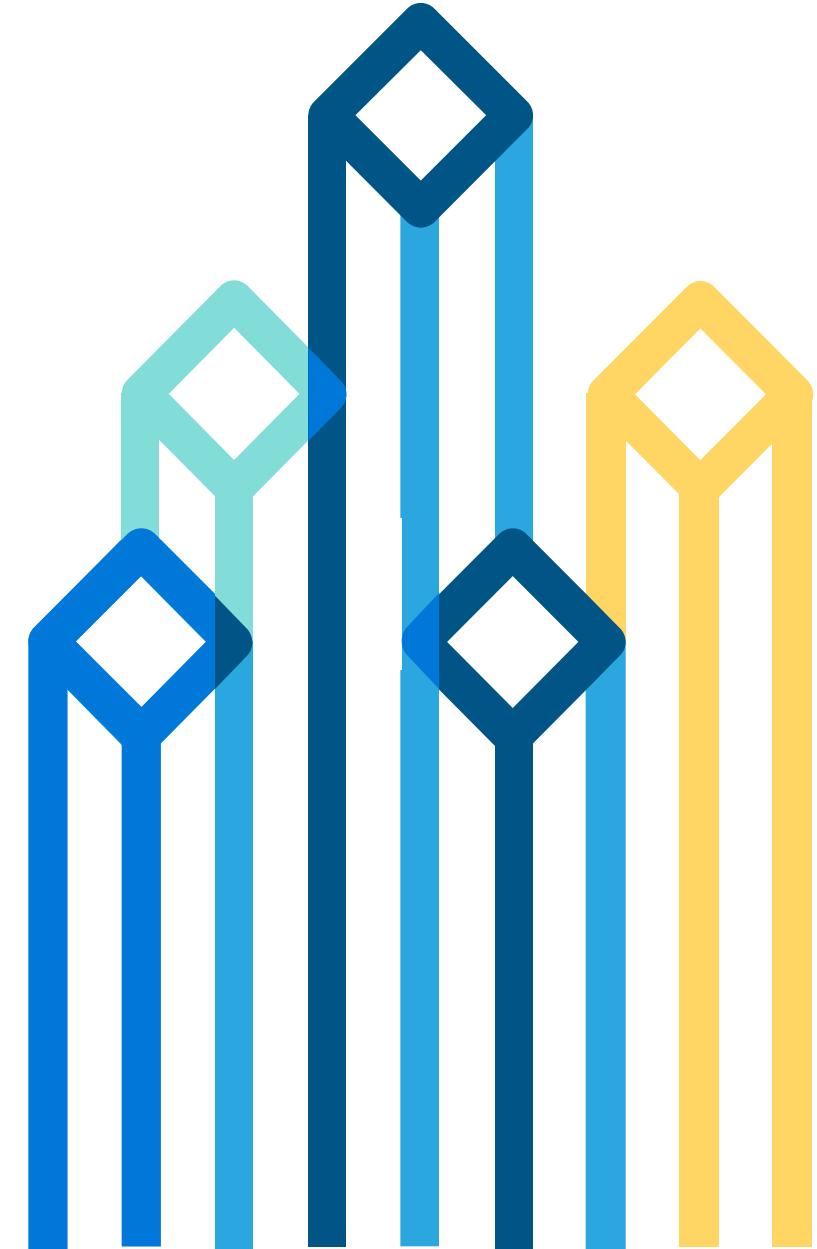


Apache HBase: Overview and Use Cases

Apekshit Sharma



About myself

Apekshit Sharma

- Software Engineer, Cloudera
 - Ex-Software Engineer, Google
- Apache HBase contributor
- Performance improvements,
replication, build infra, etc

Contents

- Motivation
- Apache HBase data model
- Overview of Architecture
- Few usage patterns

Motivation

Bigtable: A Distributed Storage System for Structured Data

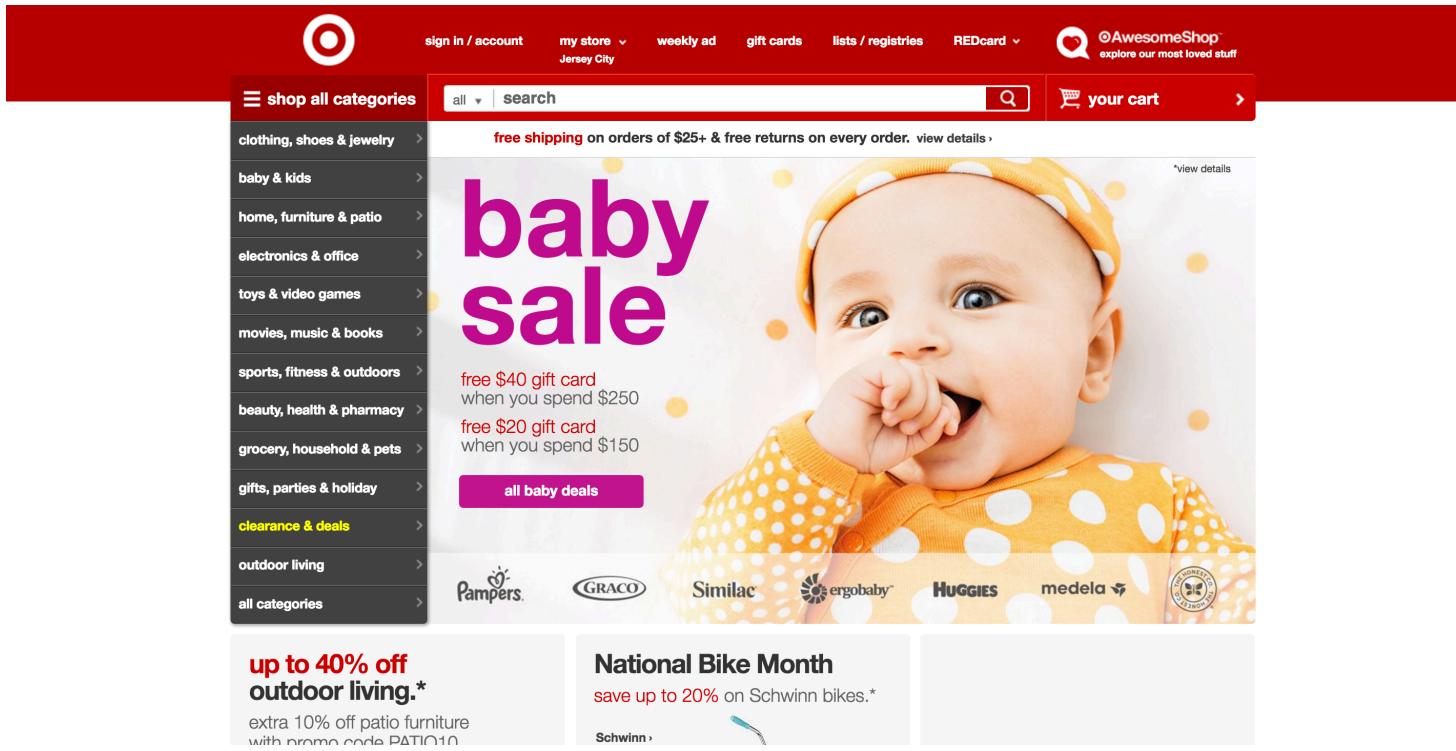
Fay Chang, Jeffrey Dean, Sanjay Ghemawat, Wilson C. Hsieh, Deborah A. Wallach
Mike Burrows, Tushar Chandra, Andrew Fikes, Robert E. Gruber

{fay,jeff,sanjay,wilsonh,kerr,m3b,tushar,fikes,gruber}@google.com

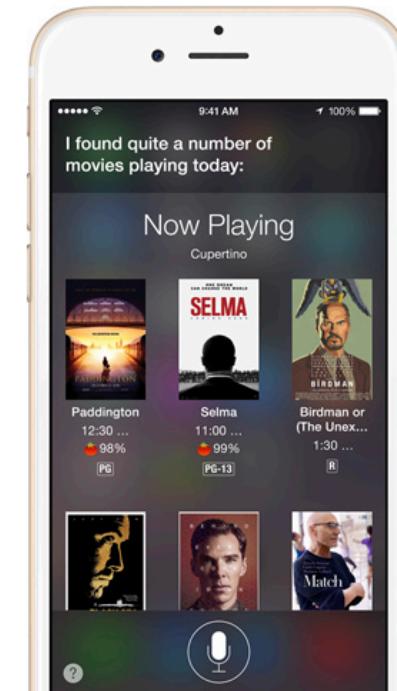
Google, Inc.

Motivation

- What if you're not trying to index the internet?



"What movies are playing today?"





