point of view, this is indistinguishable from:



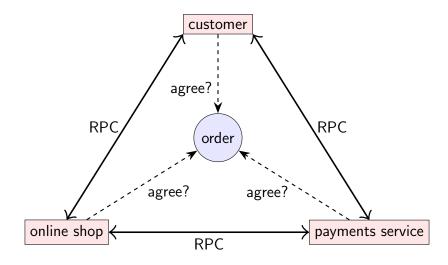
The Byzantine generals problem

- Each general is either *malicious* or *honest*
- ► Up to f generals might be malicious
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- ► Nevertheless, honest generals must agree on plan
- ▶ Theorem: need 3f + 1 generals in total to tolerate f malicious generals (i.e. $<\frac{1}{3}$ may be malicious)
- Cryptography (digital signatures) helps but problem remains hard

Trust relationships and malicious behaviour



Who can trust whom?

The Byzantine empire (650 CE)

Byzantium/Constantinople/Istanbul



 $Source: \ https://commons.wikimedia.org/wiki/File: Byzantium by 650 AD.svg$

"Byzantine" has long been used for "excessively complicated, bureaucratic, devious" (e.g. "the Byzantine tax law")



System models

We have seen two thought experiments:

- ► Two generals problem: a model of networks
- ► Byzantine generals problem: a model of node behaviour In real systems, both nodes and networks may be faulty!

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Capture assumptions in a system model consisting of:

- ► Network behaviour (e.g. message loss)
- ► Node behaviour (e.g. crashes)
- ▶ Timing behaviour (e.g. latency)

Choice of models for each of these parts.

Networks are unreliable





In the sea, sharks bite fibre optic cables

https://slate.com/technology/2014/08/

shark-attacks-threaten-google-s-undersea-internet-cables-video.html

On land, cows step on the cables

https://twitter.com/uhoelzle/status/1263333283107991558



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 A malicious adversary may interfere with messages
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 Data stored on disk survives the crash.
- Byzantine (fail-arbitrary):
 A node is faulty if it deviates from the algorithm.
 Faulty nodes may do anything, including crashing or malicious behaviour.

A node that is not faulty is called "correct"



System model: synchrony (timing) assumptions

Assume one of the following for network and nodes:

▶ Synchronous:

Message latency no greater than a known upper bound. Nodes execute algorithm at a known speed.

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► Partially synchronous:

The system is asynchronous for some finite (but unknown) periods of time, synchronous otherwise.

► Asynchronous:

Messages can be delayed arbitrarily. Nodes can pause execution arbitrarily. No timing guarantees at all.

Note: other parts of computer science use the terms "synchronous" and "asynchronous" differently.

Violations of synchrony in practice

Networks usually have quite predictable latency, which can occasionally increase:

- Message loss requiring retry
- Congestion/contention causing queueing
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Nodes usually execute code at a predictable speed, with occasional pauses:

- Operating system scheduling issues, e.g. priority inversion
- Stop-the-world garbage collection pauses
- Page faults, swap, thrashing

Real-time operating systems (RTOS) provide scheduling guarantees, but most distributed systems do not use RTOS

System models summary

For each of the three parts, pick one:

- Network: reliable, fair-loss, or arbitrary
- Nodes: crash-stop, crash-recovery, or Byzantine
- ➤ **Timing:** synchronous, partially synchronous, or asynchronous

This is the basis for any distributed algorithm. If your assumptions are wrong, all bets are off!

Availability

Online shop wants to sell stuff 24/7!Service unavailability = downtime = losing money

Availability = uptime = fraction of time that a service is functioning correctly

- ► "Two nines" = 99% up = down 3.7 days/year
- "Three nines" = 99.9% up = down 8.8 hours/year
- ► "Four nines" = 99.99% up = down 53 minutes/year
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Service-Level Objective (SLO):

e.g. "99.9% of requests in a day get a response in 200 ms"

Service-Level Agreement (SLA): contract specifying some SLO, penalties for violation



Achieving high availability: fault tolerance

Failure: system as a whole isn't working

Fault: some part of the system isn't working

- Node fault: crash (crash-stop/crash-recovery), deviating from algorithm (Byzantine)
- Network fault: dropping or significantly delaying messages

Fault tolerance:

system as a whole continues working, despite faults (up to some maximum number of faults)

Single point of failure (SPOF): node/network link whose fault leads to failure

Failure detectors

Failure detector:

algorithm that detects whether another node is faulty

Perfect failure detector:

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Problem:

cannot tell the difference between crashed node, temporarily unresponsive node, lost message, and delayed message

Failure detection in partially synchronous systems

Perfect timeout-based failure detector exists only in a synchronous crash-stop system with reliable links.

Eventually perfect failure detector:

- May temporarily label a node as crashed, even though it is correct
- May temporarily label a node as correct, even though it has crashed
- But eventually, labels a node as crashed if and only if it has crashed

Reflects fact that detection is not instantaneous, and we may have spurious timeouts

Lecture 3

Time, clocks, and ordering of events

A detective story

In the night from 30 June to 1 July 2012 (UK time), many online services and systems around the world crashed simultaneously.

Servers locked up and stopped responding.

Some airlines could not process any reservations or check-ins for several hours.

What happened?

Distributed systems often need to measure time, e.g.:

► Schedulers, timeouts, failure detectors, retry timers

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We distinguish two types of clock:

- physical clocks: count number of seconds elapsed
- ▶ logical clocks: count events, e.g. messages sent

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- Schedulers, timeouts, failure detectors, retry timers
- Performance measurements, statistics, profiling
- ▶ Log files & databases: record when an event occurred
- ▶ Data with time-limited validity (e.g. cache entries)
- ▶ Determining order of events across several nodes

We distinguish two types of clock:

- physical clocks: count number of seconds elapsed
- ▶ logical clocks: count events, e.g. messages sent

NB. Clock in digital electronics (oscillator) ≠ clock in distributed systems (source of **timestamps**)

Quartz clocks

- Quartz crystal laser-trimmed to mechanically resonate at a specific frequency
- ▶ Piezoelectric effect: mechanical force ⇔ electric field
- Oscillator circuit produces signal at resonant frequency
- Count number of cycles to measure elapsed time

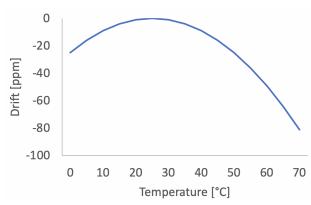




Quartz clock error: drift

- One clock runs slightly fast, another slightly slow
- Drift measured in parts per million (ppm)
- ▶ 1 ppm = 1 microsecond/second = 86 ms/day = 32 s/year
- ▶ Most computer clocks correct within ≈ 50 ppm

Temperature significantly affects drift



Atomic clocks

- ► Caesium-133 has a resonance ("hyperfine transition") at ≈ 9 GHz
- Tune an electronic oscillator to that resonant frequency
- ► 1 second = 9,192,631,770 periods of that signal
- Accuracy ≈ 1 in 10^{-14} (1 second in 3 million years)
- Price $\approx £20,000$ (?) (can get cheaper rubidium clocks for $\approx £1,000$)

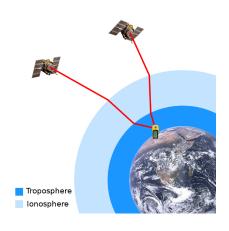


https:

//www.microsemi.com/product-directory/ cesium-frequency-references/ 4115-5071a-cesium-primary-frequency-standard

GPS as time source

- ▶ 31 satellites, each carrying an atomic clock
- satellite broadcasts current time and location
- calculate position from speed-of-light delay between satellite and receiver
- corrections for atmospheric effects, relativity, etc.
- in datacenters, need antenna on the roof



 $https://commons.wikimedia.org/wiki/File: \\ Gps-atmospheric-efects.png$

Greenwich Mean Time (GMT, solar time): it's noon when the sun is in the south, as seen from the Greenwich meridian



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Time zones and daylight savings time are offsets to UTC



Leap seconds

Every year, on 30 June and 31 December at 23:59:59 UTC, one of three things happens:

- ► The clock immediately jumps forward to 00:00:00, skipping one second (**negative leap second**)
- ▶ The clock moves to 00:00:00 after one second, as usual
- ► The clock moves to 23:59:60 after one second, and then moves to 00:00:00 after one further second (positive leap second)

This is announced several months beforehand.



http://leapsecond.com/notes/leap-watch.htm

How computers represent timestamps

Two most common representations:

- ▶ Unix time: number of seconds since 1 January 1970 00:00:00 UTC (the "epoch"), not counting leap seconds
- ► ISO 8601: year, month, day, hour, minute, second, and timezone offset relative to UTC example: 2023-11-02T09:50:17+00:00

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Conversion between the two requires:

- ► Knowledge of past and future leap seconds...?!

How most software deals with leap seconds

By ignoring them!



https://www.flickr.com/ photos/ru_boff/ 37915499055/

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However, OS and DistSys often need timings with sub-second accuracy.



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However, OS and DistSys often need timings with sub-second accuracy.

30 June 2012: bug in Linux kernel caused livelock on leap second, causing many Internet services to go down

Pragmatic solution: "smear" (spread out) the leap second over the course of a day



https://www.flickr.com/ photos/ru_boff/ 37915499055/

Clock synchronisation

Computers track physical time/UTC with a quartz clock (with battery, continues running when power is off)

Due to **clock drift**, clock error gradually increases

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Clock skew: difference between two clocks at a point in time

Clock synchronisation

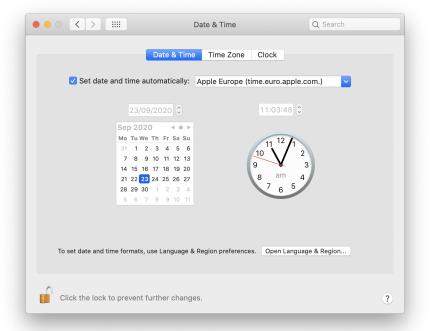
Computers track physical time/UTC with a quartz clock (with battery, continues running when power is off)

Due to **clock drift**, clock error gradually increases

Clock skew: difference between two clocks at a point in time

Solution: Periodically get the current time from a server that has a more accurate time source (atomic clock or GPS receiver)

Protocols: Network Time Protocol (**NTP**), Precision Time Protocol (**PTP**)



Network Time Protocol (NTP)

Many operating system vendors run NTP servers, configure OS to use them by default

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Hierarchy of clock servers arranged into **strata**:

- Stratum 0: atomic clock or GPS receiver
- Stratum 1: synced directly with stratum 0 device
- Stratum 2: servers that sync with stratum 1, etc.

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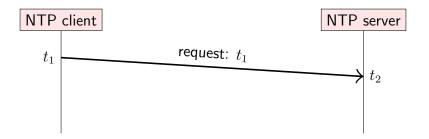
May contact multiple servers, discard outliers, average rest

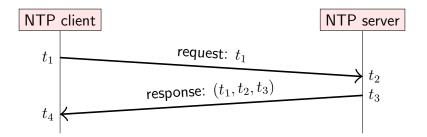
Makes multiple requests to the same server, use statistics to reduce random error due to variations in network latency

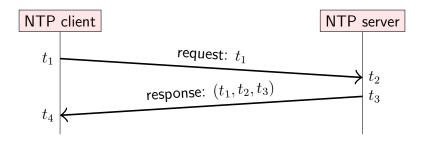
Reduces clock skew to a few milliseconds in good network conditions, but can be much worse!



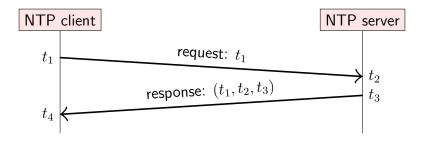






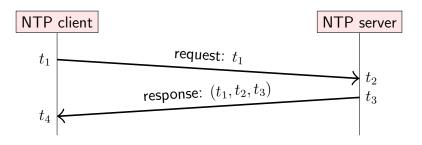


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Estimated clock skew:
$$\theta=t_3+rac{\delta}{2}-t_4=rac{t_2-t_1+t_3-t_4}{2}$$

Correcting clock skew

Once the client has estimated the clock skew θ , it needs to apply that correction to its clock.

▶ If $|\theta| < 125$ ms, **slew** the clock: slightly speed it up or slow it down by up to 500 ppm (brings clocks in sync within ≈ 5 minutes)

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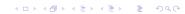
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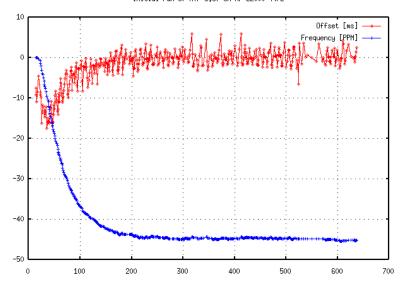
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- ▶ If $|\theta| \ge 1{,}000$ s, **panic** and do nothing (leave the problem for a human operator to resolve)

Systems that rely on clock sync need to monitor clock skew!





http://www.ntp.org/ntpfaq/NTP-s-algo.htm

```
// BAD:
long startTime = System.currentTimeMillis();
doSomething();
long endTime = System.currentTimeMillis();
long elapsedMillis = endTime - startTime;
// elapsedMillis may be negative!
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long startTime = System.nanoTime();
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long endTime = System.nanoTime();
long elapsedNanos = endTime - startTime;
// elapsedNanos is always >= 0
```

Time-of-day clock:

▶ Time since a fixed date (e.g. 1 January 1970 epoch)

Monotonic clock:

► Time since arbitrary point (e.g. when machine booted up)

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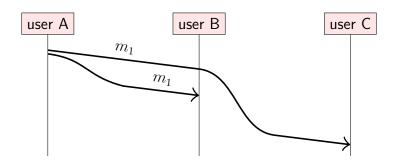
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- ▶ Java: System.currentTimeMillis()
- Linux: clock_gettime(CLOCK_REALTIME)

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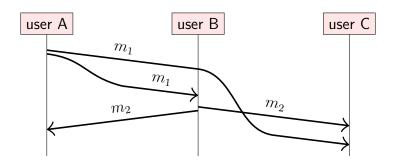


Ordering of messages



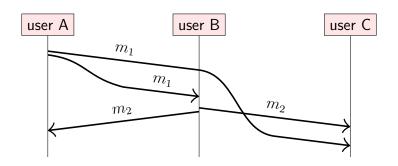
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Ordering of messages



 $m_1 =$ "A says: The moon is made of cheese!" $m_2 =$ "B says: Oh no it isn't!"

