



# **CS 5412: LECTURE 9**

## **TIMESTAMPED DATA**

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# TODAY: DRILL DOWN ON TIME

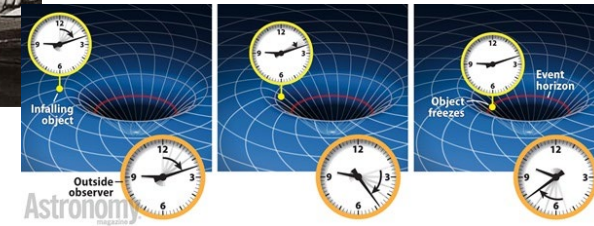


Last time we discussed time more as an active aspect of a coordinated system (one of a few dimensions in which an IoT system might be active).

But once a sensor reading is captured and stored, there is also a temporal aspect to data analysis.

What can we say about time for data and events “inside” a data store?

# TIME IN THE REAL WORLD



Einstein was first to really look closely at this topic.

It led to his theory of relativity: *Time has no absolute meaning.*

But Einstein was thinking about particles moving at near the speed of light, or near black holes. Do those ideas apply in other settings?

# TIME IN COMPUTER SYSTEMS



Often, we put “timestamps”  
on IoT sensor records

In IoT, time is tricky to work with for many reasons:

- Even with GPS receivers, it can be hard to get a good fix, so time can drift
- IoT sensors often lack GPS and their clocks need to be reset via an event, but then might drift by seconds per day
- Sensors can also fail, and this includes their clocks.

Thus a timestamped event may have inaccurate time!



# IN WHAT WAYS CAN WE TALK ABOUT TIME?

First, whenever we use time in an IoT setting, it is important to track the time source and the associated skew:

- Without GPS time, sensor time will drift by seconds/day
- With GPS time, clocks can be accurate to within about 1ms
- With special purpose hardware for synchronization, the machines in a cloud would be able to share a clock and be accurate to a few *us*.
- ... but today's cloud computers don't have that form of shared clocks, and if virtualized, clocks can be quite inaccurate! A total mess!

# VENDORS PREFER LIMITED ACCURACY!



Several recent security problems have involved an attacker who places a monitoring program on the same machine that some security code is on. The attacker is assumed to have the source code for the application it is attacking.

The monitoring program measures timing properties of the memory and caching hardware at very high accuracy and is able to deduce contents of the memory state of the attacked program.

It seems doubtful that this would work, but several exploits show that it really does work! To reduce the risk, cloud vendors make it hard to measure time.

# LAMPORT'S CAUSAL ORDER



Leslie Lamport

Leslie Lamport considered this issue, but in computing systems

- He actually started as a physicist inspired by Einstein, but went on to formalize distributed protocols, and won the Turing Award
- Primarily a theoretician, but he also was the author of Latex
- Especially good at elegant ways of posing problems and solving them

He suggested that an important aspect of consistency should involve “consistency with respect to past events”. He calls this “causal” consistency



# HOW DOES HE DEFINE CAUSALITY?

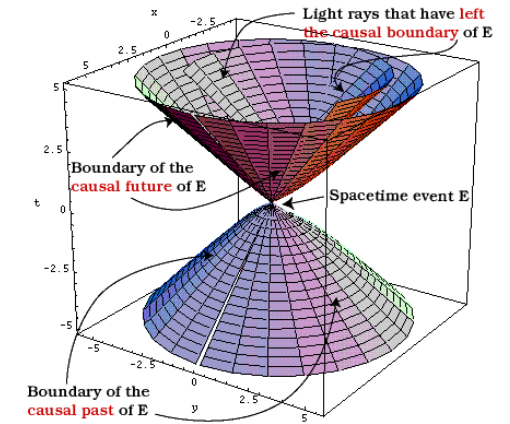
Suppose that event A occurs in a data center, and then later event B.

Did A “cause” B to happen?

- What if A was at 10am, and B at 11:30pm. Does knowing time help?
- What if A was a command to register a new student, and B was an internal action that creates her “meal card” account?
- What if A was an email from the department asking me about my teaching preferences, and B was my reply?



# HOW DOES HE DEFINE CAUSALITY



For Leslie, event A causes event B if there was a computation that somehow was triggered by A, and B was part of it. Inspired by physics!

But this is hard to discover automatically.

Instead, Leslie focused on **potential** causality: A “might” have caused B.

Under what conditions is this possible?

- Somehow, *information must flow* from A to B.



# NOTATION FOR REPRESENTING CAUSALITY

Leslie proposes that we write  $A \rightarrow B$  if  $A$  potentially caused  $B$ .

He suggests that we use the words “happened before” for  $\rightarrow$

Now the question arises: is  $\rightarrow$  just a mathematical concept, or can we build a practical tool for tracking causality in real systems?

# WHY WOULD WE WANT TO TRACK $A \rightarrow B$ ?

Consider the Securities and Exchange Commission.

For them, A might be “information about stock X” and B “a trade of X”.

An insider trade occurs if someone with non-public information takes advantage to trade a stock before that information comes out. So if “John learned that the IBM quantum computer showed promise”, then bought IBM stock, perhaps John violated the insider trading law.

# LAMPORT'S POINT

Simply seeing data records in which John talks to his friend at IBM at 10:00am and then buys IBM stock at 10:01am might not be “proof” of criminality. These days the cloud might participate in all of these events.

If the records were timestamped by the identical clock, and the clock isn't faulty, this really would be proof.

But if the records came from different computers, clock imprecision could be creating an illusion. If we track actual  $\rightarrow$ , we would be confident.

# TRACKING $A \rightarrow B$

Leslie first considered normal clocks. But they don't track  $\rightarrow$

- Here, he took his inspiration from Einstein
- “*Time is an illusion.*” Einstein went on to draw space-time diagrams.

So Leslie asked: “Can we use space-time diagrams as the basis of a new kind of “logical clock”?”

- If  $A \rightarrow B$ , then  $\text{LogicalClock}(A) < \text{LogicalClock}(B)$
- If  $\text{LogicalClock}(A) < \text{LogicalClock}(B)$ , then  $A \rightarrow B$



# DEVELOPING A SOLUTION

Suppose that every computer ( $P, Q, \dots$ ) has a local, private integer

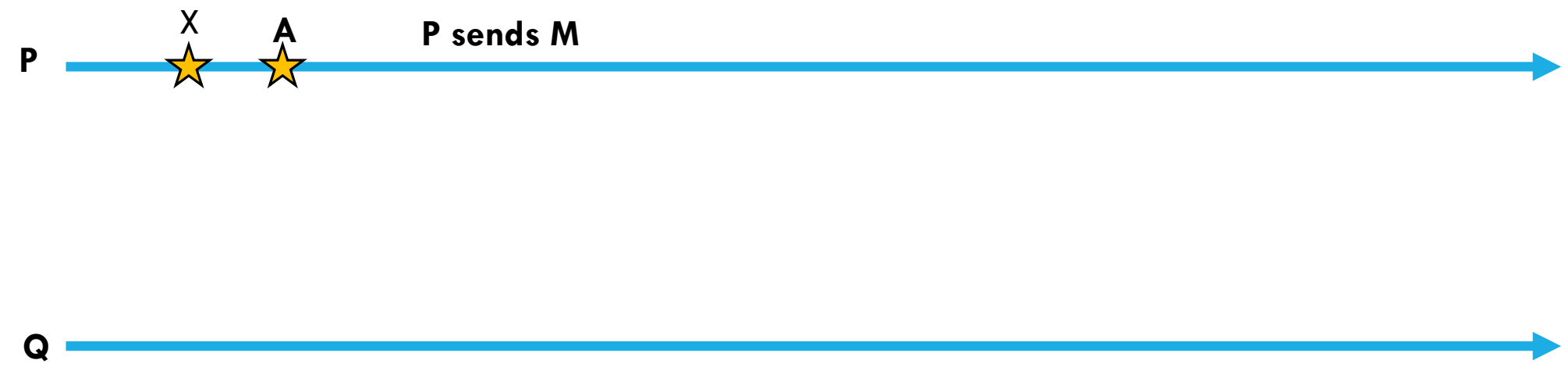
Call these  $\text{LogicalClock}_P$  and  $\text{LogicalClock}_Q$  etc.