- Open Source
- Horizontally Scalable
- Consistent
- Random access, low latency
- Built on top of HDFS

Data model

Columns

Row key	info:name	info:age	comp:base	comp:stocks
121	'tom'	'28'	'125k'	
145	'bob'	'32'	'110k'	'50' (ts=2012) '100' (ts=2014)

Row keys

Cells

Data model

- Sorted **rows** : supports billions of rows
- **Columns** : Supports millions of columns
- **Cell** : intersection of row and column.
 - Can have multiple values (which are time-stamped)
 - Can be empty. No storage/processing overheads

HBase Architecture

Unique id	Name	price	weight	store1	store2	store3
“1000000”	snickers	\$9.99	4 Oz	Yes	Yes	Yes
“3000000”	almonds	\$9.99	8 Oz	Yes	No	Yes
“8000000”	coke	\$9.99	16 Oz	Yes	Yes	Yes

HBase Architecture

Unique id	Name	price	weight	store1	store2	store3
"1000000"	snickers	\$9.99	4 Oz	Yes	Yes	Yes
"3000000"	almonds	\$9.99	8 Oz	Yes	No	Yes
"8000000"	coke	\$9.99	16 Oz	Yes	Yes	Yes
"4000000"	new	\$34.63	16 Oz	No	Yes	Yes

HBase Architecture

Unique id	Name	price	weight	store1	store2	store3
"1000000"	snickers	\$9.99	4 Oz	Yes	Yes	Yes
"3000000"	almonds	\$9.99	8 Oz	Yes	No	Yes
"8000000"	coke	\$9.99	16 Oz	Yes	Yes	Yes
"4000000"	foo	\$34.63	16 Oz	No	Yes	Yes
"5000000"	bar	\$22.54	16 Oz	Yes	Yes	Yes

HBase Architecture

Unique id	Name	price	weight	store1	store2	store3
"1000000"	snickers	\$9.99	4 Oz	Yes	Yes	Yes
"3000000"	almonds	\$9.99	8 Oz	Yes	No	Yes
"8000000"	coke	\$9.99	16 Oz	Yes	Yes	Yes
"4000000"	foo	\$34.63	16 Oz	No	Yes	Yes
"5000000"	bar	\$22.54	16 Oz	Yes	Yes	Yes
"9000000"	new1	\$2.5	16 Oz	Yes	Yes	Yes
"7000000"	new2	\$6.4	16 Oz	Yes	Yes	Yes
"2000000"	new3	\$6.4	16 Oz	Yes	Yes	Yes

HBase Architecture | Regions

["", "5000000")

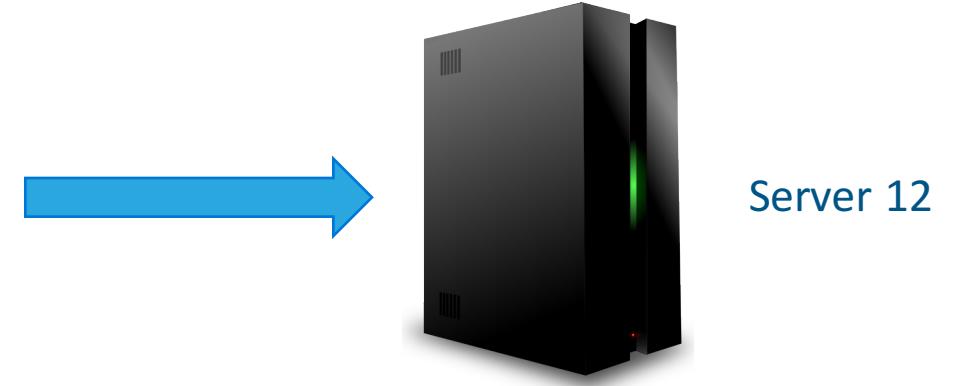
Row Key	Name	brand	price	weight	store1	store2	store3
"1000000"	snickers	xxx	\$9.99	4 Oz	Yes	Yes	Yes
"2000000"	new3	xxx	\$6.4	16 Oz	Yes	Yes	Yes
"3000000"	almonds	xxx	\$9.99	8 Oz	Yes	No	Yes
"4000000"	foo	xxx	\$34.63	16 Oz	No	Yes	Yes

["5000000", "")

Row Key	Name	brand	price	weight	store1	store2	store3
"5000000"	bar	xxx	\$22.54	16 Oz	Yes	Yes	Yes
"7000000"	new2	xxx	\$6.4	16 Oz	Yes	Yes	Yes
"8000000"	coke	xxx	\$9.99	16 Oz	Yes	Yes	Yes
"9000000"	new1	xxx	\$2.5	16 Oz	Yes	Yes	Yes

HBase Architecture | RegionServer

Row Key	Name	price	weight
"1000000"	snickers	\$9.99	4 Oz
"2000000"	new3	\$6.4	16 Oz
"3000000"	almonds	\$9.99	8 Oz
"4000000"	foo	\$34.63	16 Oz



Row Key	Name	price	weight
"5000000"	bar	\$22.54	16 Oz
"7000000"	new2	\$6.4	16 Oz
"8000000"	coke	\$9.99	16 Oz
"9000000"	new1	\$2.5	16 Oz



HBase Architecture | Regions

- Horizontal split of tables
- Regions are served by RegionServers
- Automatically (based on configurations). Can be done manually too.

HBase Architecture

Row Key	Name	price	weight	store1	store2	store3
"1000000"	snickers	\$9.99	4 Oz	Yes	Yes	Yes
"2000000"	new3	\$6.4	16 Oz	Yes	Yes	Yes
"3000000"	almonds	\$9.99	8 Oz	Yes	No	Yes
"4000000"	foo	\$34.63	16 Oz	No	Yes	Yes
"5000000"	bar	\$22.54	16 Oz	Yes	Yes	Yes
"7000000"	new2	\$6.4	16 Oz	Yes	Yes	Yes
"8000000"	coke	\$9.99	16 Oz	Yes	Yes	Yes
"9000000"	new1	\$2.5	16 Oz	Yes	Yes	Yes

HBase Architecture | Column family

Row Key	info: name	info: price	info: weight	availability: store1	availability: store2	availability: store3
"1000000"	snickers	\$9.99	4 Oz	Yes	Yes	Yes
"2000000"	new3	\$6.4	16 Oz	Yes	Yes	Yes
"3000000"	almonds	\$9.99	8 Oz	Yes	No	Yes
"4000000"	foo	\$34.63	16 Oz	No	Yes	Yes
"5000000"	bar	\$22.54	16 Oz	Yes	Yes	Yes
"7000000"	new2	\$6.4	16 Oz	Yes	Yes	Yes
"8000000"	coke	\$9.99	16 Oz	Yes	Yes	Yes
"9000000"	new1	\$2.5	16 Oz	Yes	Yes	Yes

HBase Architecture | Column family

- You can also see it as vertical splits
- Data stored in separate files
- Tune performance
 - In-memory
 - Compression
 - Version retention policies
 - Cache priority
- Needs to be specified by the user