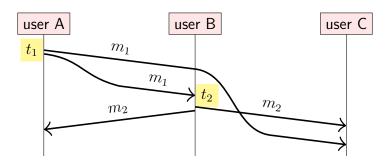
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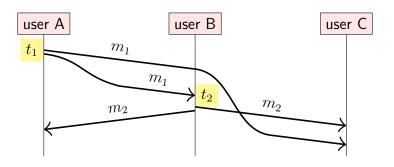
C sees m_2 first, m_1 second, even though logically m_1 happened before m_2 .

Ordering of messages using timestamps?



 $m_1=(t_1,$ "A says: The moon is made of cheese!") $m_2=(t_2,$ "B says: Oh no it isn't!")

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Problem: even with synced clocks, $t_2 < t_1$ is possible. Timestamp order is inconsistent with expected order!



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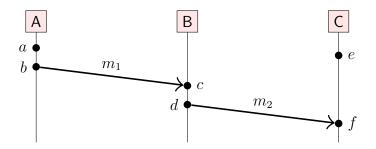
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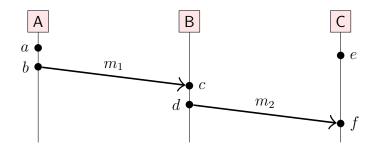
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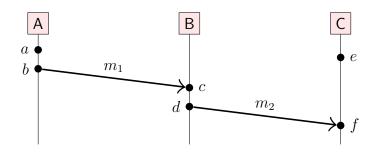
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The happens-before relation is a partial order: it is possible that neither $a \to b$ nor $b \to a$. In that case, a and b are **concurrent** (written $a \parallel b$).

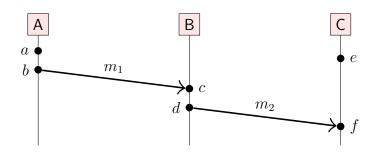




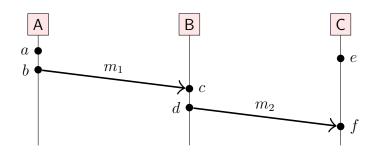
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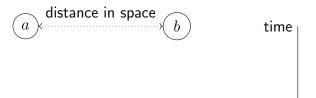


Taken from physics (relativity).

- ▶ When $a \rightarrow b$, then a might have caused b.
- ▶ When $a \parallel b$, we know that a cannot have caused b.

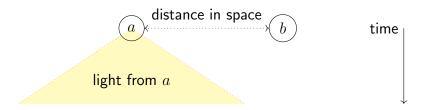
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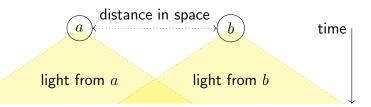
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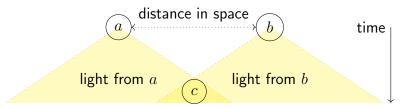
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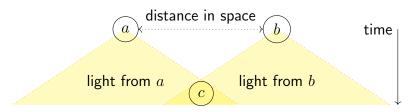
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Happens-before relation encodes potential causality.



Let \prec be a strict total order on events.

If $(a \rightarrow b) \Longrightarrow (a \prec b)$ then \prec is a causal order

(or: \prec is "consistent with causality").

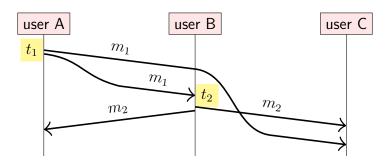
NB. "causal" \neq "casual"!



Lecture 4

Broadcast protocols and logical time

Physical timestamps inconsistent with causality



$$m_1=(t_1,$$
 "A says: The moon is made of cheese!") $m_2=(t_2,$ "B says: Oh no it isn't!")

Problem: even with synced clocks, $t_2 < t_1$ is possible. Timestamp order is inconsistent with expected order!



Logical vs. physical clocks

- Physical clock: count number of seconds elapsed
- ► Logical clock: count number of **events occurred**

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$$(e_1 \rightarrow e_2) \Longrightarrow (T(e_1) < T(e_2))$$

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We will look at two types of logical clocks:

- Lamport clocks
- Vector clocks

Lamport clocks algorithm

on initialisation do

$$t := 0$$

 \triangleright each node has its own local variable t

end on

on any event occurring at the local node do

$$t := t + 1$$

end on

on request to send message $m\ \mathbf{do}$

t:=t+1; send (t,m) via the underlying network link

end on

on receiving (t', m) via the underlying network link **do**

$$t := \max(t, t') + 1$$

deliver m to the application

end on



- ► Each node maintains a counter t, incremented on every local event e
- Let L(e) be the value of t after that increment
- Attach current t to messages sent over network
- Recipient moves its clock forward to timestamp in the message (if greater than local counter), then increments

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Properties of this scheme:

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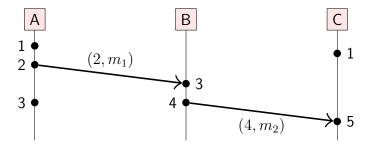
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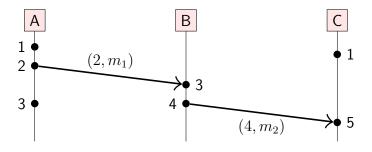
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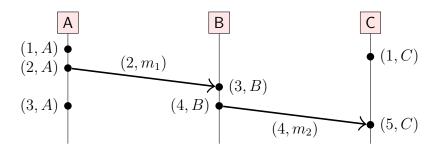
Properties of this scheme:

- ▶ If $a \to b$ then L(a) < L(b)
- ▶ However, L(a) < L(b) does not imply $a \to b$
- ▶ Possible that L(a) = L(b) for $a \neq b$

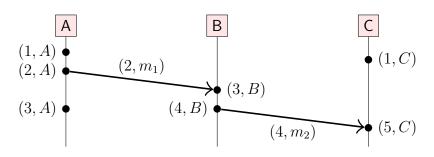




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Define a **total order** ≺ using Lamport timestamps:

$$(a \prec b) \Longleftrightarrow (L(a) < L(b) \lor (L(a) = L(b) \land N(a) < N(b)))$$

This order is **causal**: $(a \to b) \Longrightarrow (a \prec b)$

Given Lamport timestamps L(a) and L(b) with L(a) < L(b) we can't tell whether $a \to b$ or $a \parallel b$.

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If we want to detect which events are concurrent, we need **vector clocks**:

▶ Assume n nodes in the system, $N = \langle N_1, N_2, \dots, N_n \rangle$

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- ▶ Assume n nodes in the system, $N = \langle N_1, N_2, \dots, N_n \rangle$
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- ► Attach current vector timestamp to each message
- Recipient merges message vector into its local vector

Vector clocks algorithm

on initialisation at node N_i do $T:=\langle 0,0,\dots,0\rangle \qquad \qquad \rhd \text{ local variable at node } N_i$ end on

on any event occurring at node N_i do

$$T[i] := T[i] + 1$$

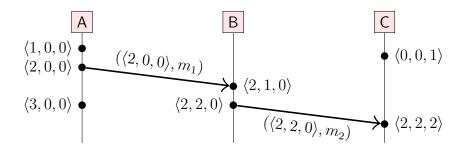
end on

on request to send message m at node N_i do T[i] := T[i] + 1; send (T, m) via network end on

on receiving (T',m) at node N_i via the network **do** $T[j] := \max(T[j], T'[j])$ for every $j \in \{1, \dots, n\}$ T[i] := T[i] + 1; deliver m to the application end on

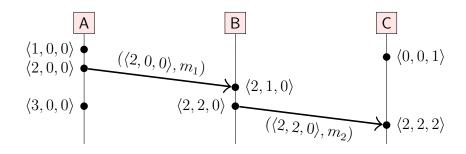
Vector clocks example

Assuming the vector of nodes is $N = \langle A, B, C \rangle$:



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The vector timestamp of an event e represents a set of events, e and its causal dependencies: $\{e\} \cup \{a \mid a \rightarrow e\}$

For example, $\langle 2, 2, 0 \rangle$ represents the first two events from A, the first two events from B, and no events from C.



Vector clocks ordering

Define the following order on vector timestamps (in a system with n nodes):

- $ightharpoonup T = T' ext{ iff } T[i] = T'[i] ext{ for all } i \in \{1, \dots, n\}$
- $T \leq T' \text{ iff } T[i] \leq T'[i] \text{ for all } i \in \{1, \dots, n\}$
- ▶ T < T' iff $T \le T'$ and $T \ne T'$
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$$V(a) \le V(b) \text{ iff } (\{a\} \cup \{e \mid e \to a\}) \subseteq (\{b\} \cup \{e \mid e \to b\})$$

Properties of this order:

- \blacktriangleright $(V(a) < V(b)) \iff (a \to b)$
- $ightharpoonup (V(a) = V(b)) \iff (a = b)$
- \blacktriangleright $(V(a) \parallel V(b)) \iff (a \parallel b)$

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Build upon system models from lecture 2:

 Can be best-effort (may drop messages) or reliable (non-faulty nodes deliver every message, by retransmitting dropped messages)

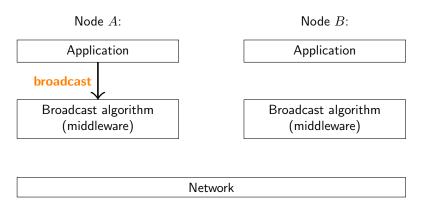
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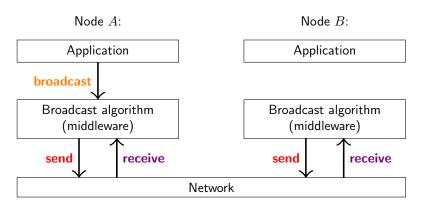
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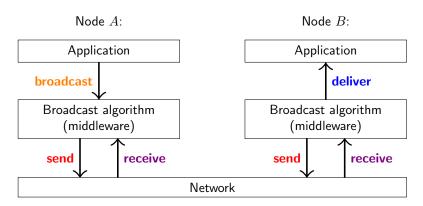
- Can be best-effort (may drop messages) or reliable (non-faulty nodes deliver every message, by retransmitting dropped messages)
- ▶ Asynchronous/partially synchronous timing model
 ⇒ no upper bound on message latency

Node A: Node B: Application Application Broadcast algorithm Broadcast algorithm (middleware) (middleware) Network





Assume network provides point-to-point send/receive



Assume network provides point-to-point send/receive

After broadcast algorithm **receives** message from network, it may buffer/queue it before **delivering** to the application

FIFO broadcast:

If m_1 and m_2 are broadcast by the same node, and broadcast $(m_1) \to \text{broadcast}(m_2)$, then m_1 must be delivered before m_2

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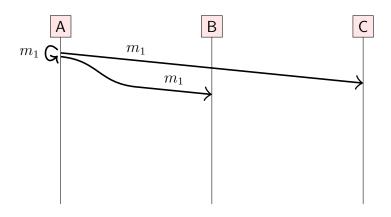
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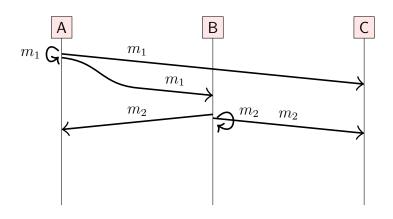
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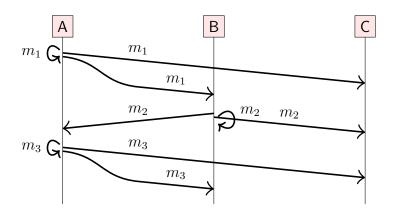
FIFO-total order broadcast:

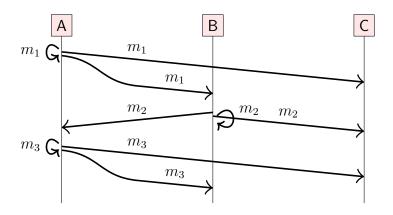
Combination of FIFO broadcast and total order broadcast





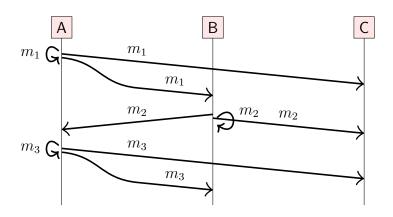






Messages sent by the same node must be delivered in the order they were sent.