Each time something happens, increment the clock.

- Now, if  $A$  and  $B$  happen at  $P$ , the  $\text{LogicalClock}_P$  can tell us that  $A \rightarrow B$ .
- But what if  $A$  is on machine  $P$ , and  $B$  happens on  $Q$ ?

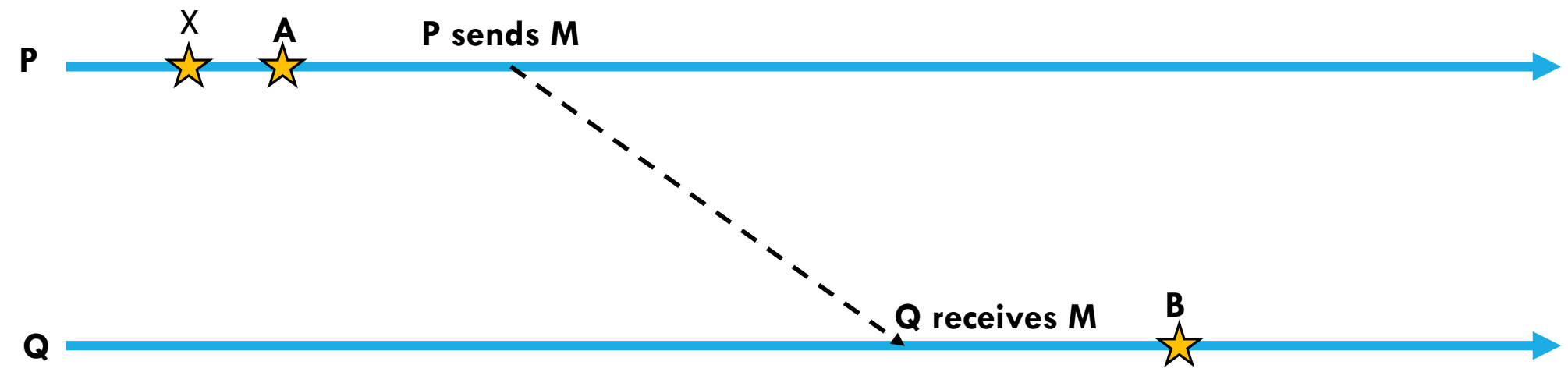


# A SPACE-TIME DIAGRAM FOR THIS CASE





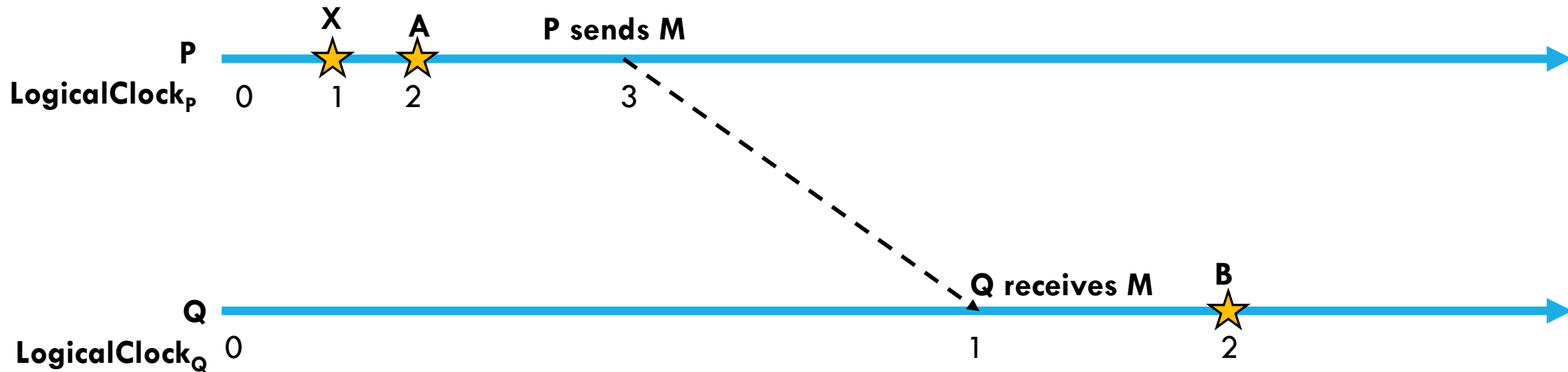
# A SPACE-TIME DIAGRAM FOR THIS CASE





# A SPACE-TIME DIAGRAM FOR THIS CASE

Uncoordinated counters don't solve our problem



Here, A and B end up with the identical Time, so we incorrectly conclude that A did not *happen before* B



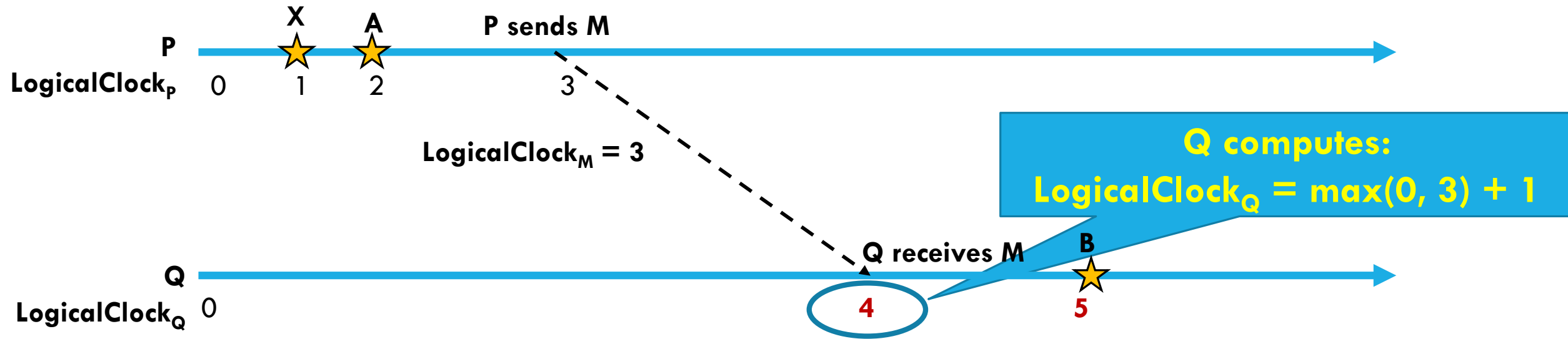
# AHA!

But notice that in the diagram, the “receive” occurs when  $\text{LogicalClock}_B = 1$ .  
Yet the “send” of  $M$  was at  $\text{LogicalClock}_A = 3$ .