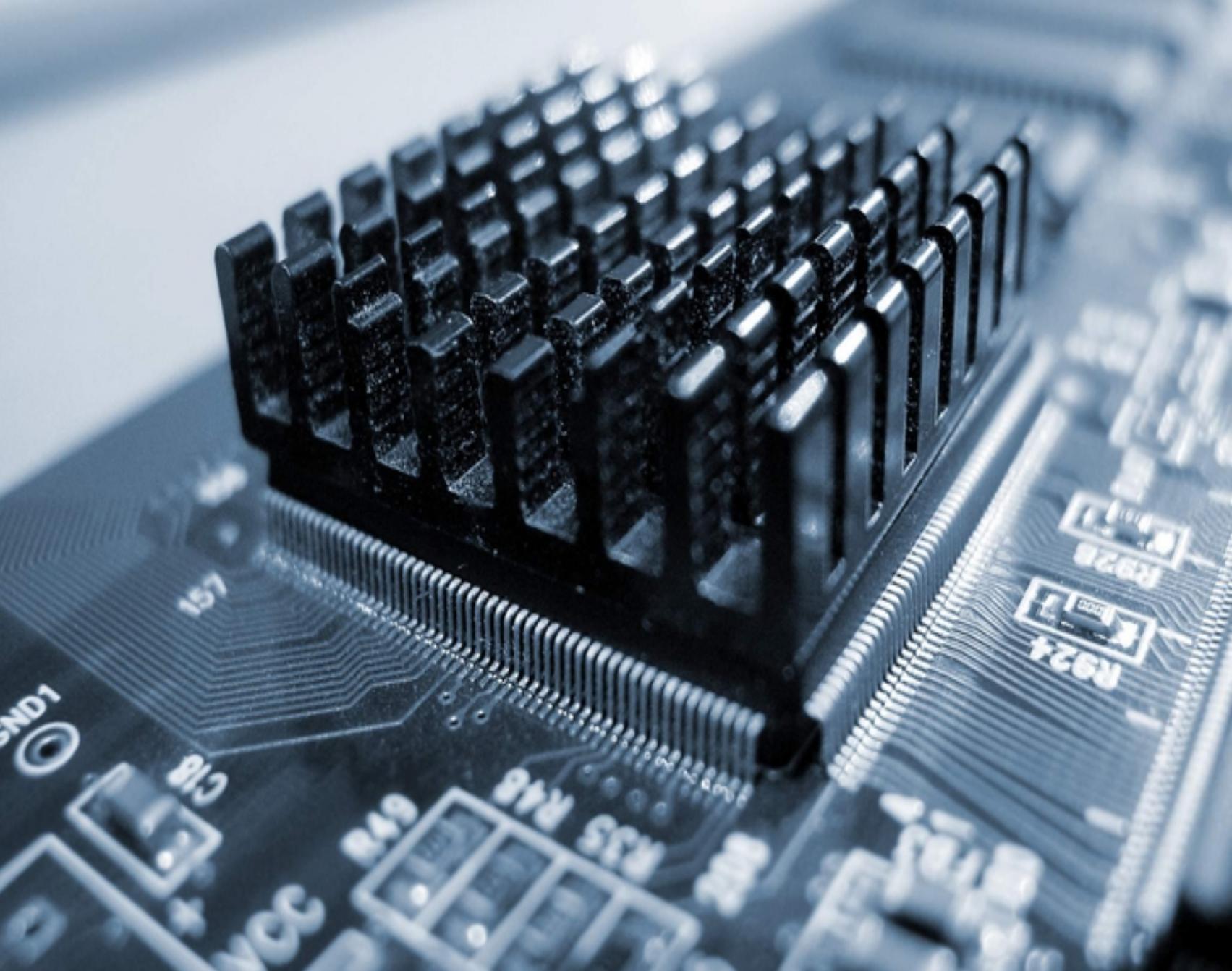
HBase Architecture

Column Family				Region
	info: name	info: price	info: weight	Column Family
"1000000"	snickers	\$9.99	4 Oz	
"2000000"	new3	\$6.4	16 Oz	
"3000000"	almonds	\$9.99	8 Oz	

	available: store1	available: store2	available: store3
"1000000"	Yes	Yes	Yes
"2000000"	Yes	Yes	Yes
"3000000"	Yes	No	Yes

Enough of Architecture

Lets move to
use patterns



Apache HBase “Nascar” Slide



Apache HBase “Nascar” Slide



Apache HBase “Nascar” Slide



Apache HBase “Nascar” Slide





What have we learned from
so many users?



There are some patterns which repeat often.

Just like a Lego block, maybe you can fit one directly in your system!



Know your ...

Data

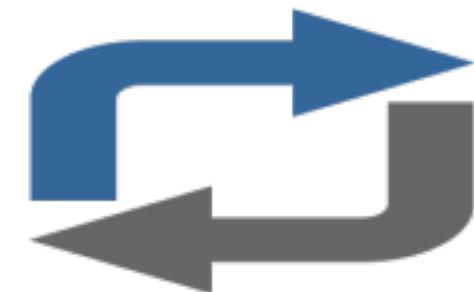
- Entity Data
- Time-centric Event Data



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How it goes in and out

- Real-time vs Batch
- Random vs Sequential



Know your data ...

There are primarily two kinds of big data workloads. They have different storage requirements.