Messages sent by different nodes can be delivered in any order.

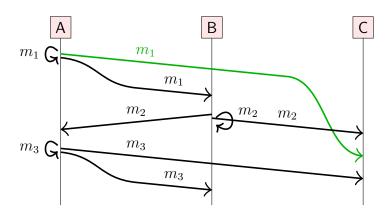


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Valid orders: (m_2, m_1, m_3) or (m_1, m_2, m_3) or (m_1, m_3, m_2)



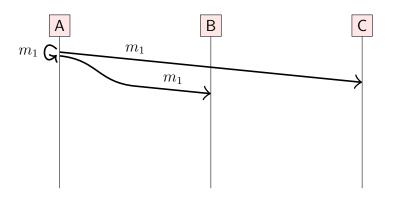


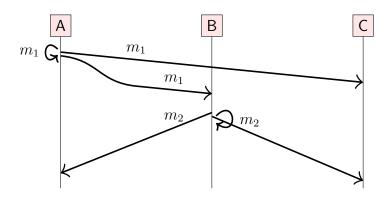
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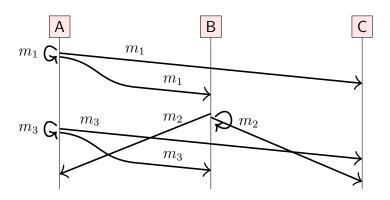
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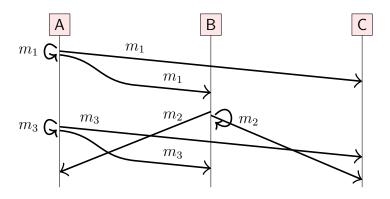
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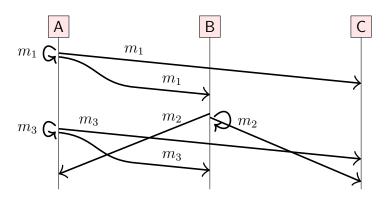








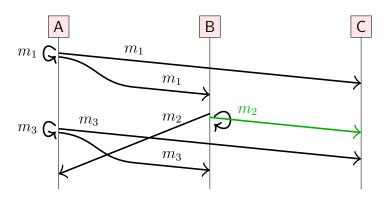
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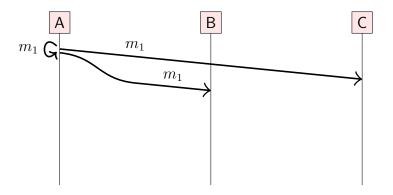
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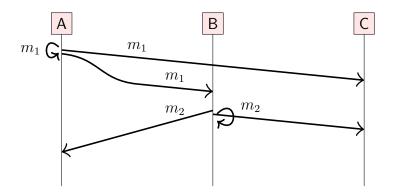
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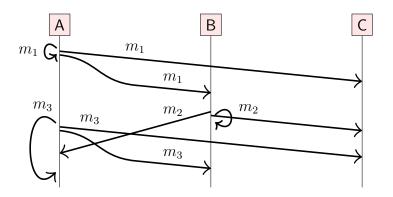


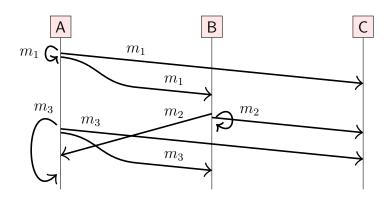
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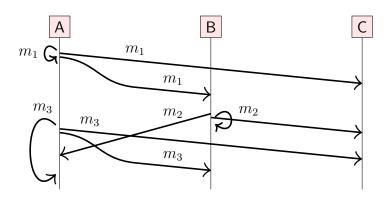








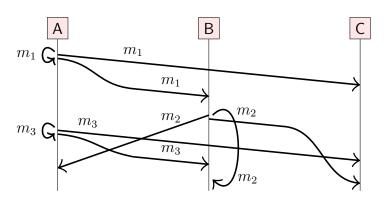
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This includes a node's deliveries to itself!



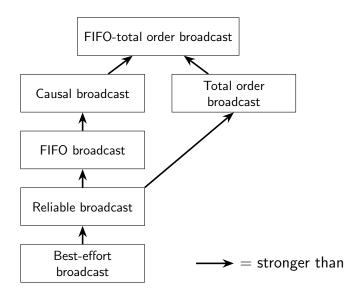


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Relationships between broadcast models



Broadcast algorithms

Break down into two layers:

- 1. Make best-effort broadcast reliable by retransmitting dropped messages
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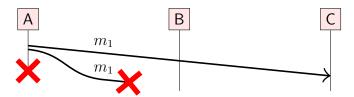
Broadcast algorithms

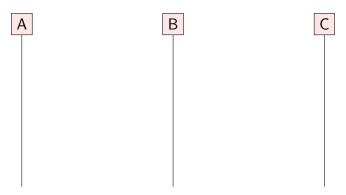
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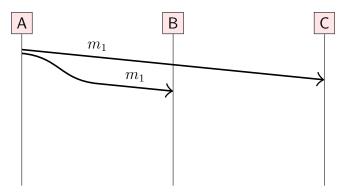
- 1. Make best-effort broadcast reliable by retransmitting dropped messages
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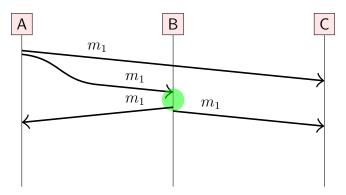
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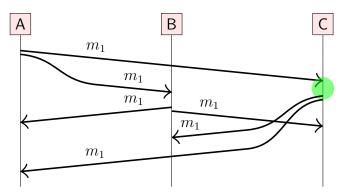
- Use reliable links (retry + deduplicate)
- Problem: node may crash before all messages delivered



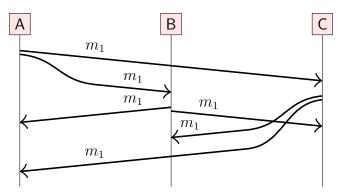




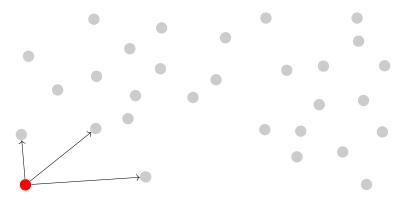


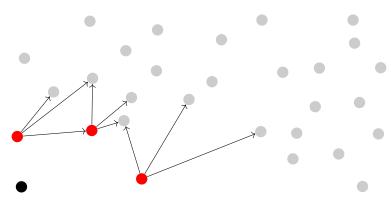


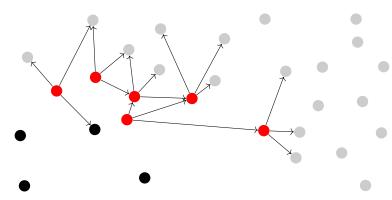
Idea: the **first time** a node receives a particular message, it **re-broadcasts** to each other node (via reliable links).

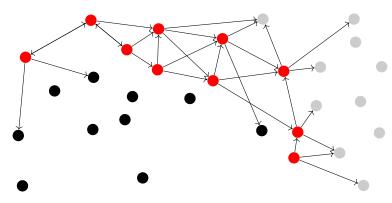


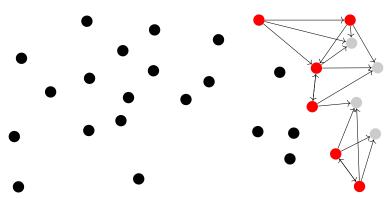
Reliable, but... up to $O(n^2)$ messages for n nodes!

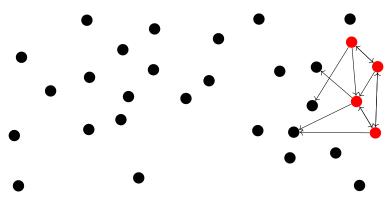




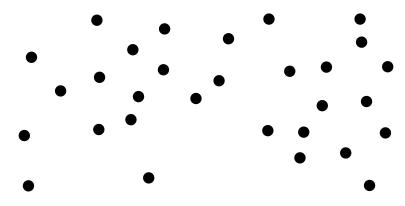








Useful when broadcasting to a large number of nodes. Idea: when a node receives a message for the first time, forward it to 3 other nodes, chosen randomly.



Eventually reaches all nodes (with high probability).



FIFO broadcast algorithm

```
on initialisation do
    sendSeg := 0; delivered := \langle 0, 0, \dots, 0 \rangle; buffer := \{\}
end on
on request to broadcast m at node N_i do
   send (i, sendSeq, m) via reliable broadcast
    sendSeq := sendSeq + 1
end on
on receiving msq from reliable broadcast at node N_i do
    buffer := buffer \cup \{msq\}
   while \exists sender, m. (sender, delivered[sender], m) \in buffer do
       deliver m to the application
       delivered[sender] := delivered[sender] + 1
   end while
end on
```

Causal broadcast algorithm

```
on initialisation do
    sendSeq := 0; delivered := \langle 0, 0, \dots, 0 \rangle; buffer := \{\}
end on
on request to broadcast m at node N_i do
    deps := delivered; deps[i] := sendSeq
   send (i, deps, m) via reliable broadcast
    sendSeq := sendSeq + 1
end on
on receiving msq from reliable broadcast at node N_i do
    buffer := buffer \cup \{msq\}
   while \exists (sender, deps, m) \in buffer. deps < delivered do
       deliver m to the application
       buffer := buffer \setminus \{(sender, deps, m)\}
       delivered[sender] := delivered[sender] + 1
    end while
end on
```

Vector clocks ordering

Define the following order on vector timestamps (in a system with n nodes):

- ightharpoonup T=T' iff T[i]=T'[i] for all $i\in\{1,\ldots,n\}$
- ▶ $T \leq T'$ iff $T[i] \leq T'[i]$ for all $i \in \{1, ..., n\}$
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- Deliver messages in total order of timestamps

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Lamport clocks approach:

- Attach Lamport timestamp to every message
- Deliver messages in total order of timestamps
- ▶ Problem: how do you know if you have seen all messages with timestamp < T? Need to use FIFO links and wait for message with timestamp $\ge T$ from *every* node

Lecture 5

- Keeping a copy of the same data on multiple nodes
- ▶ Databases, filesystems, caches, . . .
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Compare to **RAID** (Redundant Array of Independent Disks): replication within a single computer

- RAID has single controller; in distributed system, each node acts independently
- Replicas can be distributed around the world, near users



User A: The moon is not actually made of cheese!

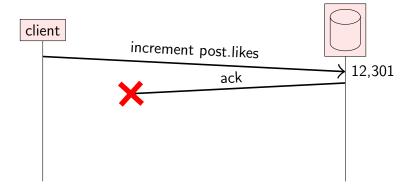
Like 12,300 people like this.

client



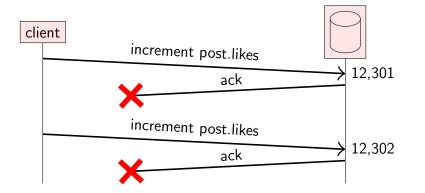
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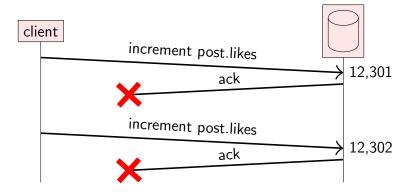
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Deduplicating requests requires that the database tracks which requests it has already seen (in stable storage)







. . .

Добро пожаловать отсюда

Default City



Лепра @leprasorium · 2h Викторианские советы Часть 2 pic.twitter.com/21PraRYBaO

Details



Лепра @leprasorium · 2h Викторианские советы Часть 1 pic.twitter.com/BVE6ao8711

Details

Go to full profile

Idempotence

A function f is idempotent if f(x) = f(f(x)).

- ▶ Not idempotent: f(likeCount) = likeCount + 1
- ▶ **Idempotent:** $f(likeSet) = likeSet \cup \{userID\}$

Idempotent requests can be retried without deduplication.

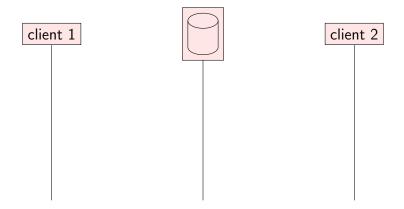
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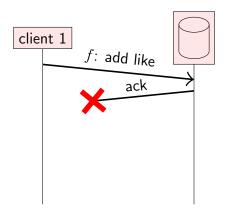
- A function f is idempotent if f(x) = f(f(x)).
 - ▶ Not idempotent: f(likeCount) = likeCount + 1
 - ▶ **Idempotent:** $f(likeSet) = likeSet \cup \{userID\}$

Idempotent requests can be retried without deduplication.