So Lamport proposes this fix:

- Each time an interesting event occurs at  $P$ , increment  $\text{LogicalClock}_P$
- If  $P$  sends  $M$  to  $Q$ , include  $\text{LogicalClock}_P$  in  $M$ . When  $Q$  receives  $M$ ,  
$$\text{LogicalClock}_Q = \text{Max}(\text{LogicalClock}_Q, \text{LogicalClock}_M) + 1$$

# A SPACE-TIME DIAGRAM FOR THIS CASE



# WE NOW HAVE A CHEAP PARTIAL SOLUTION!

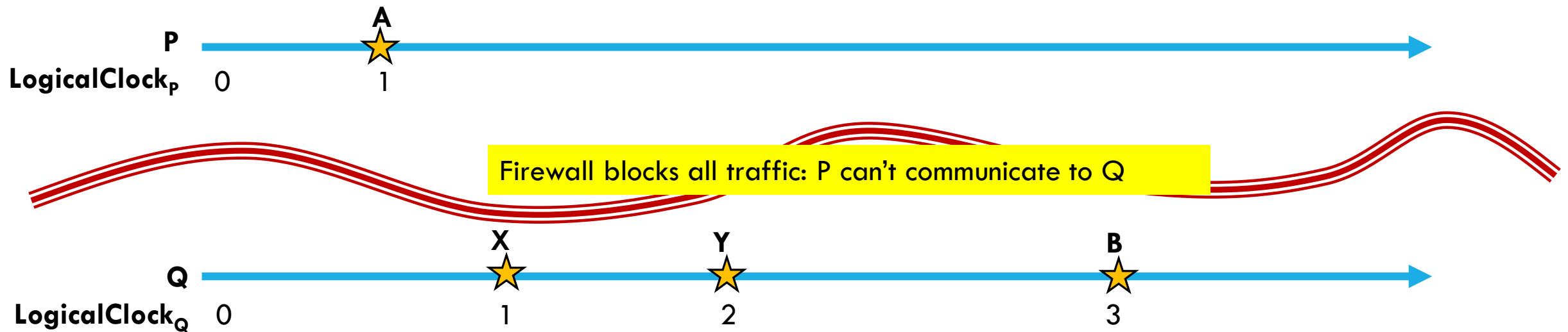
With Lamport's logical clocks, we pay a small cost (one integer per machine, to keep the clock, and some space in the message)

Let's use  $\text{LogicalClock}(X)$  to denote the relevant LogicalClock value for  $x$ . We can time-stamp events and messages.

- If  $A \rightarrow B$ , then  $\text{LogicalClock}(A) < \text{LogicalClock}(B)$
- But... if  $\text{LogicalClock}(A) < \text{LogicalClock}(B)$ , *perhaps A didn't happen before B!*

# A SPACE-TIME DIAGRAM FOR THIS CASE

With logical clocks, even if P and Q never talk, we might have  $\text{Time}(A) < \text{Time}(B)$



Here, if we claim that  $\text{LogicalClock}(A) < \text{LogicalClock}(B) \Rightarrow A \rightarrow B$ , this is nonsense! In fact  $\neg(A \rightarrow B), \neg(B \rightarrow A)$ . (A and B are “concurrent”)

# LOGICAL CLOCKS ONLY WORK IN ONE DIRECTION.

Logical clocks approximate the causal happens-before relationship, but only in an “if-then” sense, not “If and only if”.

Yet, they are useful: Lamport gives many examples where this suffices.

We actually can do better, but at the “cost” of higher space overhead.

# DETECTING INSIDER TRADING

The S.E.C.\* wants to detect that “John learned of the good news from Lilly, CEO of Zebra Corp. Then he purchased stock before the market heard.”

If A was John learning, and B was the stock purchase, then the SEC wants to look at  $\text{LogicalClock}(A) < \text{LogicalClock}(B)$ , and conclude “ $A \rightarrow B$ ”.

But logical clocks don't let us conclude this. And John might insist that “I kept a log of call times, and I spoke to Lilly after the IBM market announcement. Perhaps some clock drifted and the S.E.C. has its time sequence wrong.”

\* Securities Exchange Commission

# INTUITION BEHIND VECTOR CLOCKS

Suppose that we had a fancier clock that could act like logical clocks do (with the “take the max, then add one” rule).

But instead of a single counter, what if it were to count “events in the causal past of this point in the execution”, tracking events on a per-process basis?

For example, a VectorClock value for A = [5,7] might mean “event A happens after 5 events at P, and 7 events at Q”.



# VECTOR CLOCKS ARE EASY TO IMPLEMENT



A vector clock has one entry per machine.  $VT(A) = [3, 0, 7, 1]$

- If an event occurs at  $P$ ,  $P$  increments *its own entry* in the vector
- When  $Q$  receives  $M$  from  $P$ ,  $Q$  computes an entry-by-entry max, then increments its own entry (because a “receive” is an event, too)

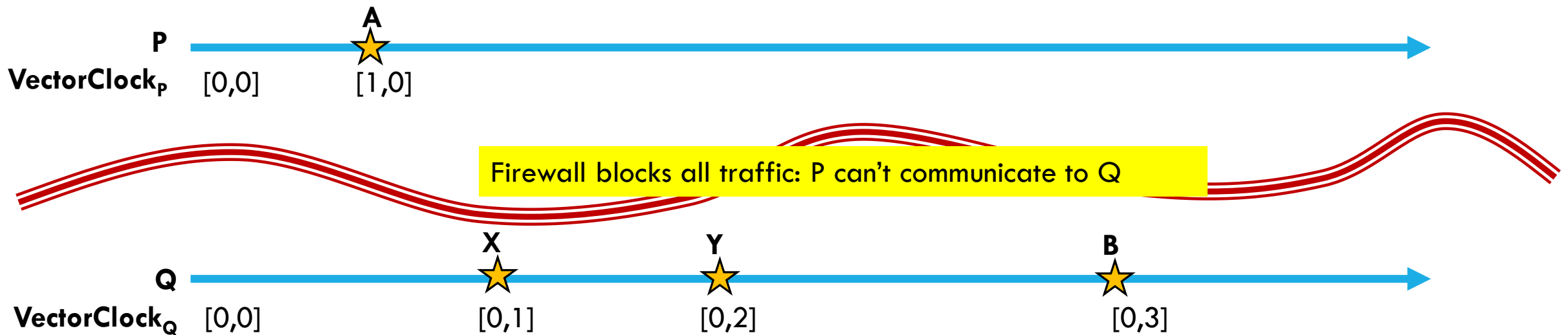
VectorClock comparison rule:

**Define  $VT(A) < VT(B)$  if  
 $VT(A) \leq VT(B)$ ,  
but  $VT(A) \neq VT(B)$**

**Now,  $VT(A) < VT(B)$  iff  $A \rightarrow B$**

# A SPACE-TIME DIAGRAM FOR THIS CASE

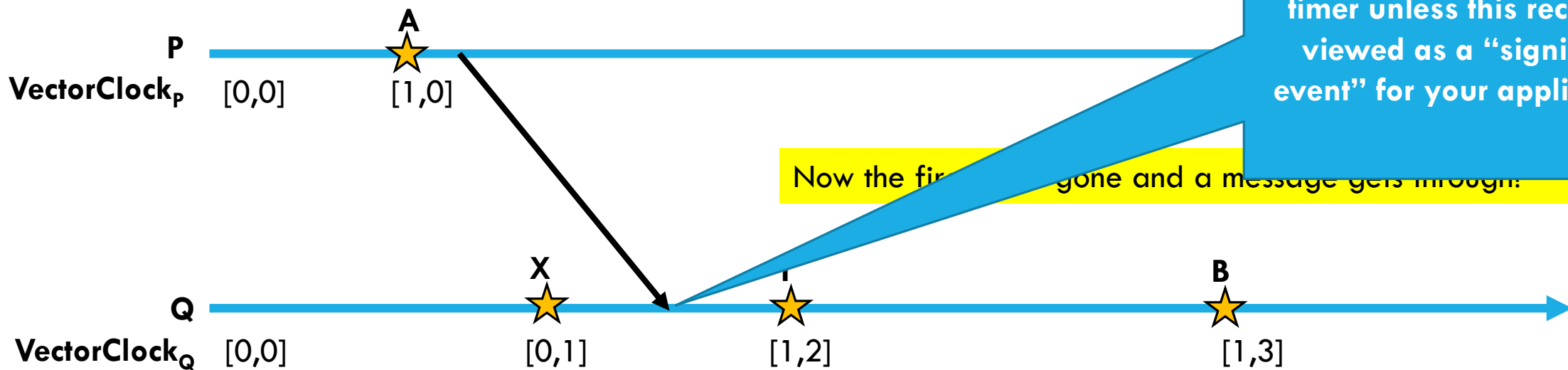
Case A: Suppose that P and Q never interact.