- Entity centric data
- Time centric event data



Entity centric data



Users



Accounts



Location



Clicks and
Metrics

- Scales up with # of entities
- Billions of distinct entities

Time centric event data



Stock Ticker Data

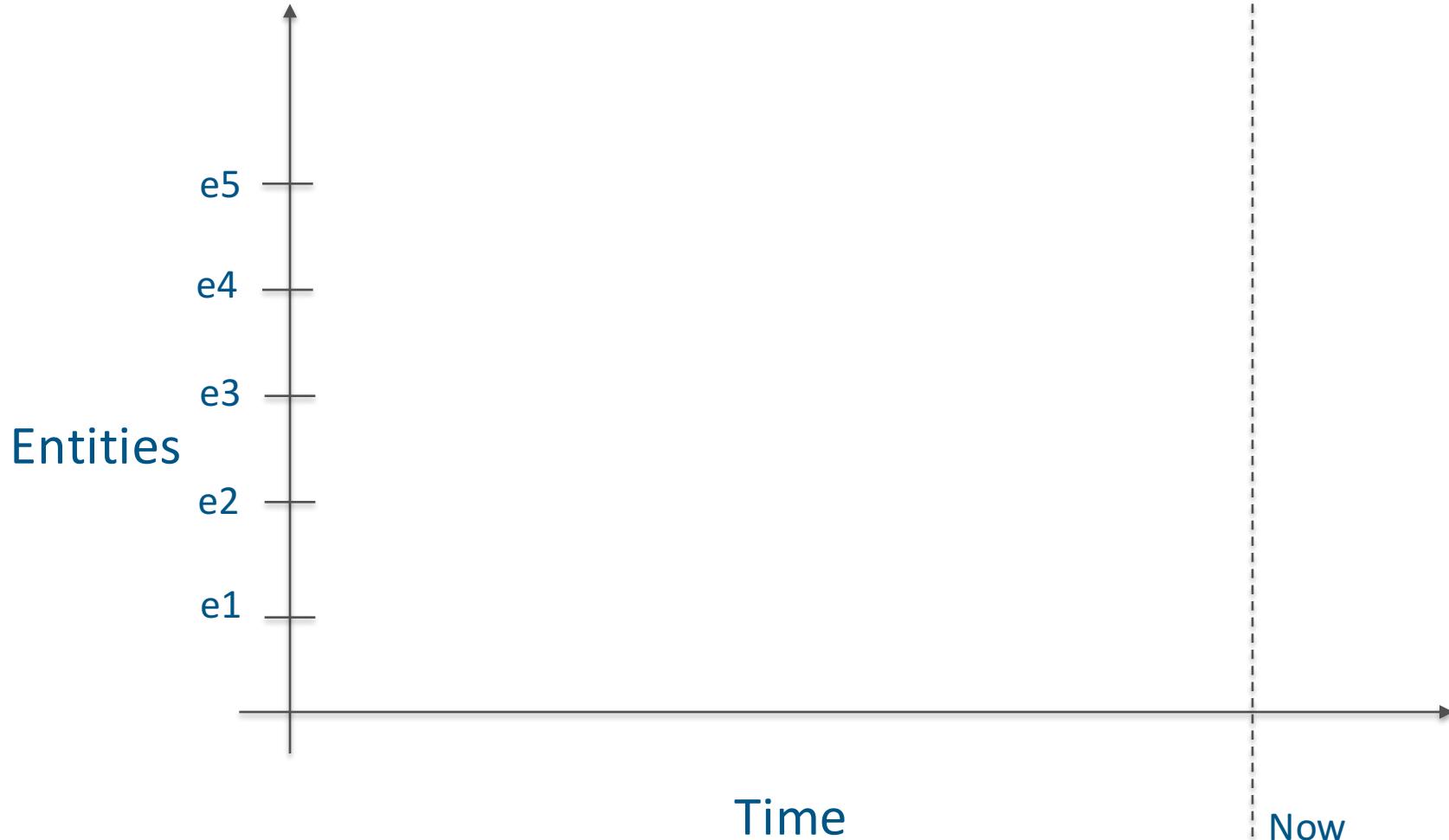


Monitoring applications

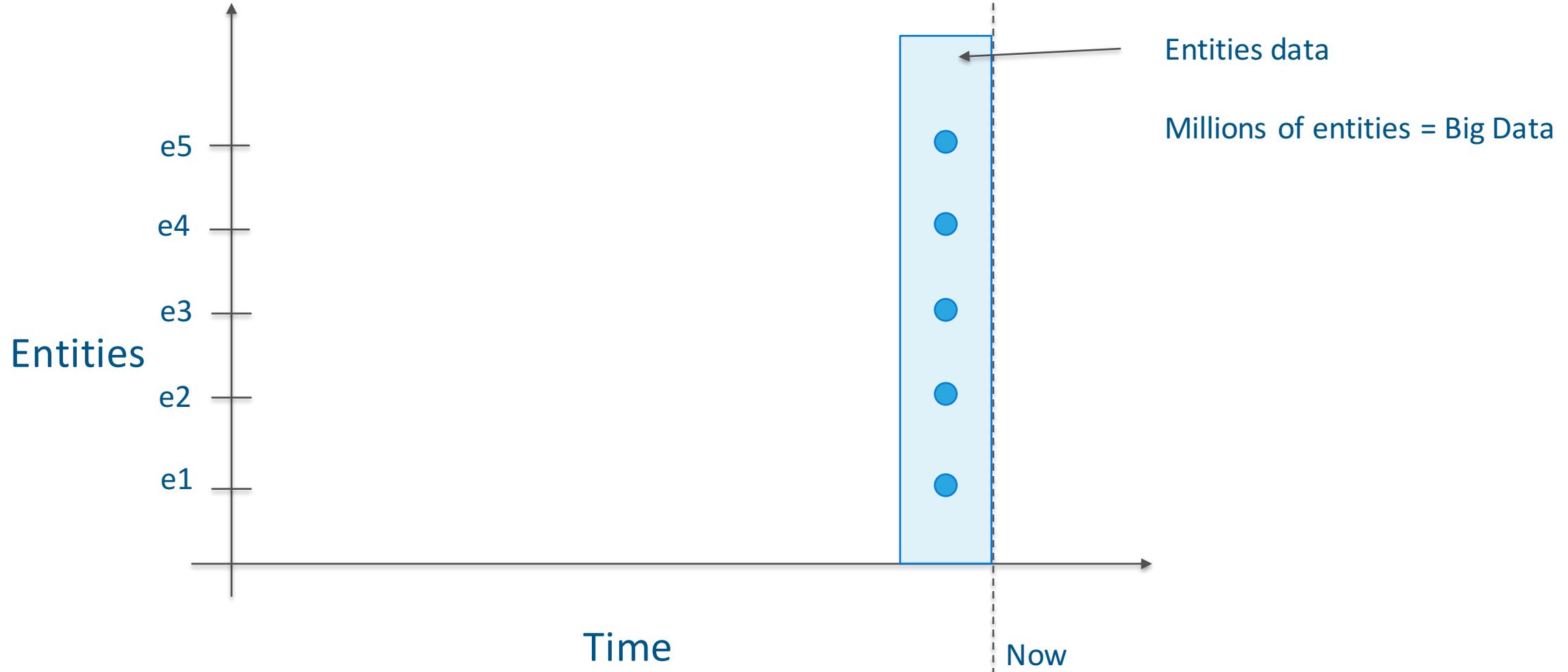


Periodic Sensor Data

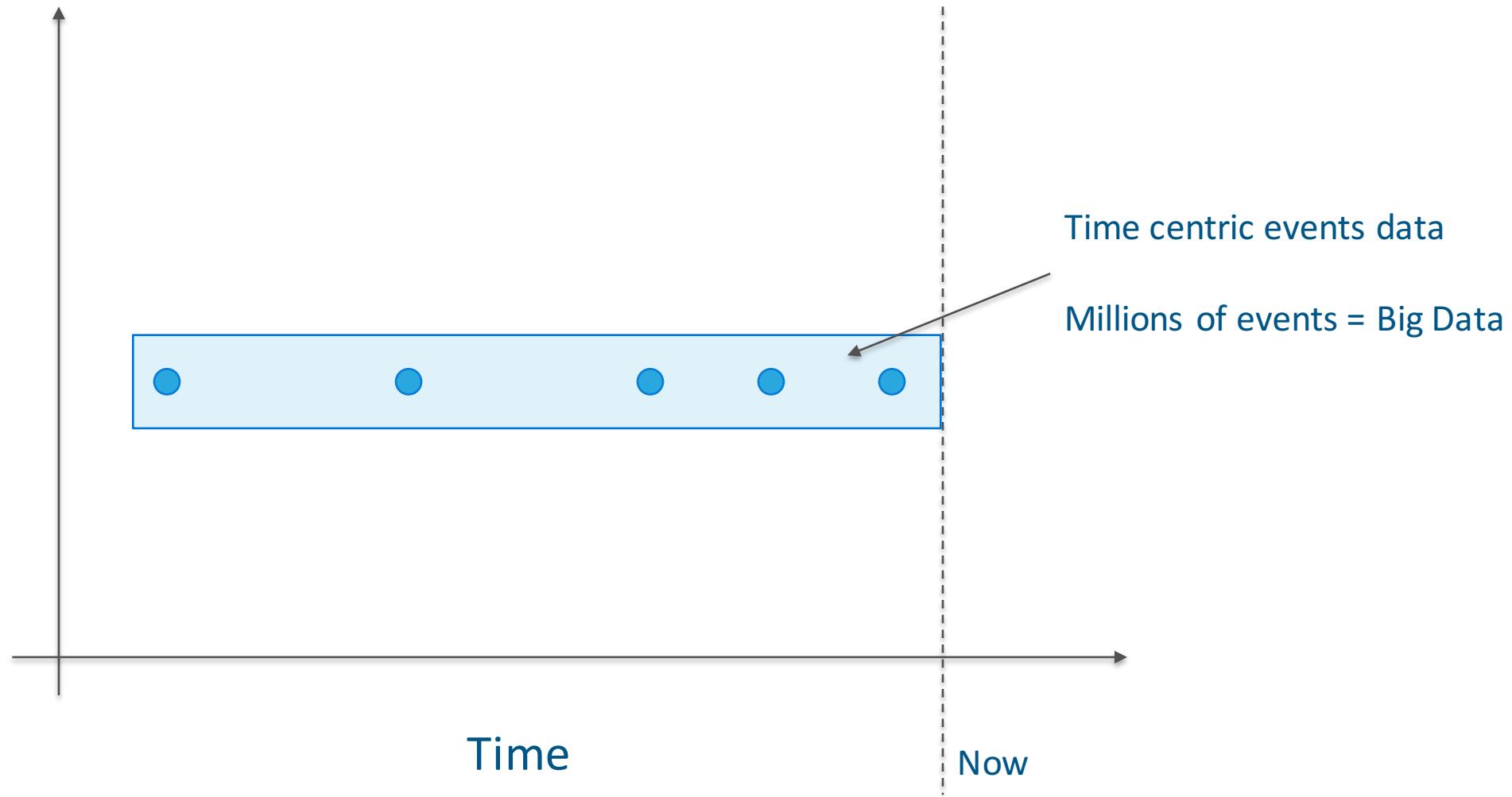
- Time-series data points over a period
- Scales up due to finer grained intervals, retention policies, and the passage of time