Choice of retry behaviour:

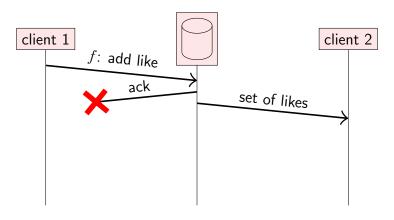
- ► At-most-once semantics: send request, don't retry, update may not happen
- ► At-least-once semantics: retry request until acknowledged, may repeat update
- Exactly-once semantics: retry + idempotence or deduplication



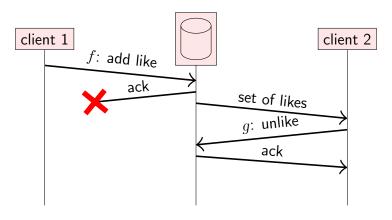


client 2

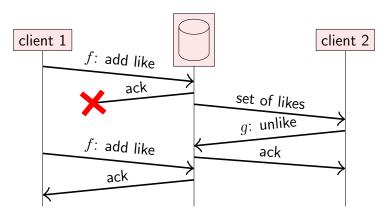
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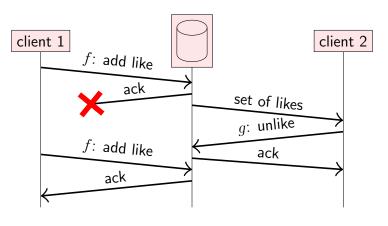


$$f(likes) = likes \cup \{userID\}$$
$$g(likes) = likes \setminus \{userID\}$$

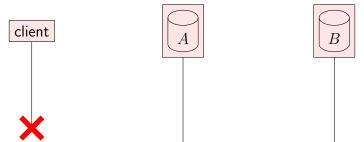


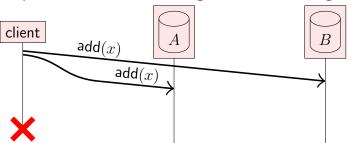
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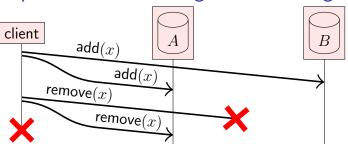
Adding and then removing again

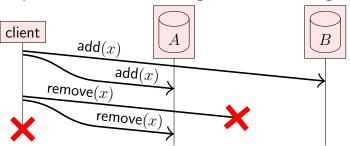


```
\begin{array}{l} f(likes) = likes \cup \{userID\} \\ g(likes) = likes \setminus \{userID\} \\ \textbf{Idempotent?} \ f(f(x)) = f(x) \ \text{but} \ f(g(f(x)) \neq g(f(x)) \end{array}
```

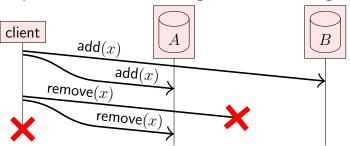




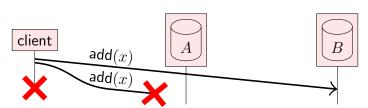


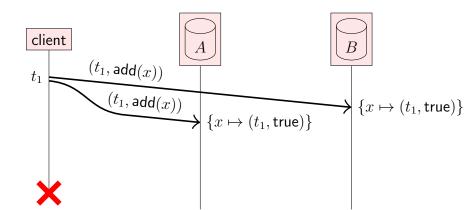


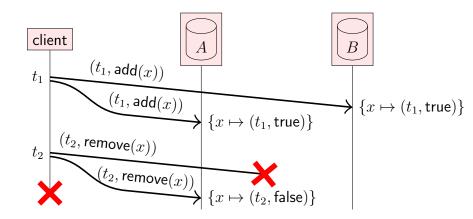
Final state $(x \notin A, x \in B)$ is the same as in this case:

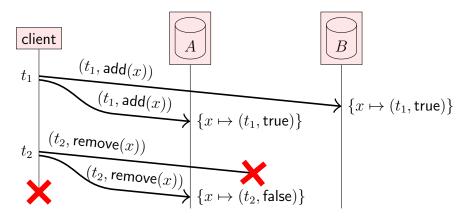


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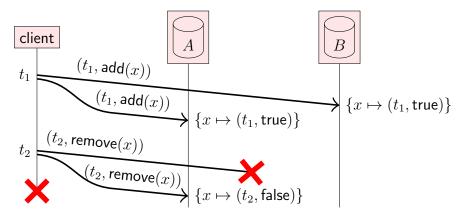








"remove(x)" doesn't actually remove x: it labels x with "false" to indicate it is invisible (a **tombstone**)



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Every record has logical timestamp of last write



Replicas periodically communicate among themselves to check for any inconsistencies.

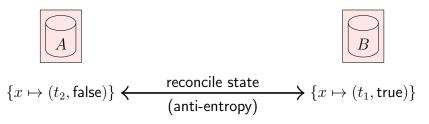


$$\{x \mapsto (t_2, \mathsf{false})\}$$

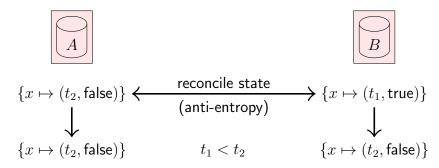


$$\{x \mapsto (t_1, \mathsf{true})\}$$

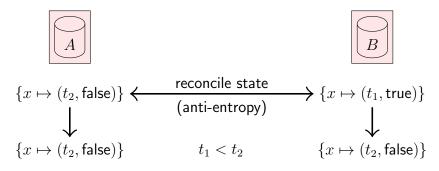
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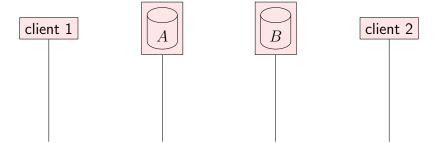
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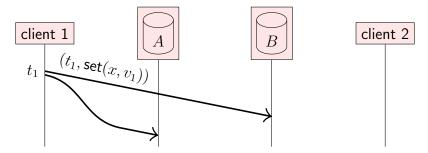


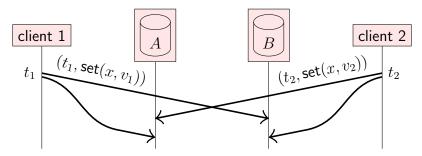
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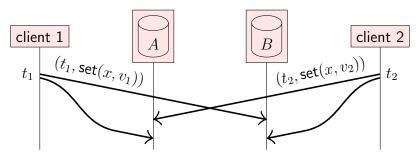


Propagate the record with the latest timestamp, discard the records with earlier timestamps (for a given key).



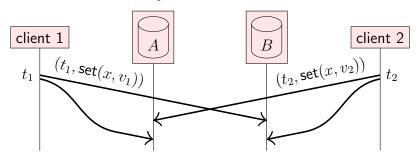






Two common approaches:

Last writer wins (LWW):
Use timestamps with total order (e.g. Lamport clock)
Keep v_2 and discard v_1 if $t_2 > t_1$. Note: **data loss**!



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- Multi-value register:
 Use timestamps with partial order (e.g. vector clock) v_2 replaces v_1 if $t_2 > t_1$; preserve both $\{v_1, v_2\}$ if $t_1 \parallel t_2$

A replica may be **unavailable** due to network partition or node fault (e.g. crash, hardware problem).

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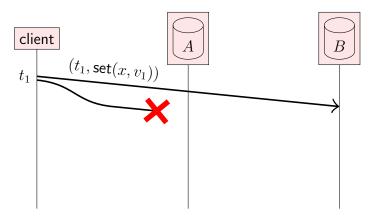
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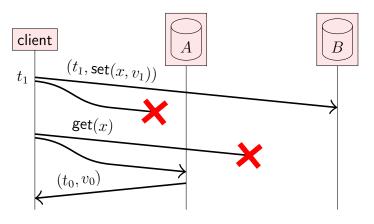
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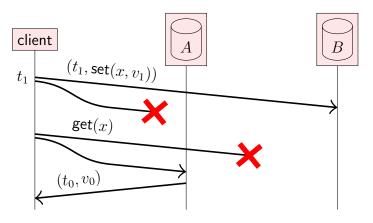
Example with p = 0.01:

replicas n	$P(\geq 1 \text{ faulty})$	$P(\geq \frac{n+1}{2} \text{ faulty})$	$P(all\ n\ faulty)$
1	0.01	0.01	0.01
3	0.03	$3 \cdot 10^{-4}$	10^{-6}
5	0.049	$1 \cdot 10^{-5}$	10^{-10}
100	0.63	$6 \cdot 10^{-74}$	10^{-200}

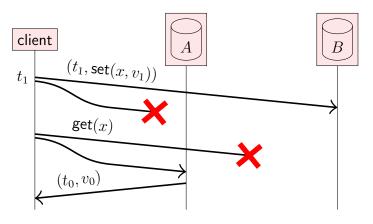






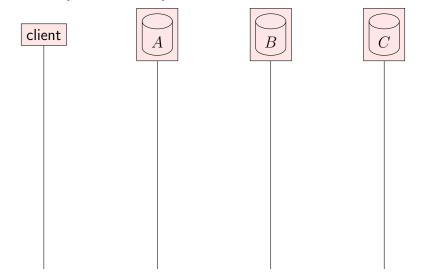


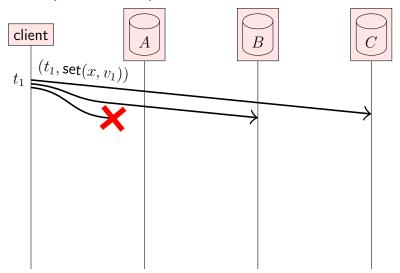
Writing to one replica, reading from another: client does not read back the value it has written

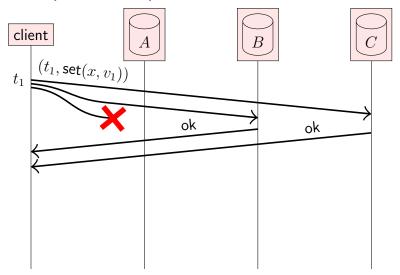


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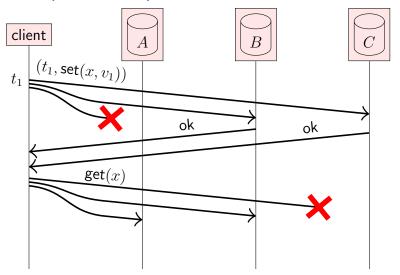
Require writing to/reading from both replicas \Longrightarrow cannot write/read if one replica is unavailable



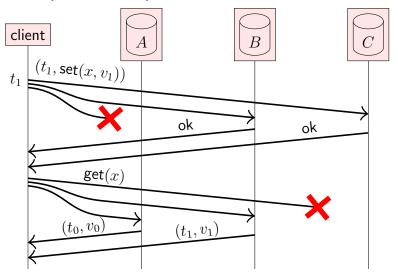




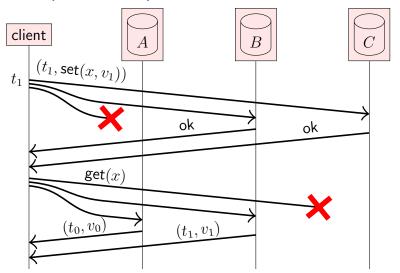
Write succeeds on B and C



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Write succeeds on B and C; read succeeds on A and B



Write succeeds on B and C; read succeeds on A and B Choose between (t_0,v_0) and (t_1,v_1) based on timestamp

Read and write quorums

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Read and write quorums

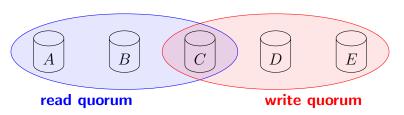
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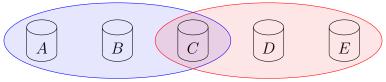
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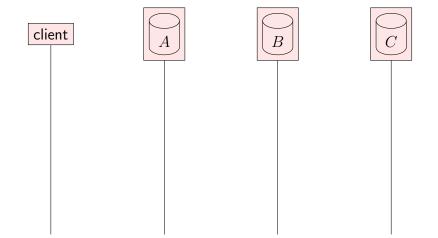
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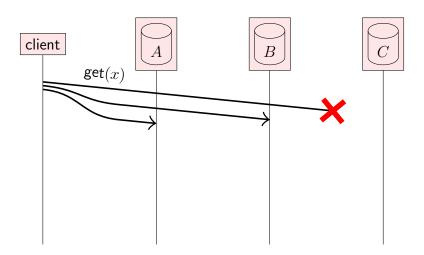
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- lacktriangle Read quorum and write quorum share ≥ 1 replica
- ▶ Typical: $r = w = \frac{n+1}{2}$ for n = 3, 5, 7, ... (majority)
- lacktriangle Reads can tolerate n-r unavailable replicas, writes n-w

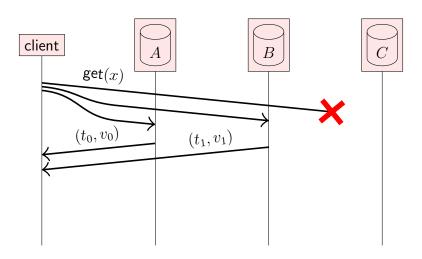


read quorum

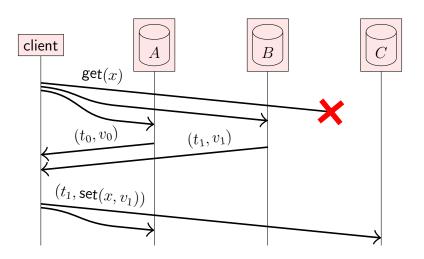
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- ▶ Replica is a **state machine**: starts in fixed initial state, goes through same sequence of state transitions in the same order ⇒ all replicas end up in the same state

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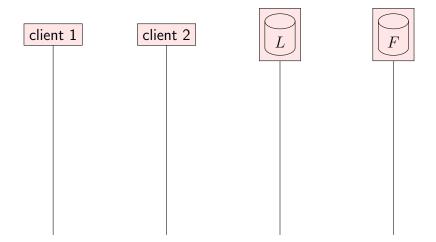
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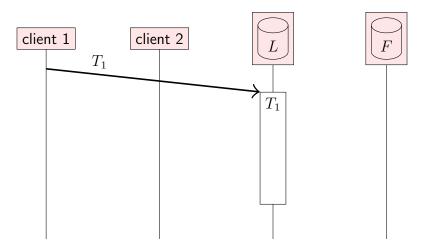
- Cannot update state immediately, have to wait for delivery through broadcast
- ▶ Need fault-tolerant total order broadcast: see lecture 6



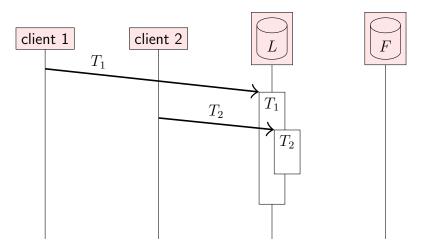
Leader database replica L ensures total order broadcast



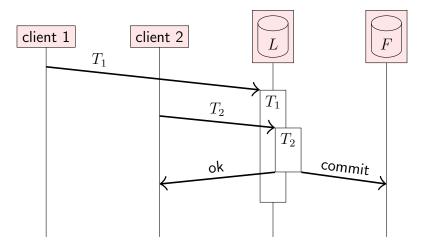
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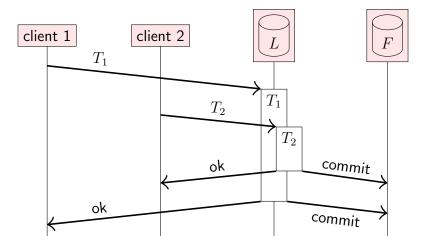
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best-effort	deterministic, commutative, idempotent, tolerates message loss

Lecture 6

Consensus

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- Can we automatically choose a new leader?