With vector clocks we can see that A is concurrent with X, Y and B. We can use the comparison rule to show this, for example that  $\neg(A \rightarrow B)$  and  $\neg(B \rightarrow A)$ .

# A SPACE-TIME DIAGRAM FOR THIS CASE

Case B: P sends a message to Q after A, and it is received



The vector timestamps show that A happens before B (and also, before Y).

# VECTOR CLOCKS SOLVE THE S.E.C. PROBLEM!

A: John spoke to his friend Lilly.

Then the message  $M$  was to tell his stock broker to “Buy IBM futures ASAP!”  
B was the purchase. Our goal: Deduce that  $A \rightarrow B$  using just a database with information about A, and information about B, including timestamps.

We just saw that

$$\text{VectorClock}(A) < \text{VectorClock}(B) \Rightarrow A \rightarrow B!$$

# SO WHY NOT ALWAYS USE VECTOR CLOCKS?

They represent happens-before with full accuracy, which is great.

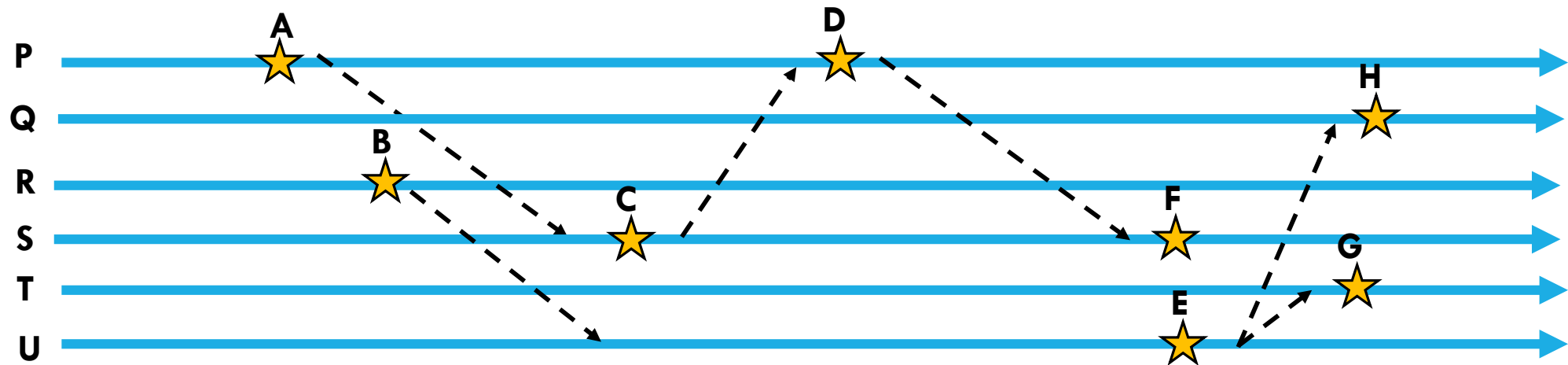
But you need one vector entry per process in your application. For a small  $\mu$ -service this would be fine, but if the vector would become large, the overheads are an issue.

So, we try to use a LogicalClock before considering a VectorClock.

# MORE FUN WITH CAUSALITY

Working with Mani Chandy (CalTech), Lamport also showed that you can use  $\rightarrow$  to define “now” in a way that makes sense even for a fully distributed system

He draws a complex space-time picture, perhaps this one:



# CONSISTENT CUTS AND SNAPSHOTS

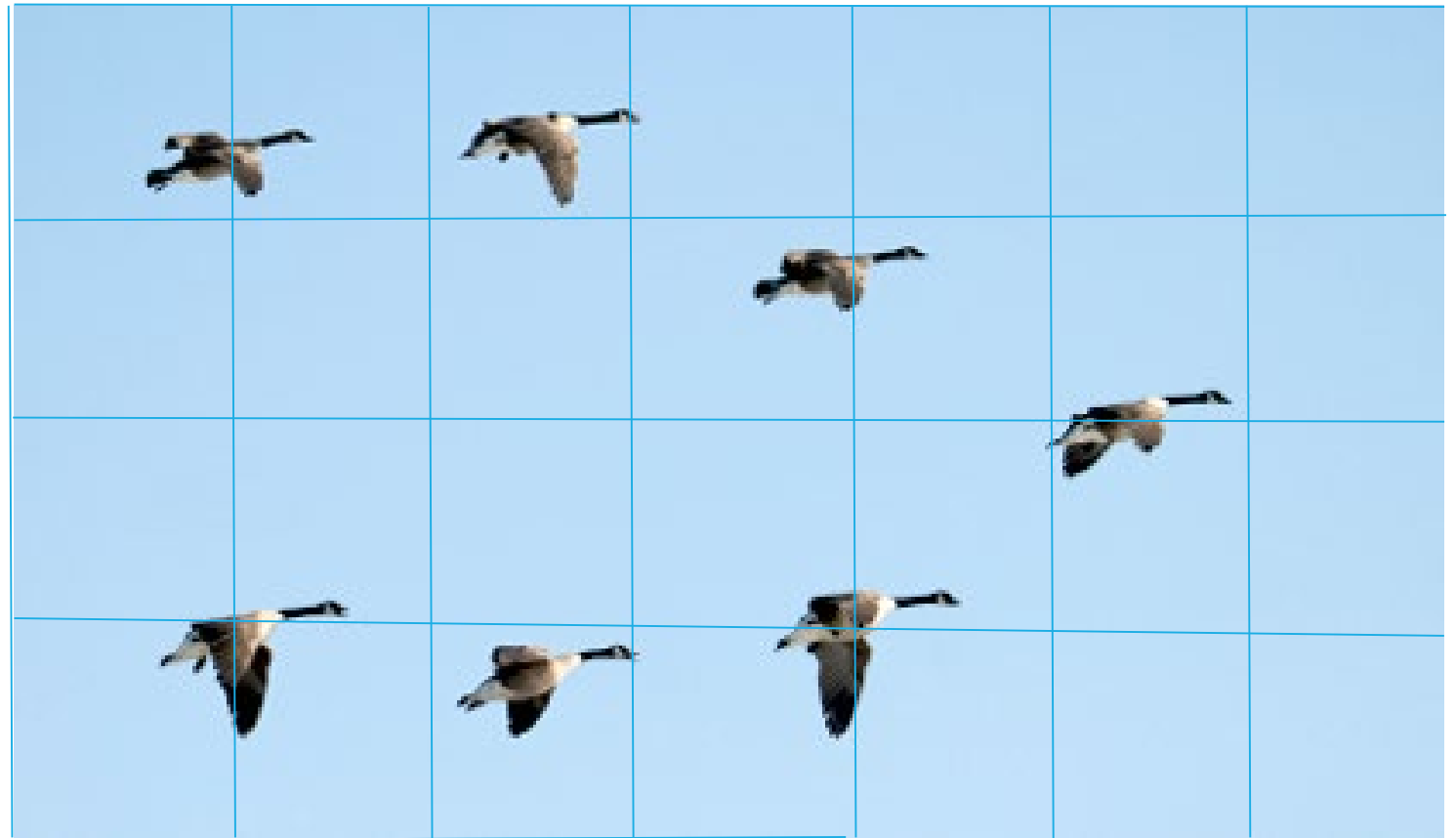
They asked: Suppose I visit each node, each at some point in time. Can we extend consistency to cover such a case (“consistent cut”)

Or even fancier: what if each node makes a checkpoint for me when I visit it along a cut. Can we end up with a “consistent snapshot”, like a photo?