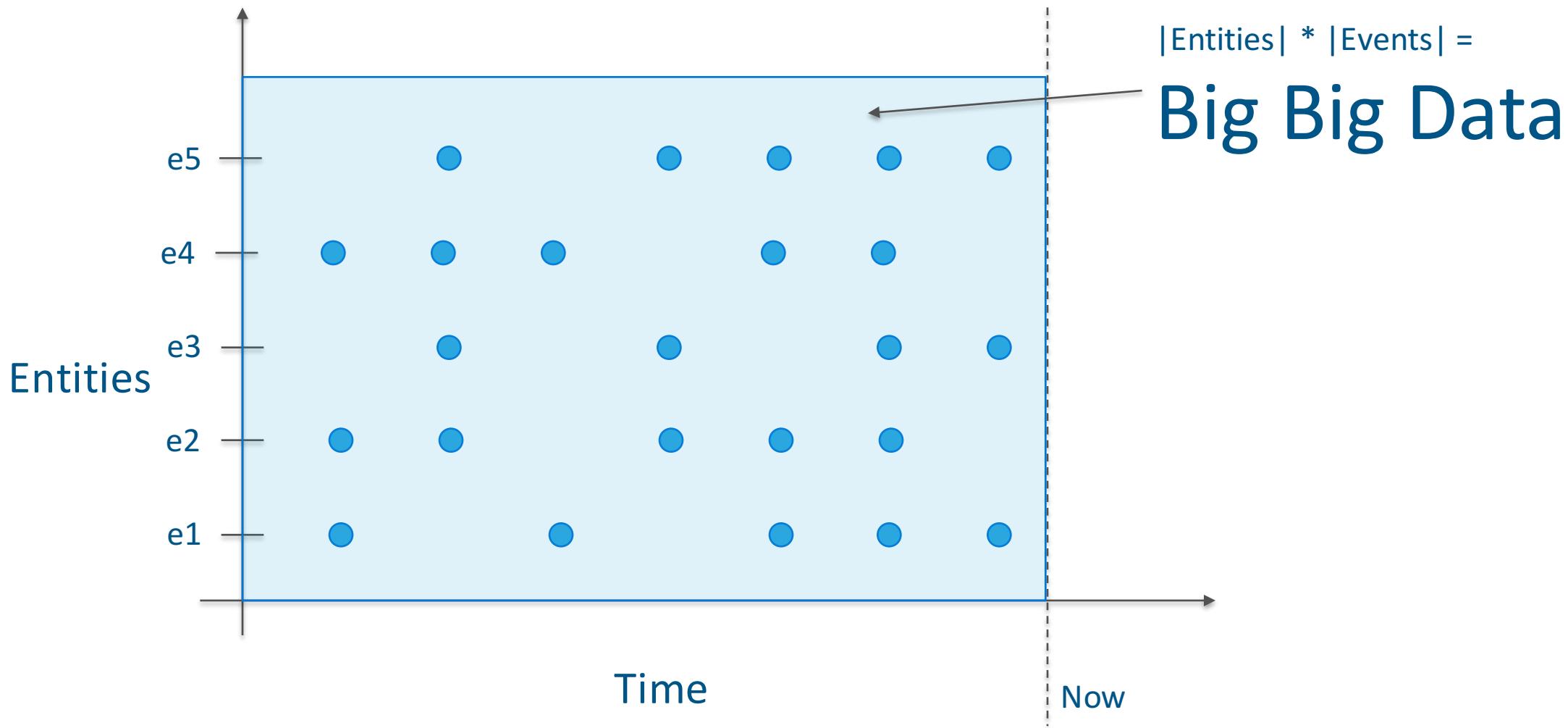
Entities data



Time-centric events data



Time-centric events about Entities



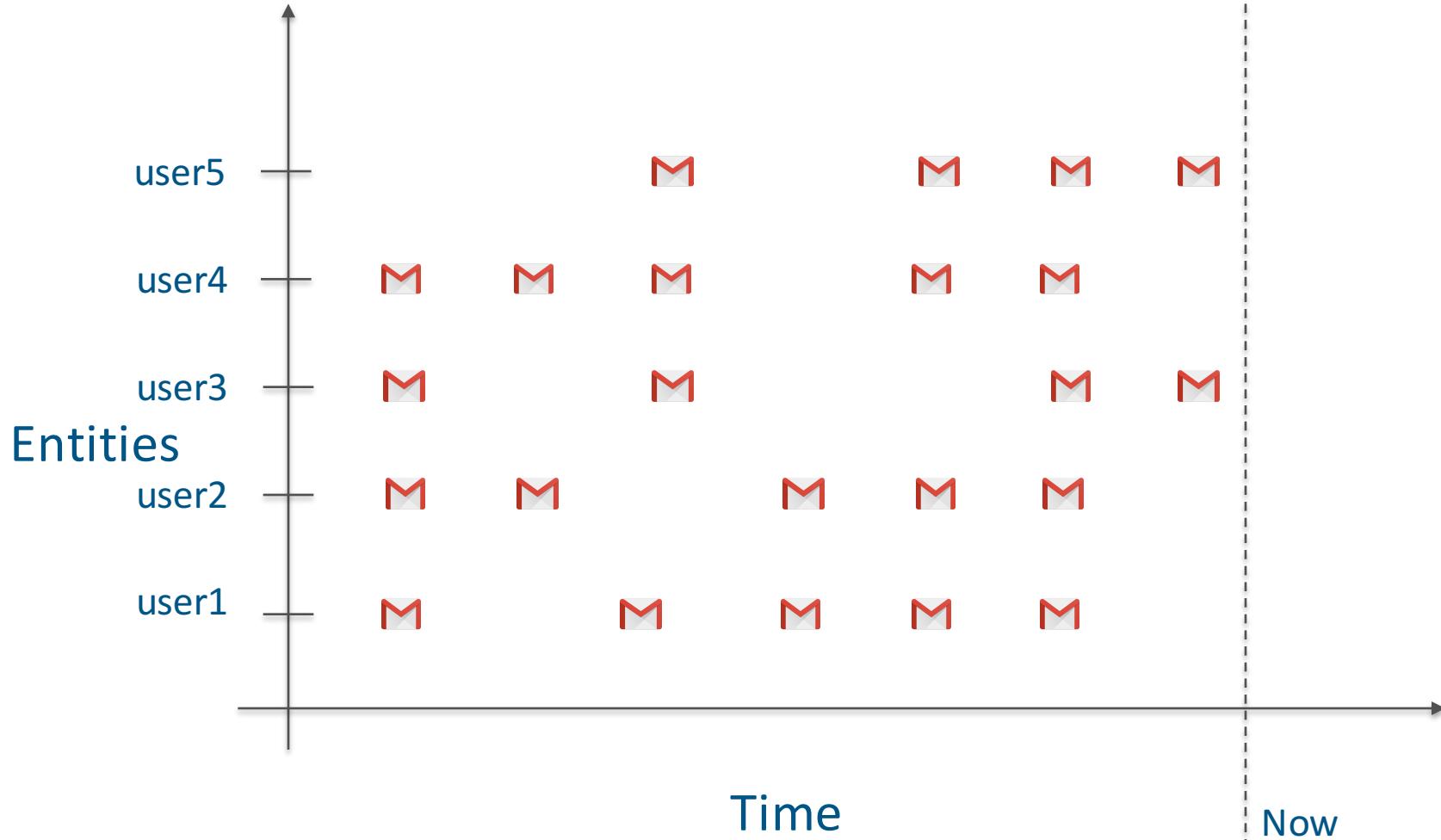
What questions do you ask?

- Do you focus in on entity first?

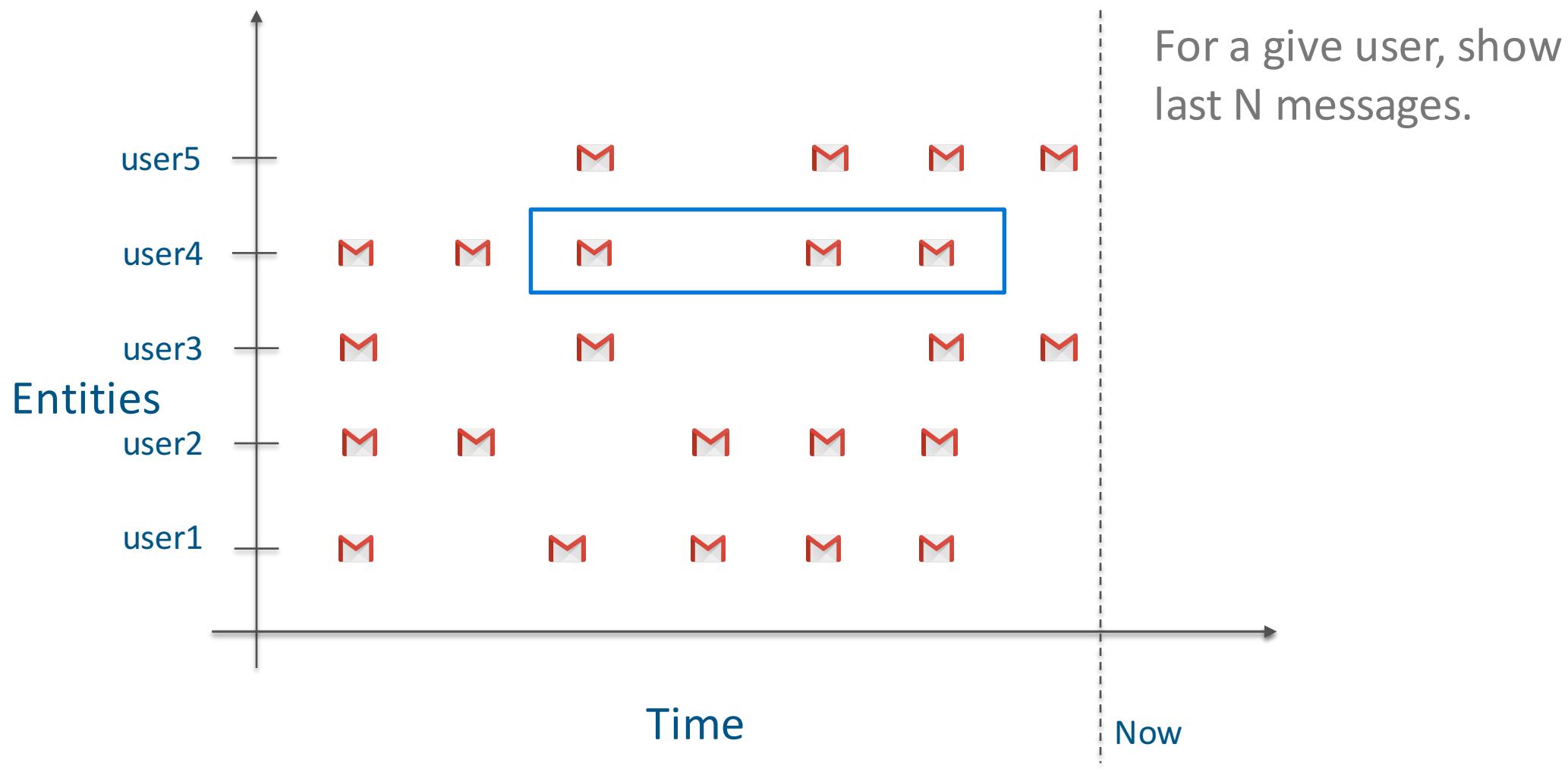
OR

- Do you focus in on time ranges first?
- Your answer will help you determine where and how to store your data.