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- Paxos: single-value consensus
 Multi-Paxos: generalisation to total order broadcast
- Raft, Viewstamped Replication, Zab: FIFO-total order broadcast by default



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Why not asynchronous?

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There are also consensus algorithms for a partially synchronous **Byzantine** system model (used in blockchains)

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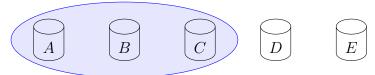


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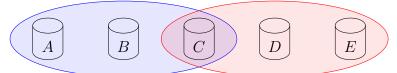
elects a leader

Multi-Paxos, Raft, etc. use a leader to sequence messages.

- ▶ Use a **failure detector** (timeout) to determine suspected crash or unavailability of leader.
- ► On suspected leader crash, **elect a new one**.
- Prevent two leaders at the same time ("split-brain")!

Ensure ≤ 1 leader per **term**:

- Term is incremented every time a leader election is started
- A node can only vote once per term
- ▶ Require a **quorum** of nodes to elect a leader in a term



elects a leader

cannot elect a different leader because C already voted

Can guarantee unique leader **per term**.

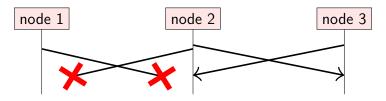
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Example: node 1 is leader in term t, but due to a network partition it can no longer communicate with nodes 2 and 3:

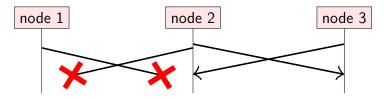


Nodes 2 and 3 may elect a new leader in term t + 1.

Can guarantee unique leader per term.

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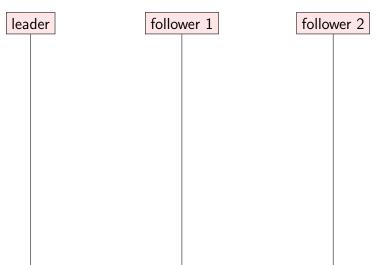
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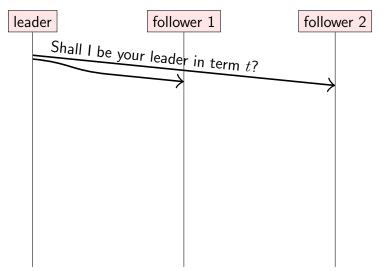


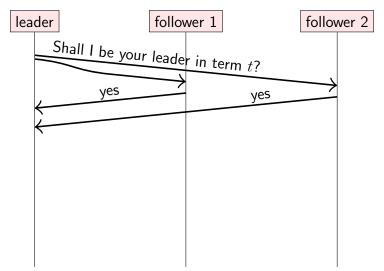
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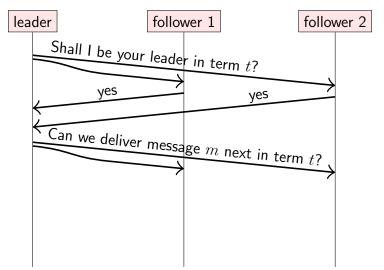
Node 1 may not even know that a new leader has been elected!

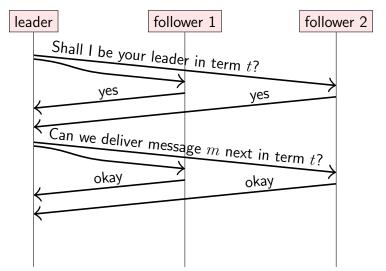


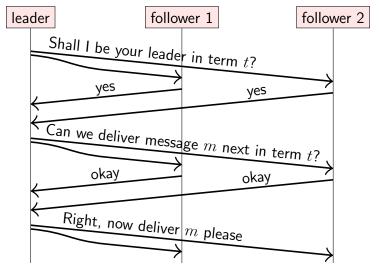


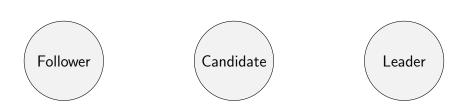


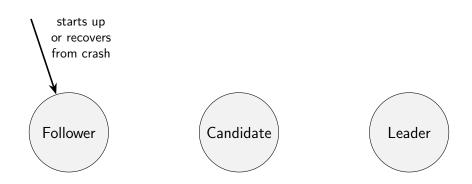


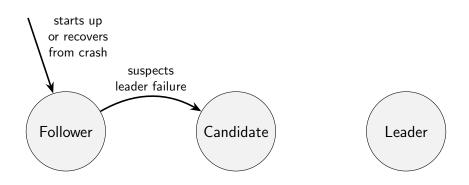


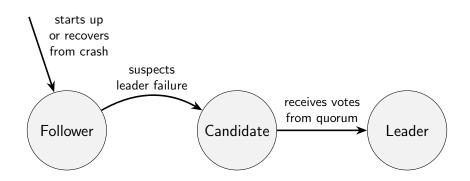


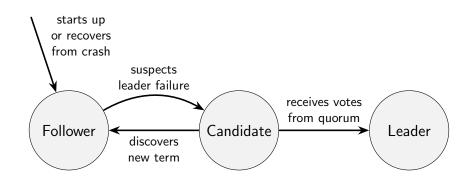


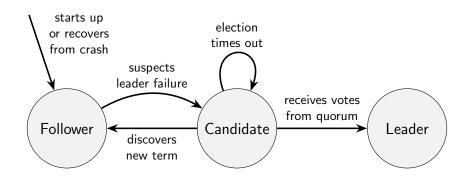


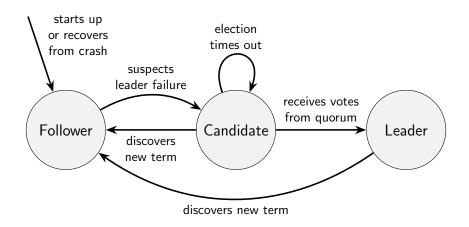












Raft (1/9): initialisation

```
on initialisation do
    currentTerm := 0; votedFor := null
    log := \langle \rangle; commitLength := 0
    currentRole := follower; currentLeader := null
    votesReceived := \{\}; sentLength := \langle \rangle; ackedLength := \langle \rangle
end on
on recovery from crash do
    currentRole := follower; currentLeader := null
    votesReceived := \{\}; sentLength := \langle \rangle; ackedLength := \langle \rangle
end on
on node nodeId suspects leader has failed, or on election timeout do
    currentTerm := currentTerm + 1; currentRole := candidate
    votedFor := nodeId; votesReceived := \{nodeId\}; lastTerm := 0
    if log.length > 0 then lastTerm := log[log.length - 1].term; end if
    msg := (VoteRequest, nodeId, currentTerm, log.length, lastTerm)
    for each node \in nodes: send msg to node
```

start election timer

Raft (1/9): initialisation

```
on initialisation do
```

 $\begin{aligned} currentTerm := 0; \ votedFor := \mathsf{null} \\ log := \langle \rangle; \ commitLength := 0 \end{aligned}$

log[0] log[1] log

currentRole := follower; currentLeader := null

 $votesReceived := \{\}; \ sentLength := \langle \rangle; \ ackedLength := \langle \rangle$

end on

on recovery from crash ${\bf do}$

```
currentRole := follower; \ currentLeader := null \ votesReceived := \{\}; \ sentLength := \langle \rangle; \ ackedLength := \langle \rangle
```

end on

on node nodeId suspects leader has failed, or on election timeout do $currentTerm := currentTerm + 1; \ currentRole := \text{candidate}$ $votedFor := nodeId; \ votesReceived := \{nodeId\}; \ lastTerm := 0$ if log.length > 0 then lastTerm := log[log.length - 1].term; end if msg := (VoteRequest, nodeId, currentTerm, log.length, lastTerm) for each $node \in nodes$: send msg to node start election timer

Raft (2/9): voting on a new leader

```
on receiving (VoteRequest, cId, cTerm, cLogLength, cLogTerm)
        at node nodeId do
   if cTerm > currentTerm then
       currentTerm := cTerm; currentRole := follower
       votedFor := null
   end if
   lastTerm := 0
   if log.length > 0 then lastTerm := log[log.length - 1].term; end if
   logOk := (cLogTerm > lastTerm) \lor
         (cLogTerm = lastTerm \land cLogLength > log.length)
   if cTerm = currentTerm \land logOk \land votedFor \in \{cId, null\} then
       votedFor := cId
       send (VoteResponse, nodeId, currentTerm, true) to node cId
   else
       send (VoteResponse, nodeId, currentTerm, false) to node cId
   end if
end on
```

4D + 4B + 4B + B + 900

Raft (2/9): voting on a new leader

end if end on

c for candidate

```
on receiving (VoteRequest, cId, cTerm, cLogLength, cLogTerm)
        at node nodeId do
   if cTerm > currentTerm then
       currentTerm := cTerm; currentRole := follower
       votedFor := null
   end if
   lastTerm := 0
   if log.length > 0 then lastTerm := log[log.length - 1].term; end if
   logOk := (cLogTerm > lastTerm) \lor
         (cLogTerm = lastTerm \land cLogLength > log.length)
   if cTerm = currentTerm \land logOk \land votedFor \in \{cId, null\} then
       votedFor := cId
       send (VoteResponse, nodeId, currentTerm, true) to node cId
   else
       send (VoteResponse, nodeId, currentTerm, false) to node cId
```

Raft (3/9): collecting votes

```
on receiving (VoteResponse, voterId, term, granted) at nodeId do
   if currentRole = candidate \land term = currentTerm \land granted then
       votesReceived := votesReceived \cup \{voterId\}
       if |votesReceived| > \lceil (|nodes| + 1)/2 \rceil then
           currentRole := leader; currentLeader := nodeId
           cancel election timer
           for each follower \in nodes \setminus \{nodeId\} do
              sentLength[follower] := log.length
              ackedLength[follower] := 0
              ReplicateLog(nodeId, follower)
           end for
       end if
   else if term > currentTerm then
       currentTerm := term
       currentRole := follower
       votedFor := null
       cancel election timer
   end if
end on
                                                4 D > 4 B > 4 B > 4 B > 9 Q P
```

Raft (4/9): broadcasting messages

```
on request to broadcast msg at node nodeId do
   if currentRole = leader then
       append the record (msg: msg, term: currentTerm) to log
       ackedLength[nodeId] := log.length
       for each follower \in nodes \setminus \{nodeId\} do
          ReplicateLog(nodeId, follower)
       end for
   else
       forward the request to currentLeader via a FIFO link
   end if
end on
periodically at node nodeId do
   if currentRole = leader then
       for each follower \in nodes \setminus \{nodeId\} do
          ReplicateLog(nodeId, follower)
       end for
   end if
end do
                                              4 D > 4 B > 4 B > 4 B > 9 Q P
```

Raft (5/9): replicating from leader to followers

Called on the leader whenever there is a new message in the log, and also periodically. If there are no new messages, suffix is the empty list. LogRequest messages with $suffix = \langle \rangle$ serve as heartbeats, letting followers know that the leader is still alive.