# CONSISTENT AND INCONSISTENT SNAPSHOT

*Truth: 7 Geese in a V formation*

*Imagine taking  
photos of our  
geese one by  
one and creating  
a tiled mashup*

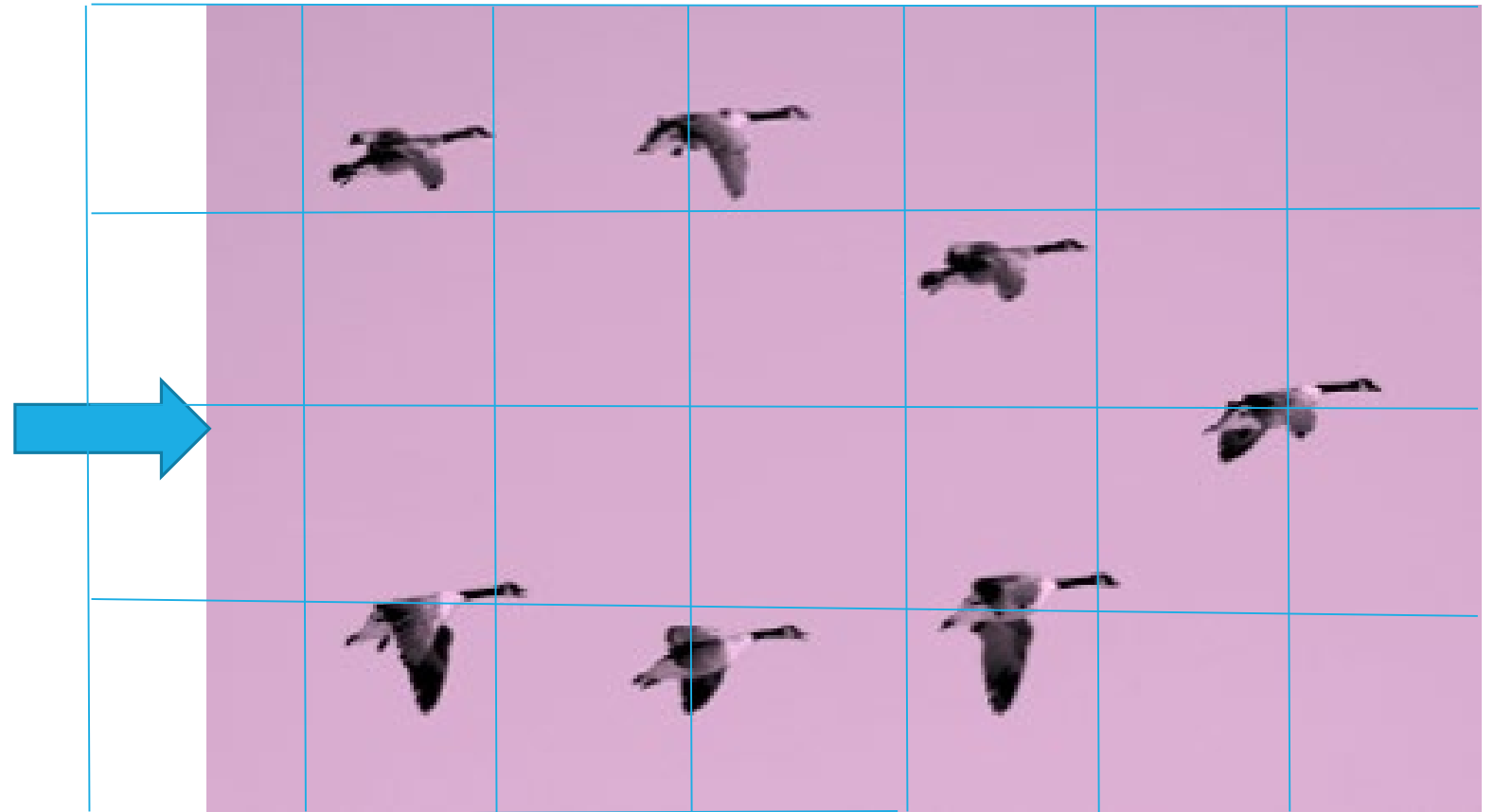




# CONSISTENT AND INCONSISTENT SNAPSHOT

*Truth: 7 Geese in a V formation*

*But suppose  
they were in  
motion while  
you did this*

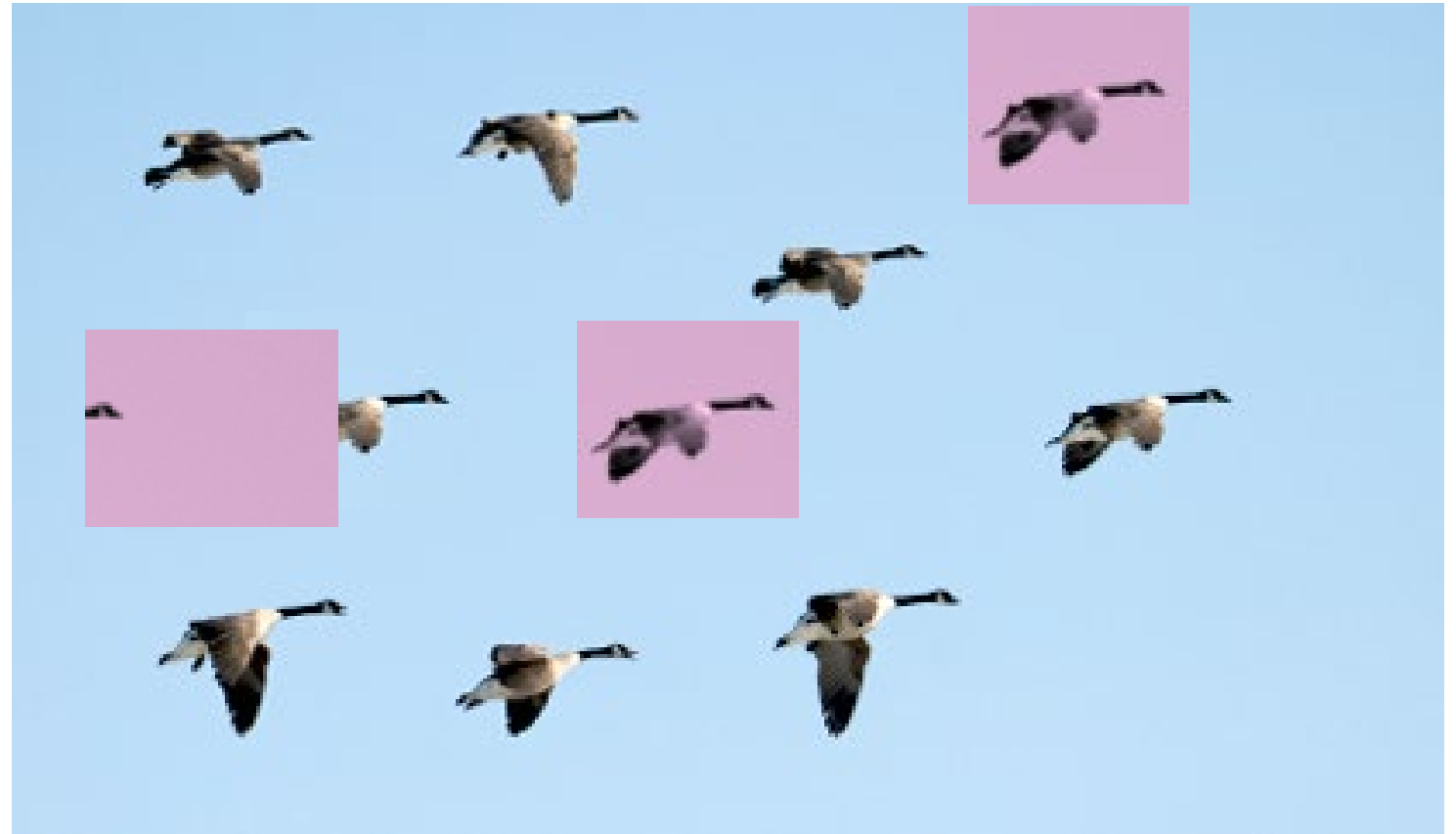


# CONSISTENT AND INCONSISTENT SNAPSHOT

*Truth: 7 Geese in a V formation*

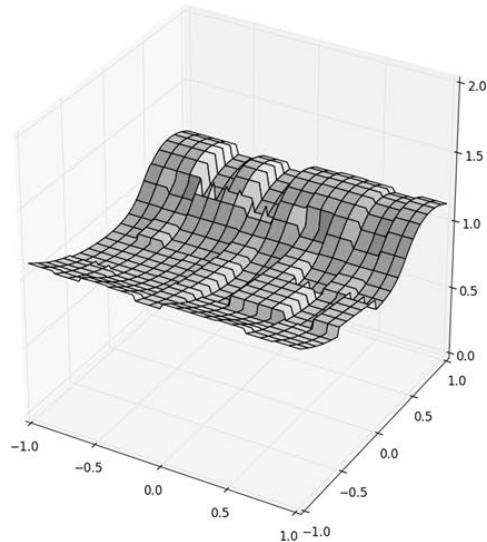
*Without coordination, some vanish, some are duplicated!*

*You might even see a goose that shifted to avoid collision with another goose, but see that other goose in a much earlier place, before it even got close.*