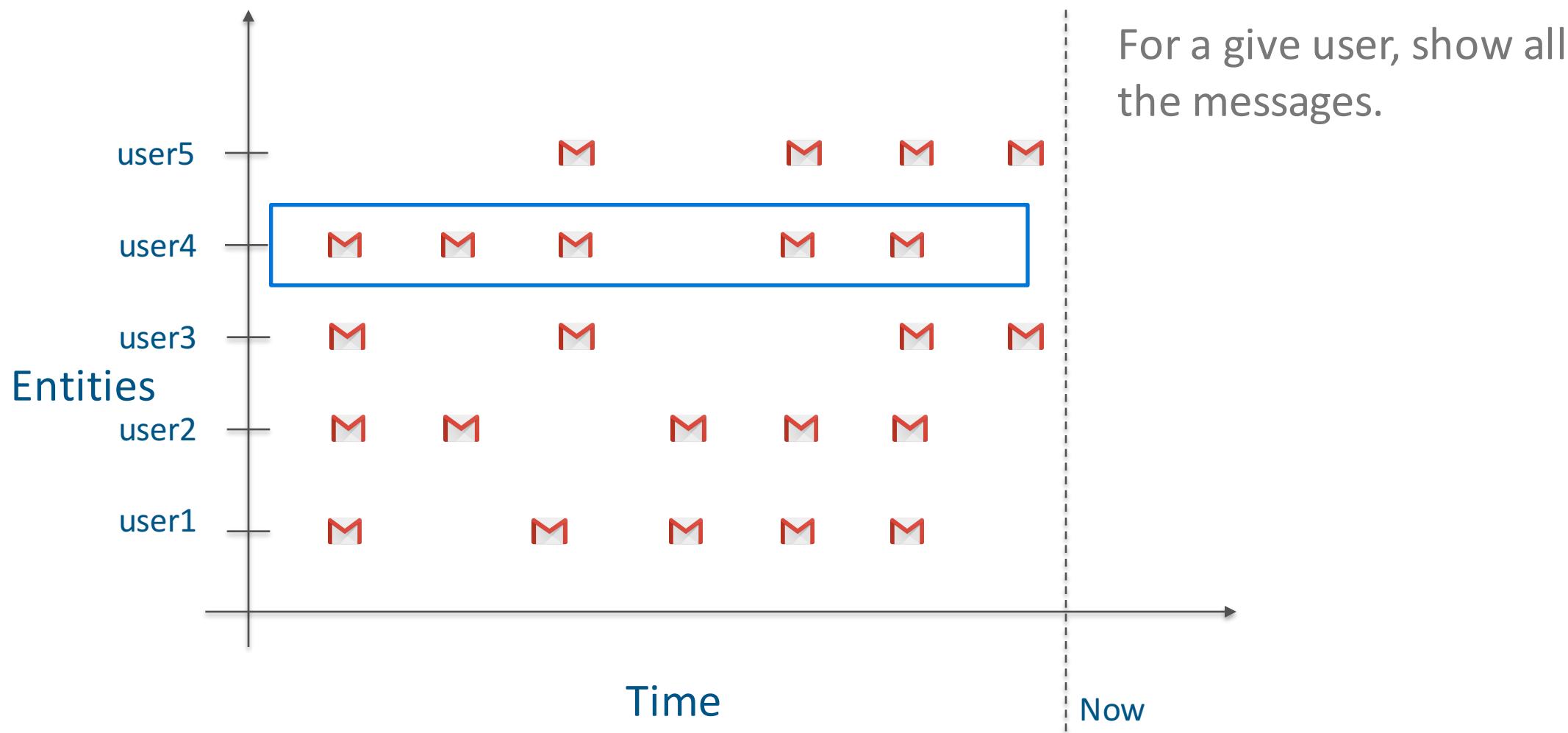
Let's say you start your own e-mail service....



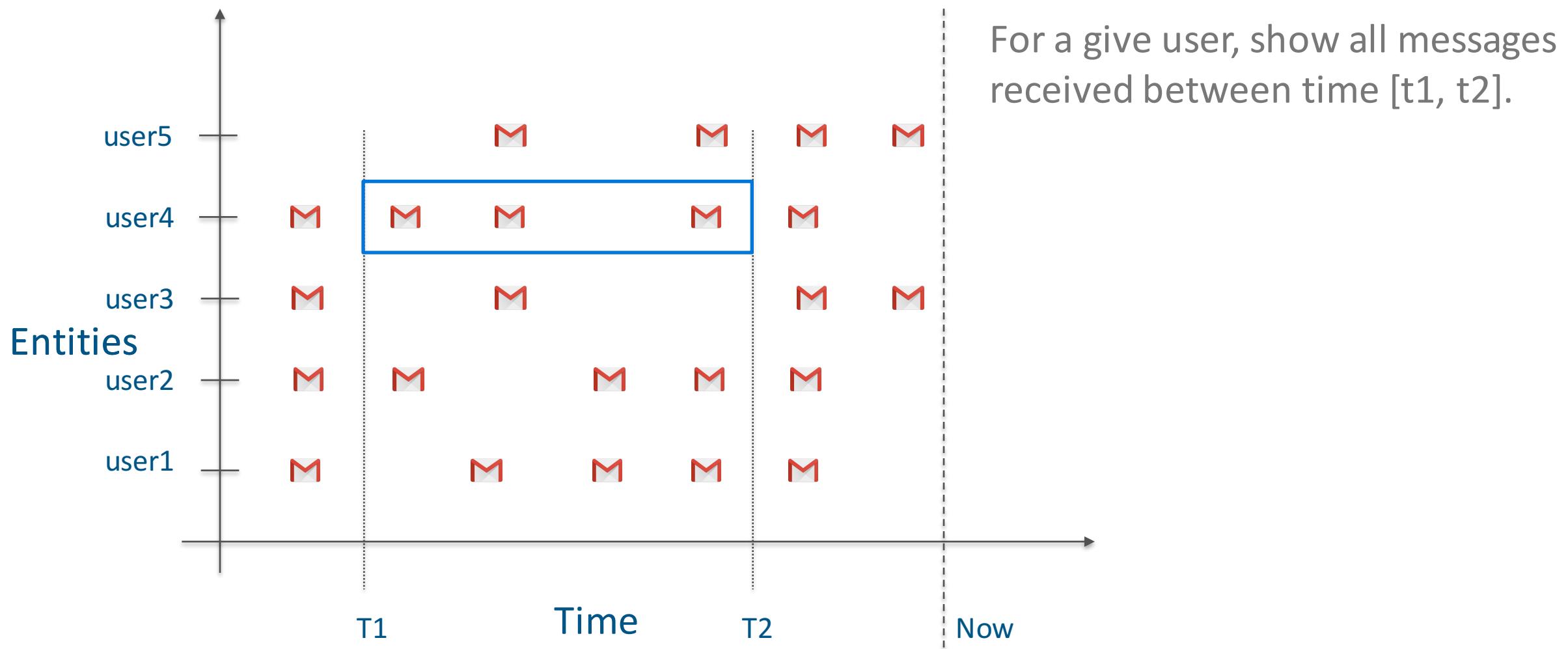
Entity first questions...