Raft (6/9): followers receiving messages

```
on receiving (LogRequest, leaderId, term, prefixLen, prefixTerm,
             leaderCommit, suffix) at node nodeId do
   if term > currentTerm then
       currentTerm := term; \ votedFor := null
       cancel election timer
   end if
   if term = currentTerm then
       currentRole := follower; currentLeader := leaderId
   end if
   logOk := (log.length \ge prefixLen) \land
             (prefixLen = 0 \lor log[prefixLen - 1].term = prefixTerm)
   if term = currentTerm \wedge logOk then
       Appendentries (prefixLen, leaderCommit, suffix)
       ack := prefixLen + suffix.length
       send (LogResponse, nodeId, currentTerm, ack, true) to leaderId
   else
       send (LogResponse, nodeId, currentTerm, 0, false) to leaderId
   end if
end on
```

Raft (7/9): updating followers' logs

```
function APPENDENTRIES(prefixLen, leaderCommit, suffix)
   if suffix.length > 0 \land log.length > prefixLen then
       index := \min(log.length, prefixLen + suffix.length) - 1
       if log[index].term \neq suffix[index - prefixLen].term then
           log := \langle log[0], log[1], \ldots, log[prefixLen-1] \rangle
       end if
   end if
   if prefixLen + suffix.length > log.length then
       for i := log.length - prefixLen to suffix.length - 1 do
          append suffix[i] to log
       end for
   end if
   if leaderCommit > commitLength then
       for i := commitLength to leaderCommit - 1 do
          deliver log[i].msg to the application
       end for
       commitLength := leaderCommit
   end if
end function
```

Raft (8/9): leader receiving log acknowledgements

```
on receiving (LogResponse, follower, term, ack, success) at nodeId do
   if term = currentTerm \land currentRole = leader then
      if success = true \land ack \ge ackedLength[follower] then
          sentLength[follower] := ack
          ackedLength[follower] := ack
          COMMITLOGENTRIES()
      else if sentLength[follower] > 0 then
          sentLength[follower] := sentLength[follower] - 1
          ReplicateLog(nodeId, follower)
      end if
   else if term > currentTerm then
      currentTerm := term
      currentRole := follower
      votedFor := null
      cancel election timer
   end if
end on
```

Raft (9/9): leader committing log entries

Any log entries that have been acknowledged by a quorum of nodes are ready to be committed by the leader. When a log entry is committed, its message is delivered to the application.

```
define acks(length) = |\{n \in nodes \mid ackedLength[n] \geq length\}|
function CommittogEntries
    minAcks := \lceil (\lfloor nodes \rfloor + 1)/2 \rceil
    ready := \{len \in \{1, \dots, log. length\} \mid acks(len) > minAcks\}
   if ready \neq \{\} \land \max(ready) > commitLength \land
           log[max(ready) - 1].term = currentTerm then
       for i := commitLength to max(ready) - 1 do
           deliver log[i].msg to the application
       end for
       commitLength := max(ready)
   end if
end function
```

Lecture 7

Replica consistency

A word that means many different things in different contexts!

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► ACID: a transaction transforms the database from one "consistent" state to another

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Here, "consistent" = satisfying application-specific invariants

e.g. "every course with students enrolled must have at least one lecturer"

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► Read-after-write consistency (lecture 5)

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```
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```

e.g. "every course with students enrolled must have at least one lecturer"

- ► Read-after-write consistency (lecture 5)
- Replication: replica should be "consistent" with other replicas

```
"consistent" = in the same state? (when exactly?)
```

- "consistent" = read operations return same result?
- ► Consistency model: many to choose from



Recall **atomicity** in the context of ACID transactions:

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If the transaction updates data on multiple nodes, this implies:

- Either all nodes must commit, or all must abort
- ▶ If any node crashes, all must abort

Ensuring this is the **atomic commitment** problem. Looks a bit similar to consensus?



Atomic commit versus consensus

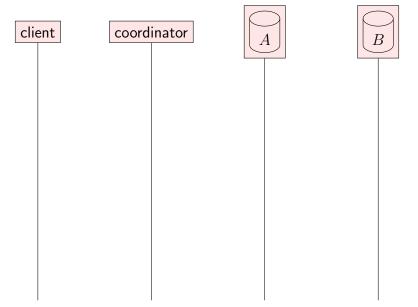
Consensus	Atomic commit
One or more nodes propose a value	Every node votes whether to commit or abort

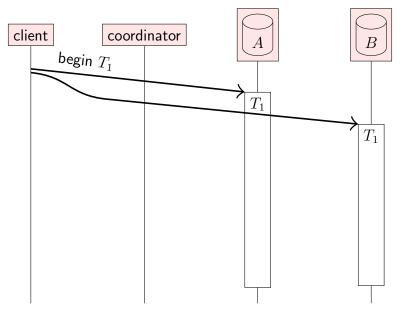
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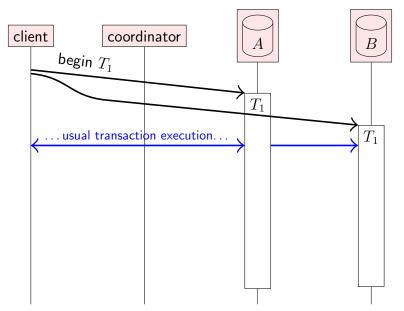
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Any one of the proposed values is decided	Must commit if all nodes vote to commit; must abort if ≥ 1 nodes vote to abort

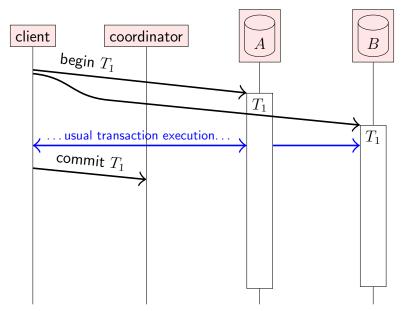
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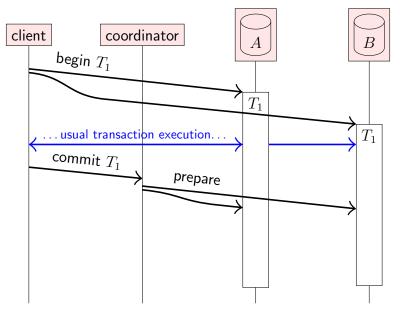
Consensus	Atomic commit
One or more nodes propose a value	Every node votes whether to commit or abort
Any one of the proposed values is decided	Must commit if all nodes vote to commit; must abort if ≥ 1 nodes vote to abort
Crashed nodes can be tolerated, as long as a quorum is working	Must abort if a participating node crashes

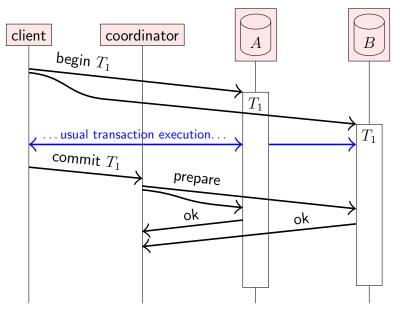


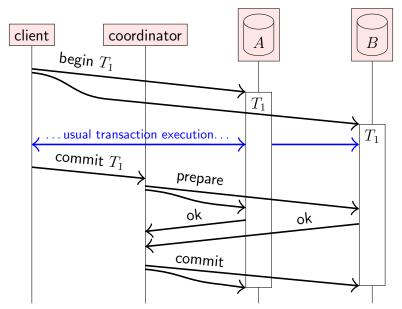


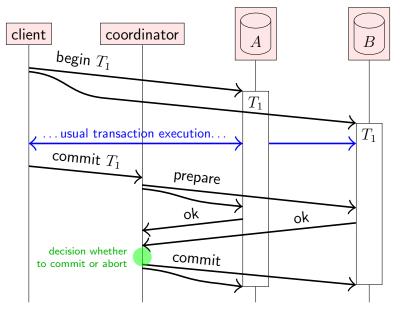












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- Replicas participating in transaction cannot commit or abort after responding "ok" to the *prepare* request (otherwise we risk violating atomicity)
- Algorithm is blocked until coordinator recovers

Fault-tolerant two-phase commit (1/2)

end on

```
on initialisation for transaction T do
   commitVotes[T] := \{\}; replicas[T] := \{\}; decided[T] := false
end on
on request to commit transaction T with participating nodes R do
   for each r \in R do send (Prepare, T, R) to r
end on
on receiving (Prepare, T, R) at node replicaId do
   replicas[T] := R
   ok = "is transaction T able to commit on this replica?"
   total order broadcast (Vote, T, replicaId, ok) to replicas[T]
end on
on a node suspects node replicaId to have crashed do
   for each transaction T in which replicald participated do
       total order broadcast (Vote, T, replicaId, false) to replicas[T]
   end for
```

Fault-tolerant two-phase commit (2/2)

```
on delivering (Vote, T, replicald, ok) by total order broadcast do
   if replicaId \notin commitVotes[T] \land replicaId \in replicas[T] \land
               \neg decided[T] then
       if ok = \text{true then}
           commitVotes[T] := commitVotes[T] \cup \{replicaId\}
           if commitVotes[T] = replicas[T] then
               decided[T] := true
               commit transaction T at this node
           end if
       else
           decided[T] := true
           abort transaction T at this node
       end if
   end if
end on
```

Linearizability

Multiple nodes concurrently accessing replicated data. How do we define "consistency" here?

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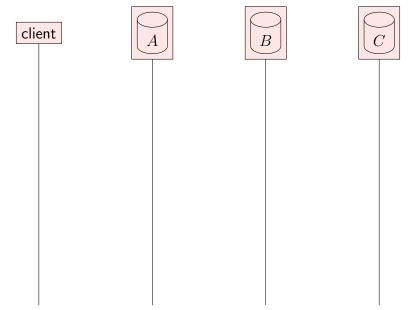
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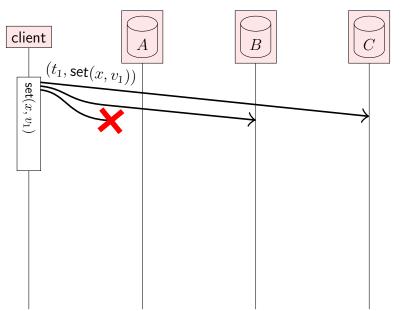
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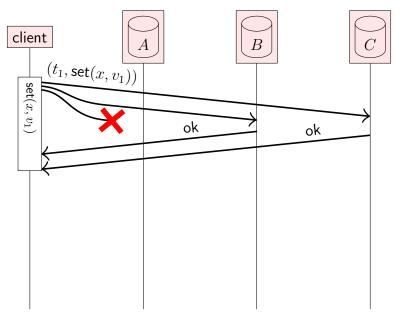
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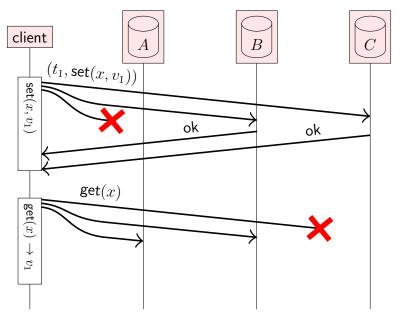
Note: linearizability \neq serializability!

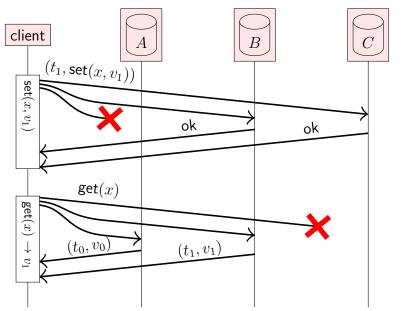


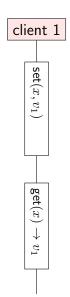




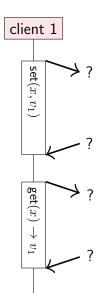




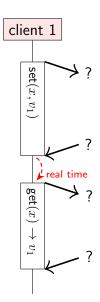




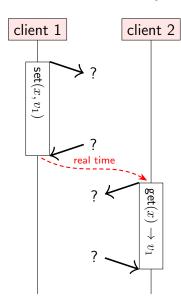
► Focus on client-observable behaviour: when and what an operation returns



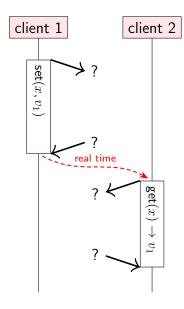
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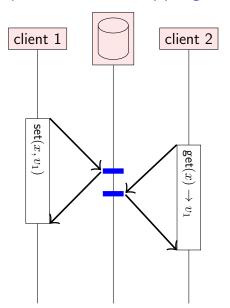
- Focus on client-observable behaviour: when and what an operation returns
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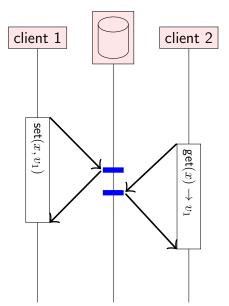
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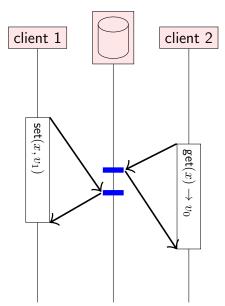
- Focus on client-observable behaviour: when and what an operation returns
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- Even if the operations are on different nodes?
- ➤ This is not happens-before: we want client 2 to read value written by client 1, even if the clients have not communicated!