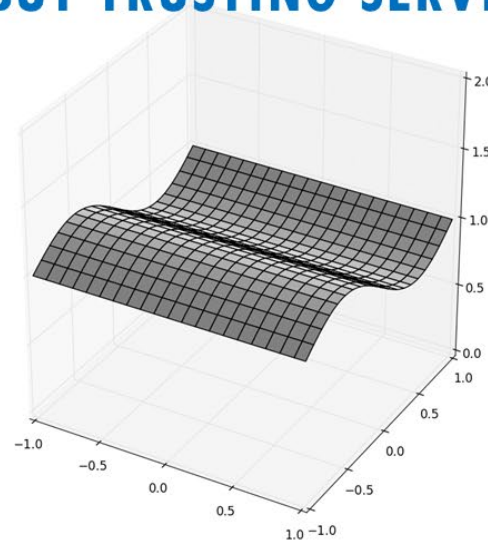


# THIS IS THE SAME ISSUE WE SAW IN OUR LECTURE ON CASCADE

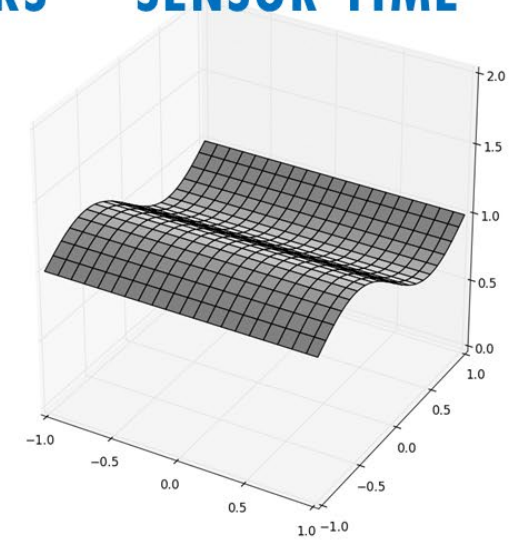
**HDFS**



**CASCADE USING CONSISTENT CUTS BUT TRUSTING SERVER CLOCKS**



**CASCADE WITH SENSOR TIME**



Inconsistent data is a problem. How does Cascade achieve consistency?



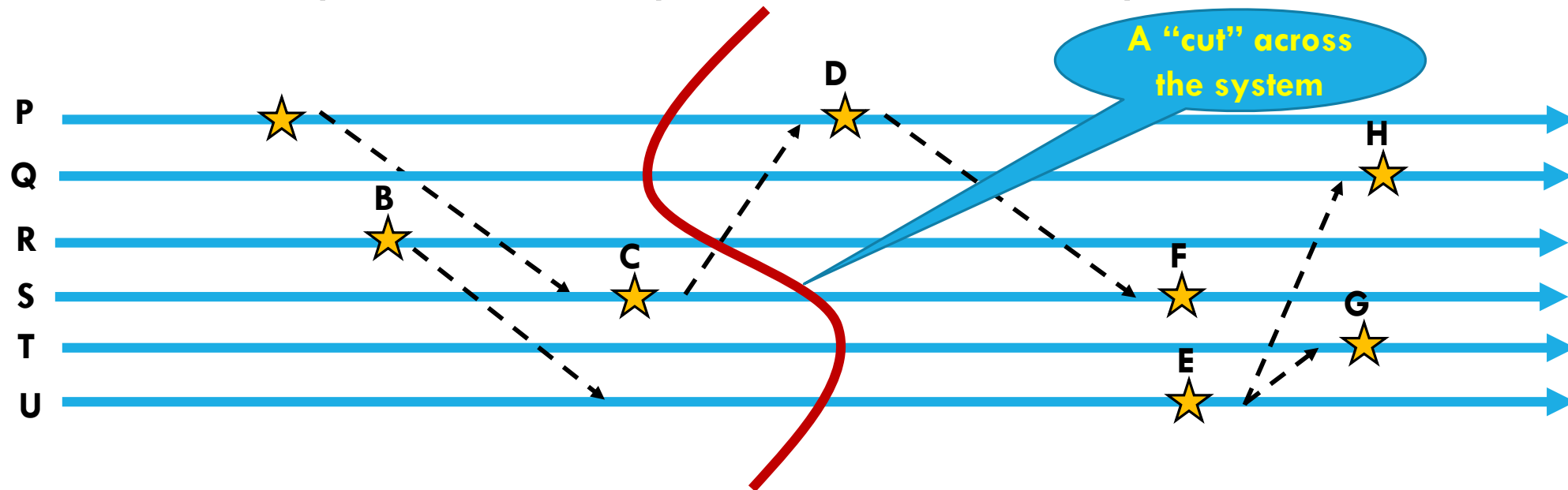
# CONSISTENT SNAPSHOT

If we use the method of Chandy and Lamport we get a consistent snapshot: there won't be any duplicates or mashup effect!

Goal of a consistent snapshot is to let us combine data from multiple processes (machines) in a distributed system, but only count each thing once, with no causal gaps or duplication.