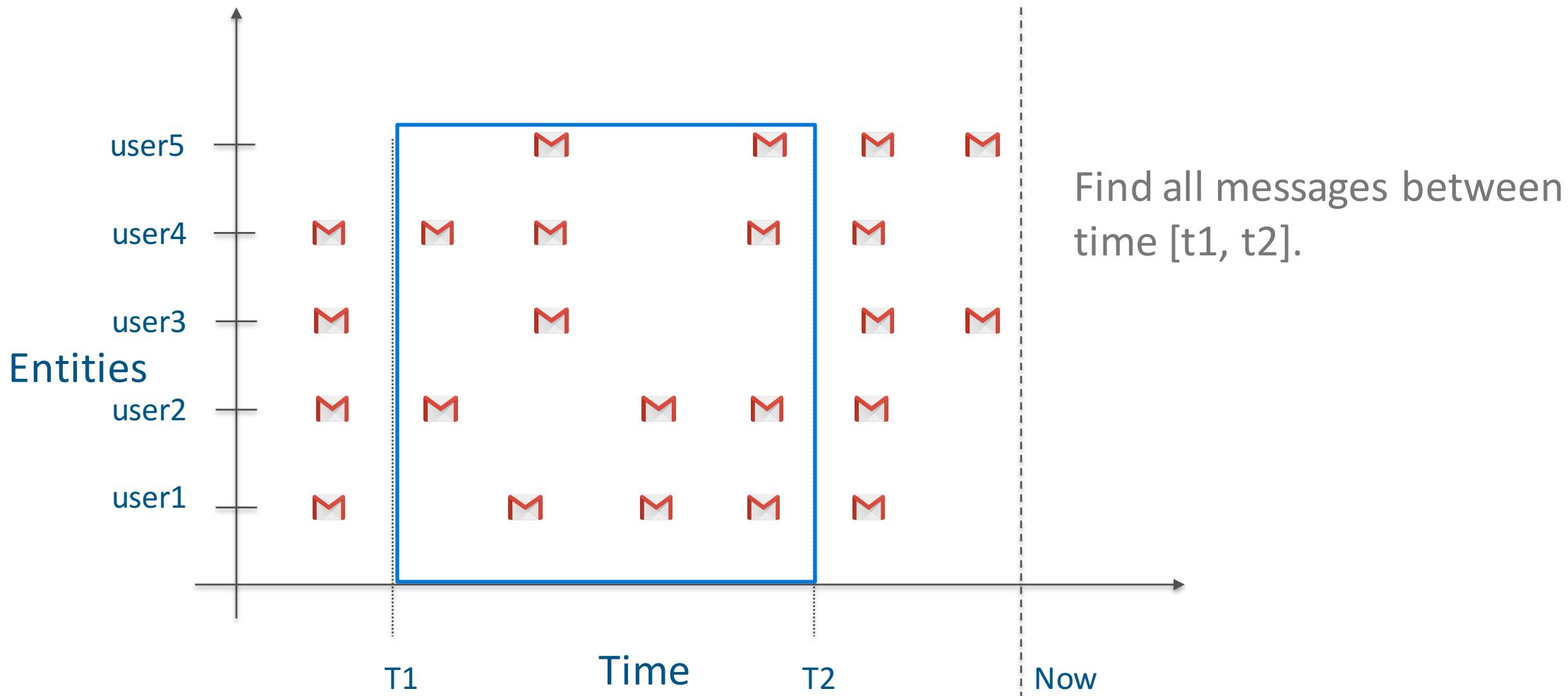
Entity first questions...



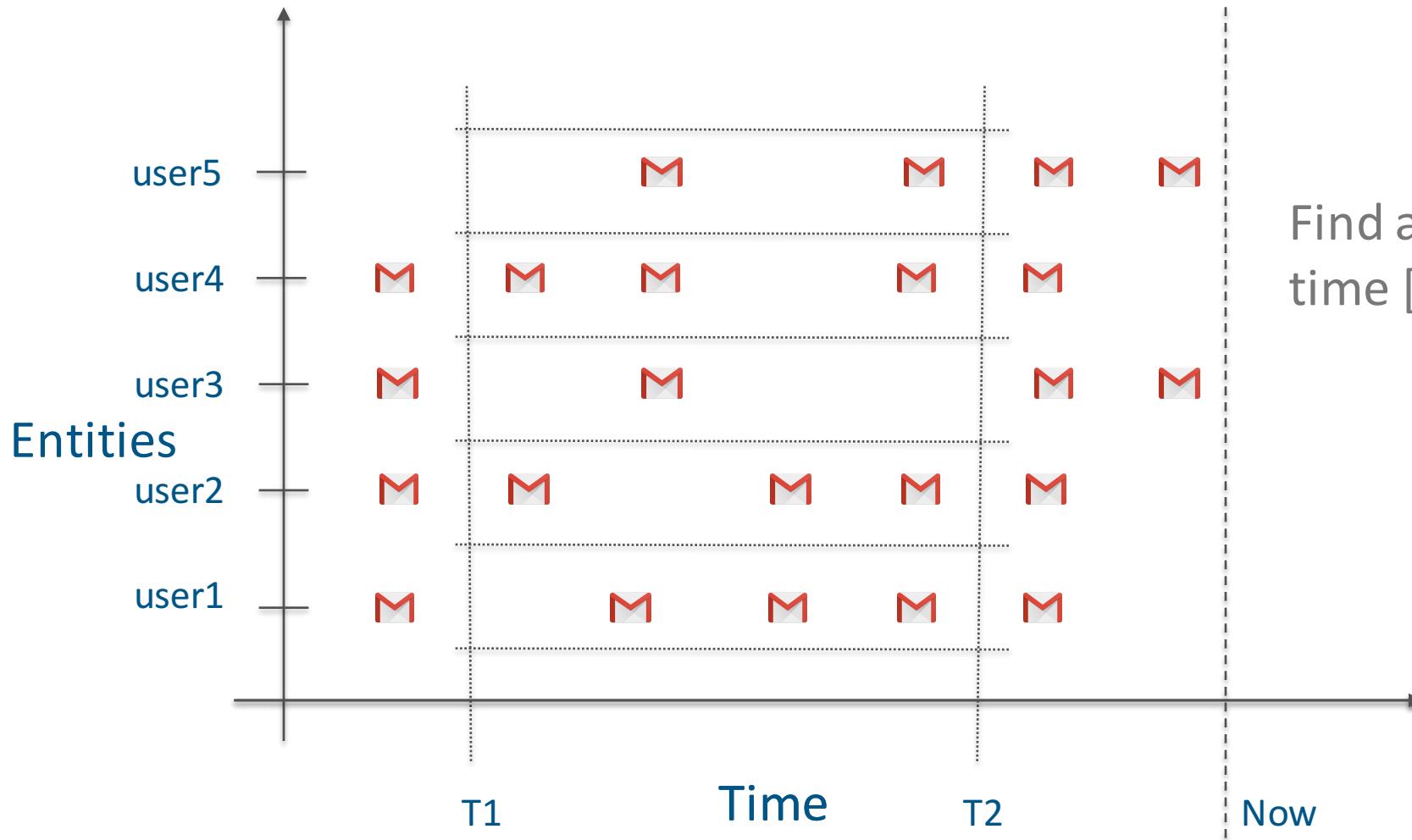
Entity first questions...



Time centric event first questions...

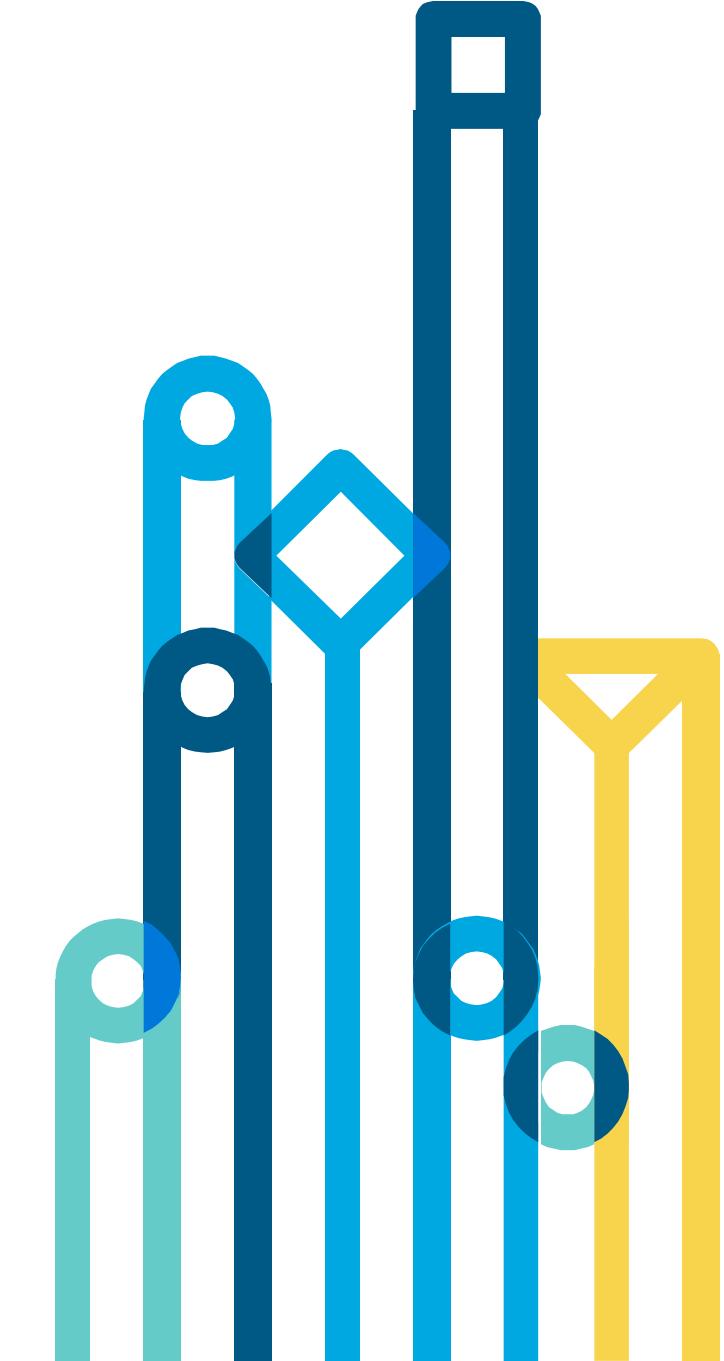


Time centric event first questions...