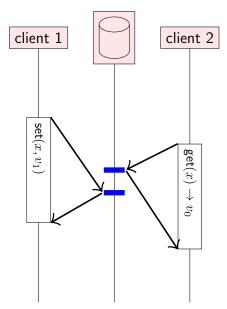
Client 2's get operation overlaps in time with client 1's set operation



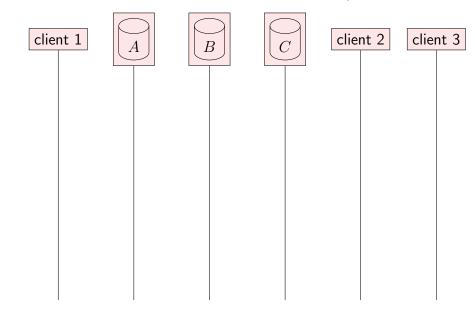
- Client 2's get operation overlaps in time with client 1's set operation
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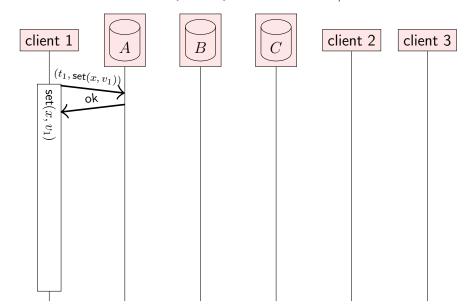


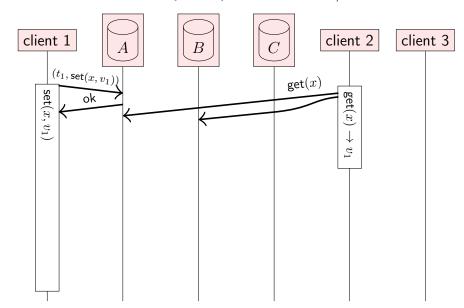
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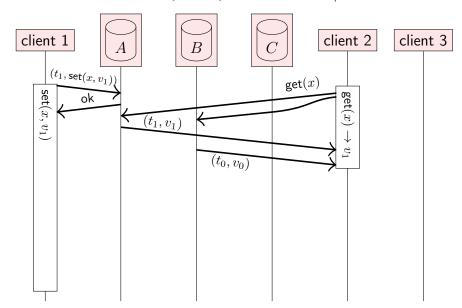


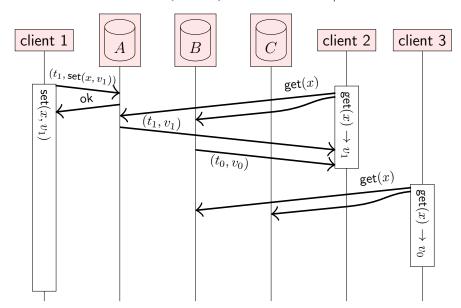
- Client 2's get operation overlaps in time with client 1's set operation
- Maybe the set operation takes effect first?
- Just as likely, the get operation may be executed first
- Either outcome is fine in this case

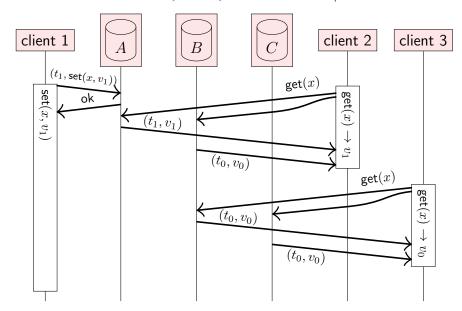


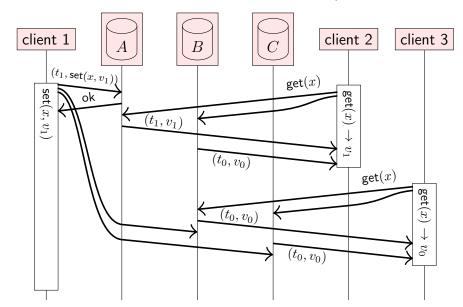


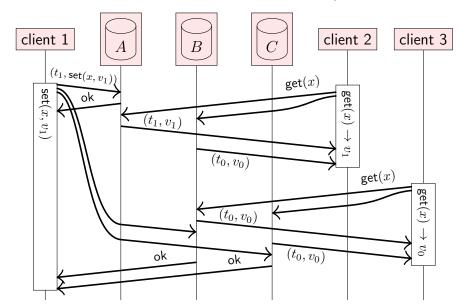




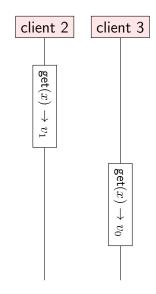


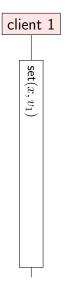


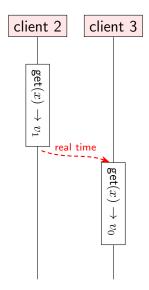


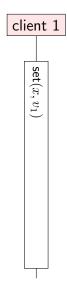




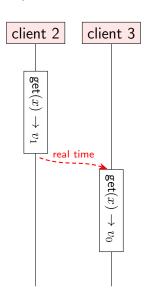


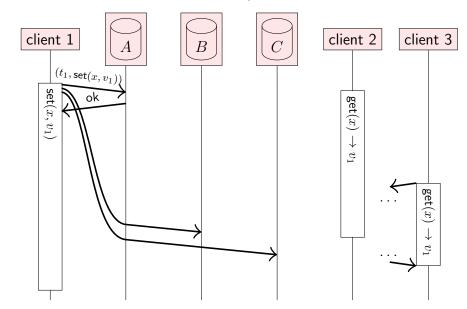


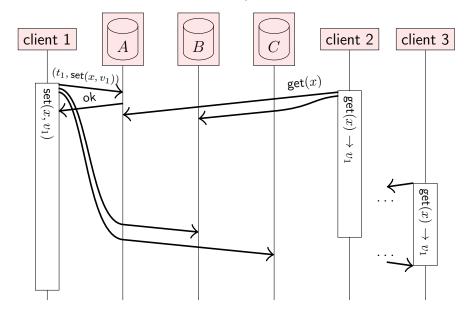


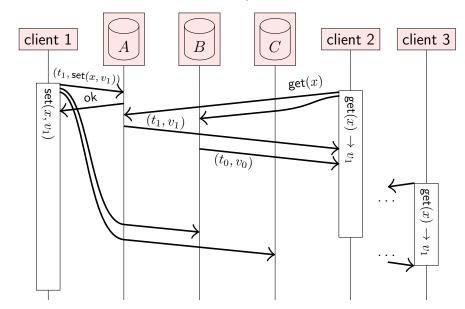


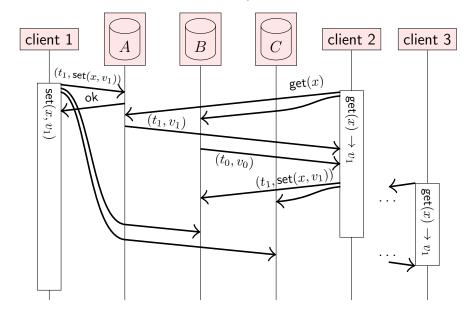
- Client 2's operation finishes before client 3's operation starts
- Linearizability therefore requires client 3's operation to observe a state no older than client 2's operation
- This example violates linearizability because v_0 is older than v_1

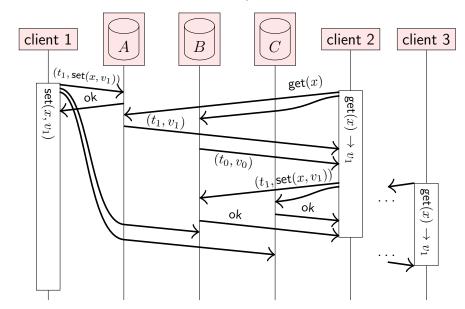




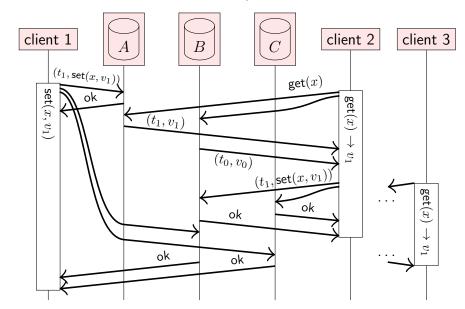








ABD: Making quorum reads/writes linearizable



This ensures linearizability of get (quorum read) and set (**blind write** to quorum)

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- ightharpoonup CAS(x, oldValue, newValue) sets x to newValue iff current value of x is oldValue
- Previously discussed in shared memory concurrency

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- Previously discussed in shared memory concurrency
- Can we implement linearizable compare-and-swap in a distributed system?
- **Yes:** total order broadcast to the rescue again!



Linearizable compare-and-swap (CAS)

```
on request to perform get(x) do
   total order broadcast (get, x) and wait for delivery
end on
on request to perform CAS(x, old, new) do
   total order broadcast (CAS, x, old, new) and wait for delivery
end on
on delivering (get, x) by total order broadcast do
   return localState[x] as result of operation get(x)
end on
on delivering (CAS, x, old, new) by total order broadcast do
   success := false
   if localState[x] = old then
       localState[x] := new; success := true
   end if
   return success as result of operation CAS(x, old, new)
end on
                                              4D + 4B + 4B + B + 900
```

Linearizability advantages:

- Makes a distributed system behave as if it were non-distributed
- ► Simple for applications to use

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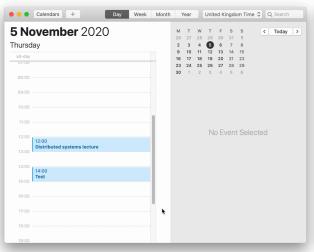
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Eventual consistency: a weaker model than linearizability. Different trade-off choices.

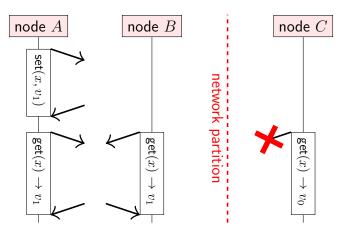




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12:00	Distr	ibuted	systems	s lectu	'e	
13:00						
14:00	Test					
15:00						
16:00						
17:00						
18:00						

The CAP theorem

A system can be either strongly **Consistent** (linearizable) or **Available** in the presence of a network **Partition**



C must either wait indefinitely for the network to recover, or return a potentially stale value



Replicas process operations based only on their local state.

If there are no more updates, **eventually** all replicas will be in the same state. (No guarantees how long it might take.)

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Properties:

- ▶ Does not require waiting for network communication
- Causal broadcast (or weaker) can disseminate updates
- ► Concurrent updates ⇒ **conflicts** need to be resolved



trength of assumptions

Problem	Must wait for communication	Requires synchrony
atomic commit	all participating nodes	partially synchronous

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atomic commit	all participating nodes	partially synchronous	,
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linearizable get/set	quorum	asynchronous	
eventual consistency, causal broadcast, FIFO broadcast	local replica only	asynchronous	

Lecture 8

Case studies

Nowadays we use a lot of **collaboration software**:

Examples: calendar sync (last lecture), Google Docs, . . .

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Families of algorithms:

- ► Conflict-free Replicated Data Types (CRDTs)
 - Operation-based
 - State-based
- ► Operational Transformation (**OT**)



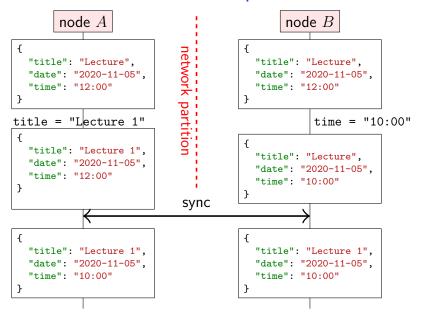
```
\mathsf{node}\ A
                                                         \mathsf{node}\ B
                                 network partition
"title": "Lecture",
                                                  "title": "Lecture",
"date": "2020-11-05",
                                                  "date": "2020-11-05",
"time": "12:00"
                                                  "time": "12:00"
```

```
\mathsf{node}\ A
  "title": "Lecture",
  "date": "2020-11-05",
  "time": "12:00"
title = "Lecture 1"
  "title": "Lecture 1",
  "date": "2020-11-05",
  "time": "12:00"
}
```

```
\mathsf{node}\ B
"title": "Lecture",
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```

```
node B
"title": "Lecture",
"date": "2020-11-05",
"time": "12:00"
          time = "10:00"
"title": "Lecture",
"date": "2020-11-05",
"time": "10:00"
```



Operation-based map CRDT

```
on initialisation do
    values := \{\}
end on
on request to read value for key k do
    if \exists t, v. (t, k, v) \in values then return v else return null
end on
on request to set key k to value v do
    t := \text{newTimestamp}() \triangleright \text{globally unique, e.g. Lamport timestamp}
    broadcast (set, t, k, v) by reliable broadcast (including to self)
end on
on delivering (set, t, k, v) by reliable broadcast do
    previous := \{(t', k', v') \in values \mid k' = k\}
    if previous = \{\} \lor \forall (t', k', v') \in previous. \ t' < t \ then
        values := (values \setminus previous) \cup \{(t, k, v)\}
    end if
end on
```

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Operation-based CRDTs

Reliable broadcast may deliver updates in any order:

- broadcast (set, t_1 , "title", "Lecture 1")
- broadcast (set, t_2 , "time", "10:00")

Operation-based CRDTs

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Recall strong eventual consistency:

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Operation-based CRDTs

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CRDT algorithm implements this:

- Reliable broadcast ensures every operation is eventually delivered to every (non-crashed) replica
- ► Applying an operation is **commutative**: order of delivery doesn't matter



State-based map CRDT

end on

The operator \sqcup merges two states s_1 and s_2 as follows:

```
s_1 \sqcup s_2 = \{(t, k, v) \in (s_1 \cup s_2) \mid \not\exists (t', k', v') \in (s_1 \cup s_2). \ k' = k \land t' > t\}
  on initialisation do
      values := \{\}
  end on
  on request to read value for key k do
      if \exists t, v. (t, k, v) \in values then return v else return null
  end on
  on request to set key k to value v do
      t := \text{newTimestamp}() \triangleright \text{globally unique, e.g. Lamport timestamp}
      values := \{(t', k', v') \in values \mid k' \neq k\} \cup \{(t, k, v)\}
      broadcast values by best-effort broadcast
  end on
  on delivering V by best-effort broadcast do
      values := values \sqcup V
```

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State-based CRDTs

Merge operator \sqcup must satisfy: $\forall s_1, s_2, s_3...$

- ▶ Commutative: $s_1 \sqcup s_2 = s_2 \sqcup s_1$.
- ▶ Associative: $(s_1 \sqcup s_2) \sqcup s_3 = s_1 \sqcup (s_2 \sqcup s_3)$.
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- State-based CRDT can tolerate message loss/duplication

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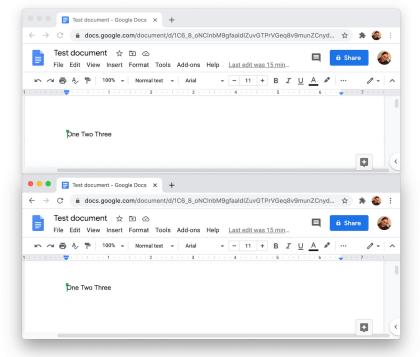
State-based versus operation-based:

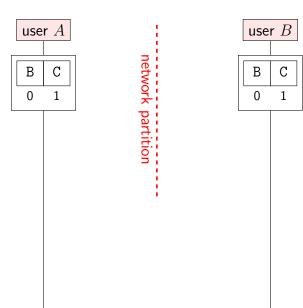
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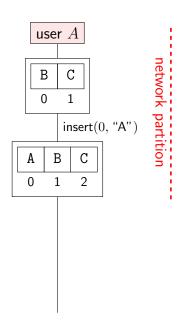
Not necessarily uses broadcast:

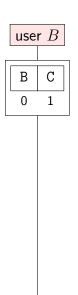
► Can also merge concurrent updates to replicas e.g. in quorum replication, anti-entropy, . . .

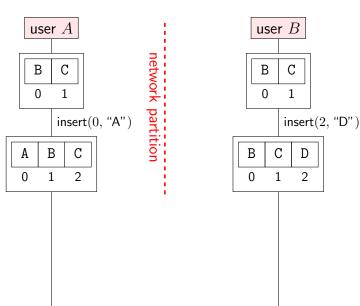


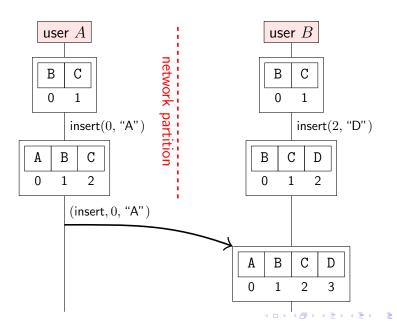


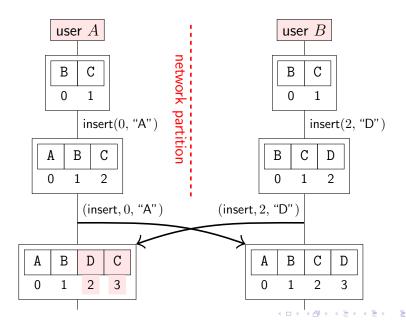




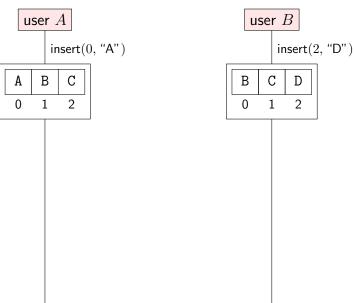




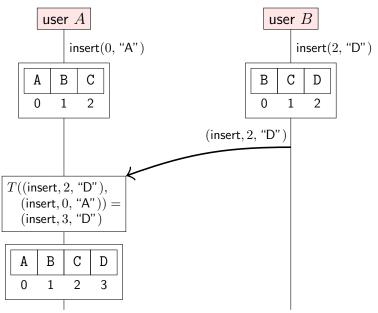




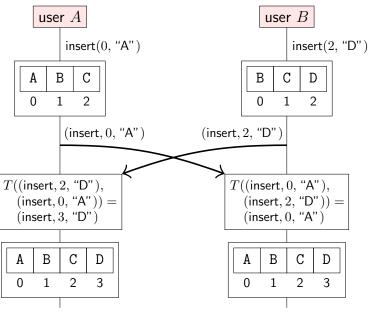
Operational transformation

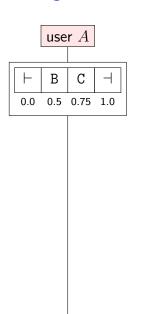


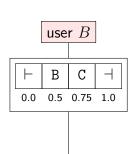
Operational transformation

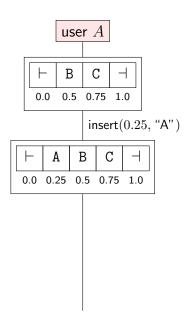


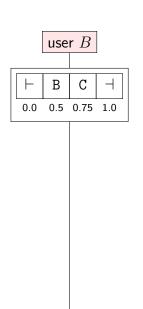
Operational transformation

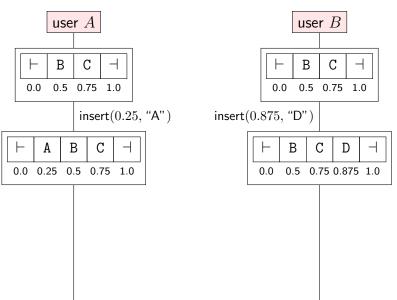


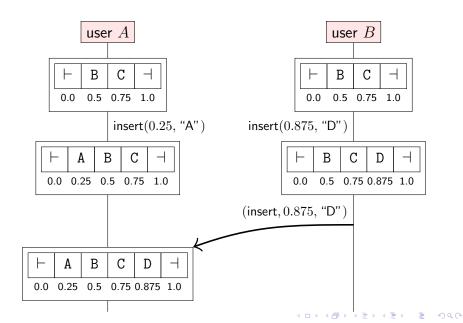


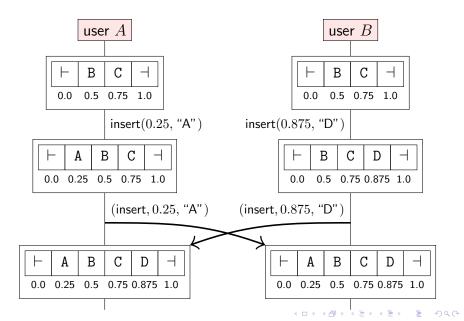












Operation-based text CRDT (1/2)

end on

```
function ElementAt(chars, index)
    min = the unique triple (p, n, v) \in chars such that
           \nexists (p', n', v') \in chars. \ p' 
    if index = 0 then return min
   else return Element At(chars \setminus \{min\}, index - 1)
end function
on initialisation do
    chars := \{(0, \mathsf{null}, \vdash), (1, \mathsf{null}, \dashv)\}
end on
on request to read character at index index do
   let (p, n, v) := \text{ElementAt}(chars, index + 1); return v
end on
on request to insert character v at index index at node nodeId do
    let (p_1, n_1, v_1) := \text{ElementAt}(chars, index)
    let (p_2, n_2, v_2) := \text{ELEMENTAT}(chars, index + 1)
```

broadcast (insert, $(p_1 + p_2)/2$, nodeId, v) by causal broadcast

4□ ► 4□ ► 4 ≧ ► 4 ≧ ► 9 < 0</p>

Operation-based text CRDT (2/2)

```
on delivering (insert, p, n, v) by causal broadcast do
    chars := chars \cup \{(p, n, v)\}
end on
on request to delete character at index index do
    let (p, n, v) := \text{ElementAt}(chars, index + 1)
    broadcast (delete, p, n) by causal broadcast
end on
on delivering (delete, p, n) by causal broadcast do
    chars := \{(p', n', v') \in chars \mid \neg (p' = p \land n' = n)\}
end on
```

- Use causal broadcast so that insertion of a character is delivered before its deletion
- ▶ Insertion and deletion of different characters commute

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The interesting bit: read-only transactions require no locks!



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- lacktriangle Read-only transaction T_r has snapshot timestamp t_r
- ► T_r ignores values with $t_w > t_r$; observes most recent value with $t_w < t_r$



Must ensure that whenever $T_1 \to T_2$ we have $t_1 < t_2$.

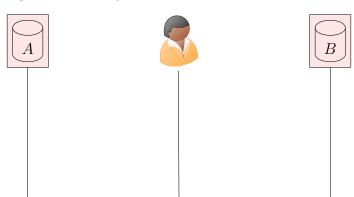
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- Problem: linearizability depends on real-time order, and logical clocks may not reflect this!

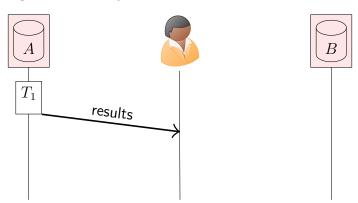
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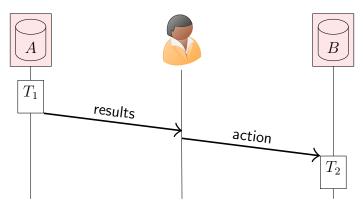
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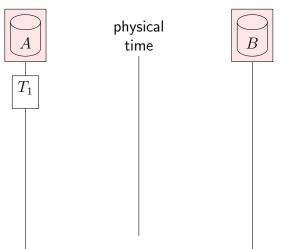


Obtaining commit timestamps

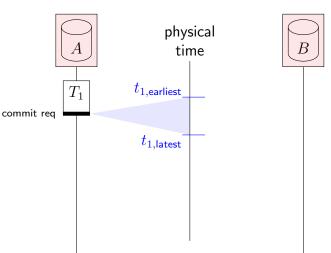
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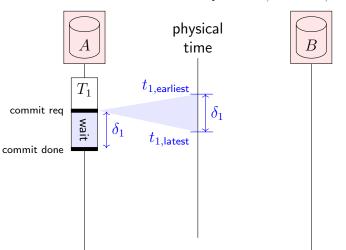




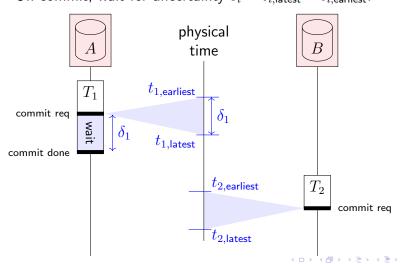
Spanner's TrueTime clock returns $[t_{\text{earliest}}, t_{\text{latest}}]$. True physical timestamp must lie within that range.



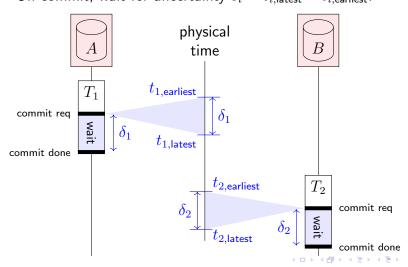
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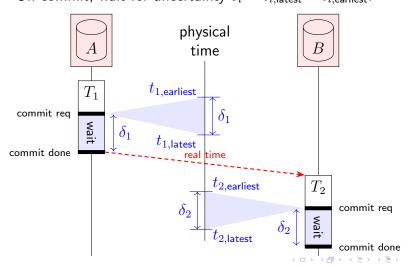
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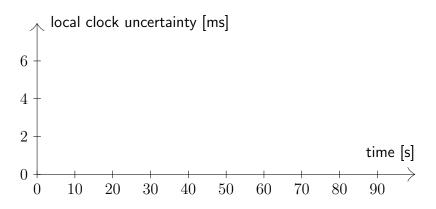
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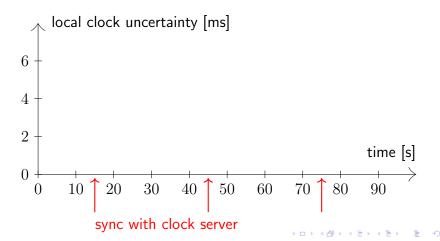
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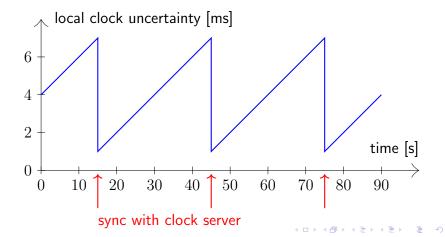
Clock servers with **atomic clock** or **GPS receiver** in each datacenter; servers report their clock uncertainty.



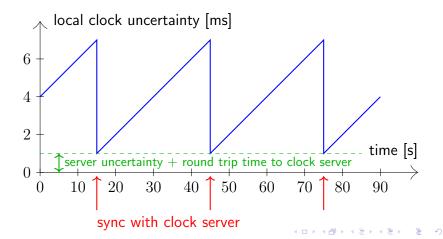
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That's all, folks!

Any questions? Email tim.harris@gmail.com!

Summary:

- Distributed systems are everywhere
- You use them every day: e.g. web apps
- Key goals: availability, scalability, performance
- Key problems: concurrency, faults, unbounded latency
- Key abstractions: replication, broadcast, consensus
- No one right way, just trade-offs