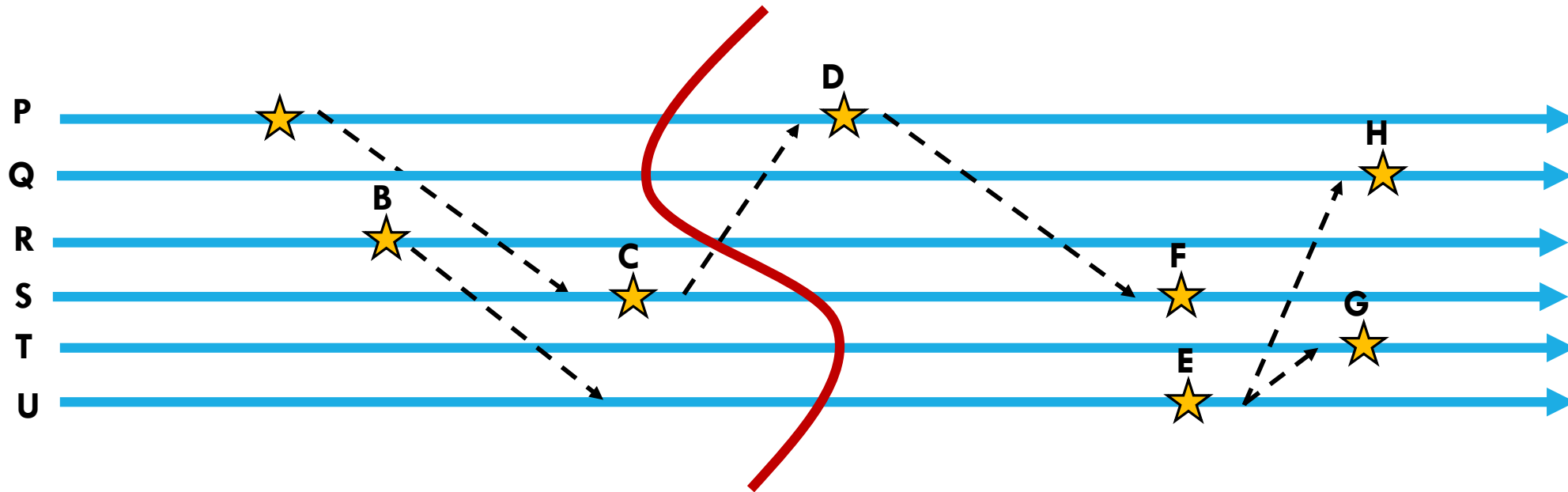
# CONSISTENT CUTS AND SNAPSHOTS

Recall: Lamport looks at “pictures” of such a system, like these



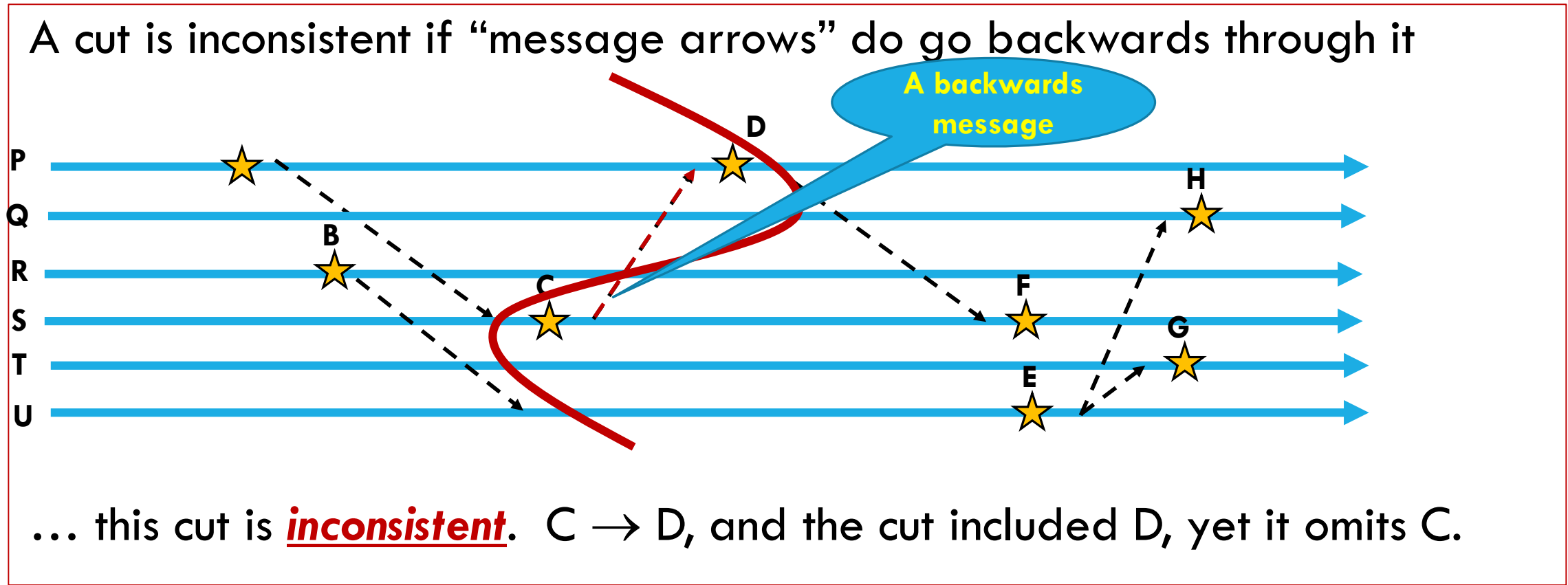
# CONSISTENT CUTS AND SNAPSHOTS

A cut is consistent if no “message arrows” go backwards through it



... this cut is a consistent one.

# CONSISTENT CUTS AND SNAPSHOTS



# A CONSISTENT CUT IS LIKE A PHOTO

It shows a state the system might actually have once been in

You could use that state for garbage collection, or to do an audit of a bank, or to detect deadlocks.

But an inconsistent cut is *broken*. It omits parts of the past and any conclusion from it would be incorrect. A real system could never have been in an inconsistent state of this kind.



# HOW TO CREATE A CONSISTENT SNAPSHOT

One option is to briefly **pause the whole system**.

When all processes have paused, make snapshots of their “state”. Also record the contents of any message channels.

This is provably a consistent snapshot. But the pause might be noticeable

# FANCIER ALGORITHM

Chandy and Lamport also have a fancier “online” algorithm

We will only give an overview today, but you can find all the details in their paper (linked to the syllabus).

It avoids the need to pause the system, but does require sending messages.

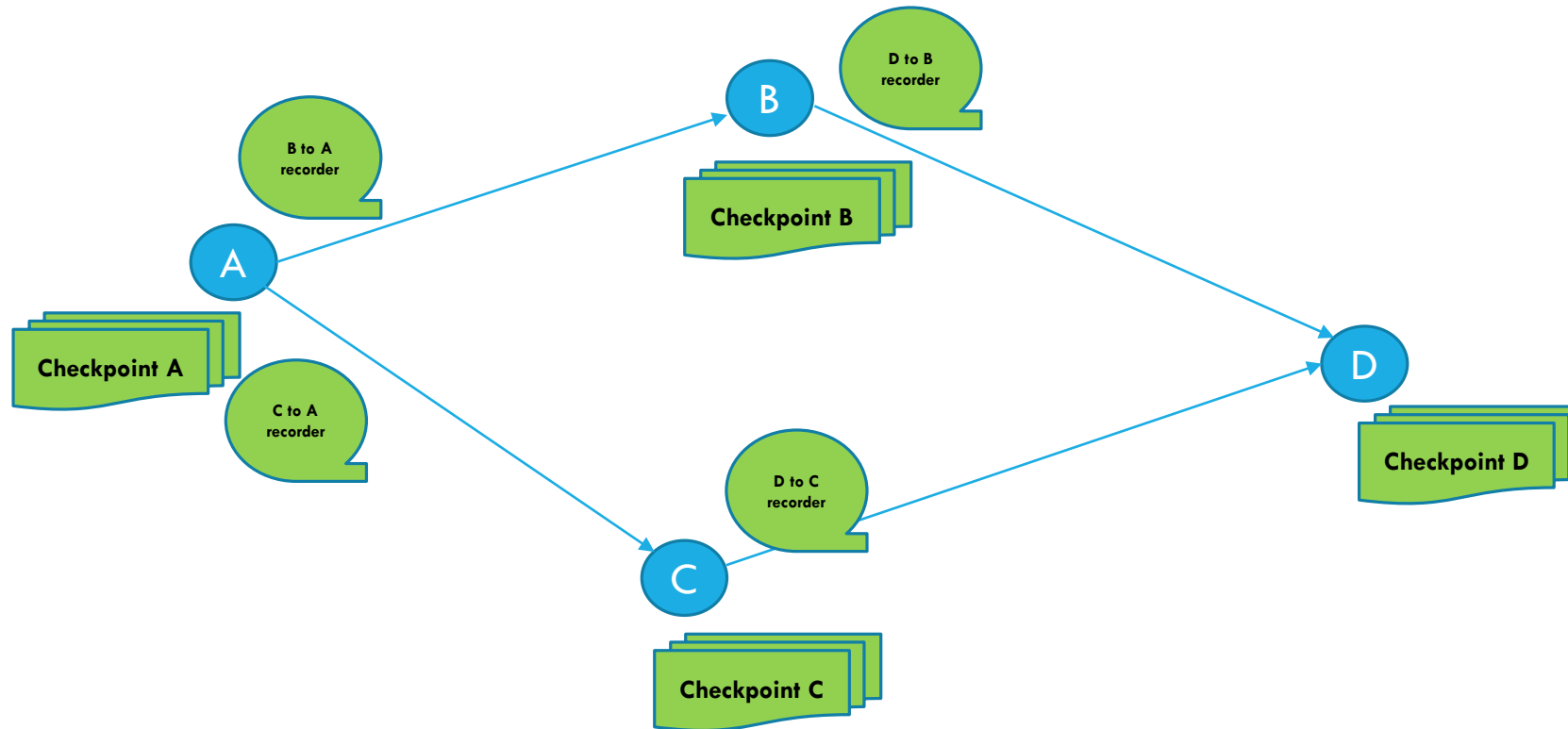
# THE ONLINE CHANDY/LAMPORT ALGORITHM

Each process is assumed to have a checkpoint mechanism, and a channel “recording” tool. Snapshot = {(checkpoint, channel-contents)} from all processes.

To start a snapshot, some process

1. Takes a checkpoint
2. Turns on the channel records for its connections to other processes.
3. Sends them a message: “Snapshot in progress”
4. When it receives “Snapshot is in progress” from a process, it turns off the channel recorder. Once all channels are recorded, its snapshot is done.

# THE ONLINE CHANDY/LAMPORT ALGORITHM



Note that we don't need a channel recording for channels we received "snapshot underway" from at the outset. Those have "empty" state

# CASCADE HAS AN EVEN EASIER OPTION

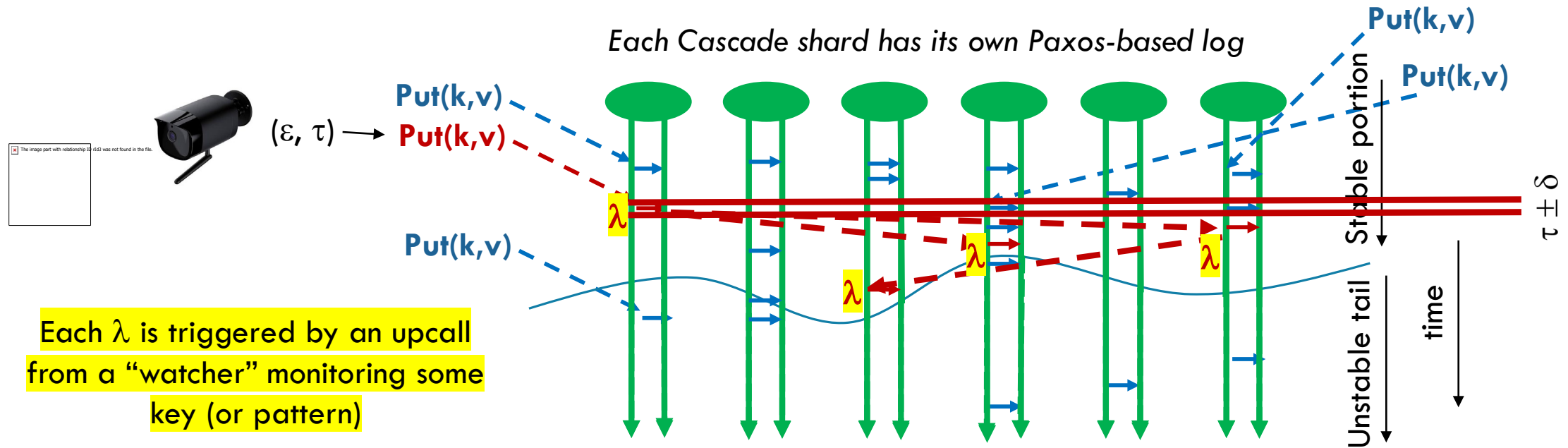
In Cascade, as data is accumulated, it becomes stable and that portion of the log or history won't change again.

This enables Cascade to use a special kind of timestamp to “find” a consistent cut in the logs. The cut is just a point in each log such that the collection of log “prefixes” is closed under  $\rightarrow$ .

The Cascade algorithm doesn't pause the system or send extra messages

# VISUALIZING CASCADE CONSISTENCY

Each Cascade shard has its own Paxos-based log

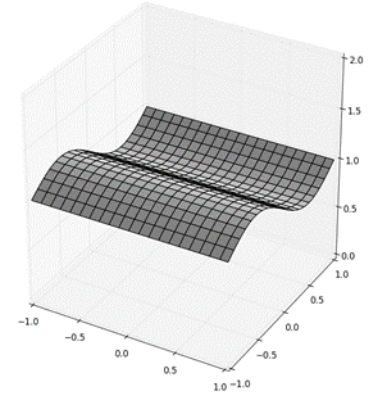


Each  $\lambda$  is triggered by an upcall from a “watcher” monitoring some key (or pattern)

A temporal query for time  $\tau$  sees a consistent cut at  $\tau \pm \delta_{\text{clock}}$ .

Queries to unstable data must wait, but updates are stable within 50us.

# EVERY GET “SEES” A CONSISTENT CUT



Cascade allows you to **get** an object for a specific instant in time.

A time-indexed **get** of multiple objects returns data along a consistent cut.

This allows you to “collect” data into a vector, array or higher dimensional tensor. Our animation was a 3-D tensor: time, plus a 15x15 array of values from power-grid sensors (synchrophasor IoT devices)

# SUMMARY



Systems with imperfect clocks, should integrate causality with their clock, not rely purely on clock timestamps.

This is done with Lamport's "causal" timestamp or a vector timestamp.

- A causal timestamp is just one integer, so many systems use it
- A vector timestamp would cover all cases, but needs one integer *per machine*. So these vectors can be too large for practical use.
- The mechanism is already integrated into Cascade