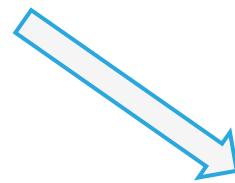
Find all messages between
time $[t_1, t_2]$ for all users.

How does the data get in and out
of HBase?

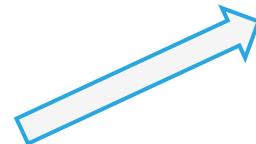


Getting data in...

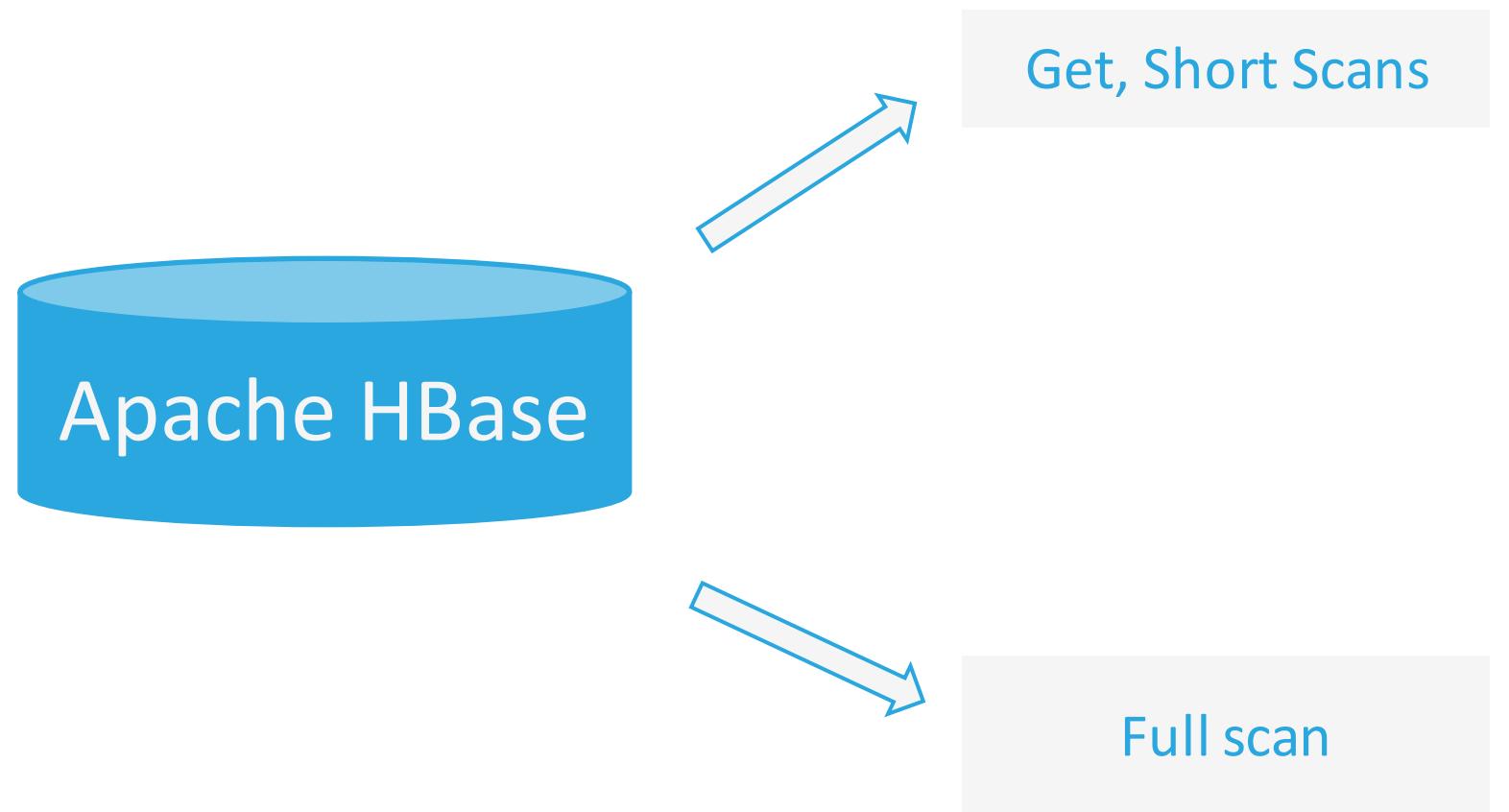
Put, Incr, Append



Bulk Import



Getting data out...

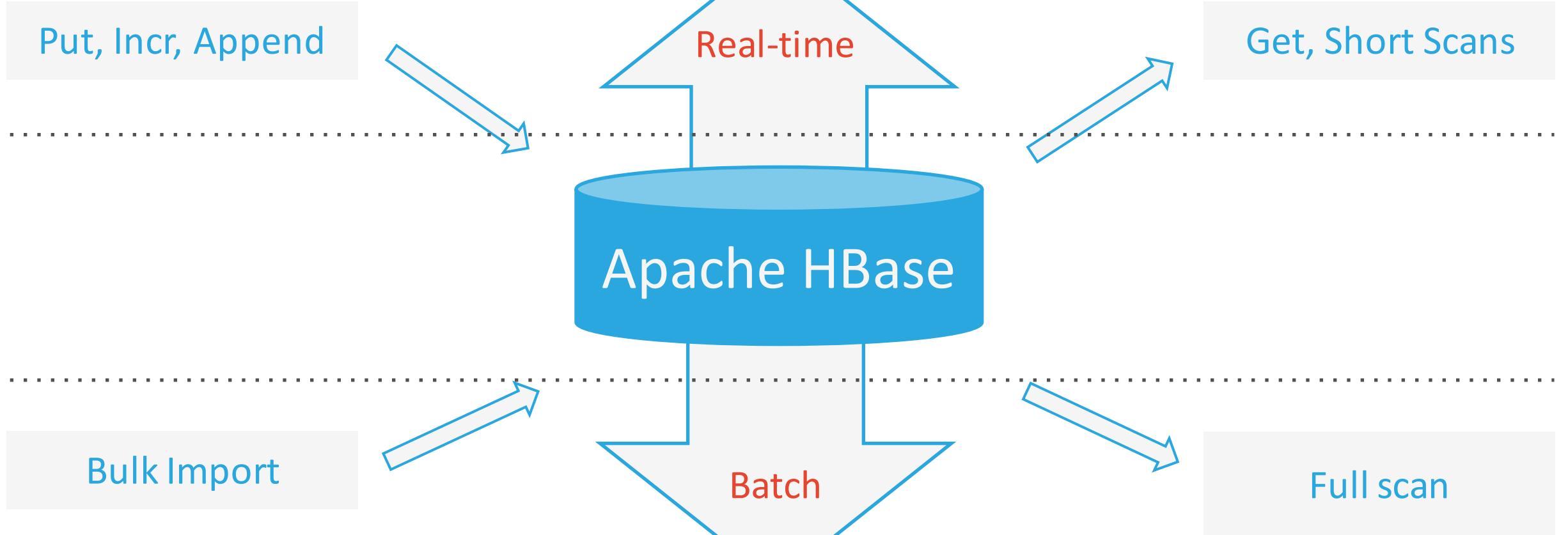


So, what's the best way?

Depends on your use case

Bottom-line: Disk I/O takes times.

- Limited disk read-write heads in a cluster
- Use the I/O bandwidth of your cluster efficiently

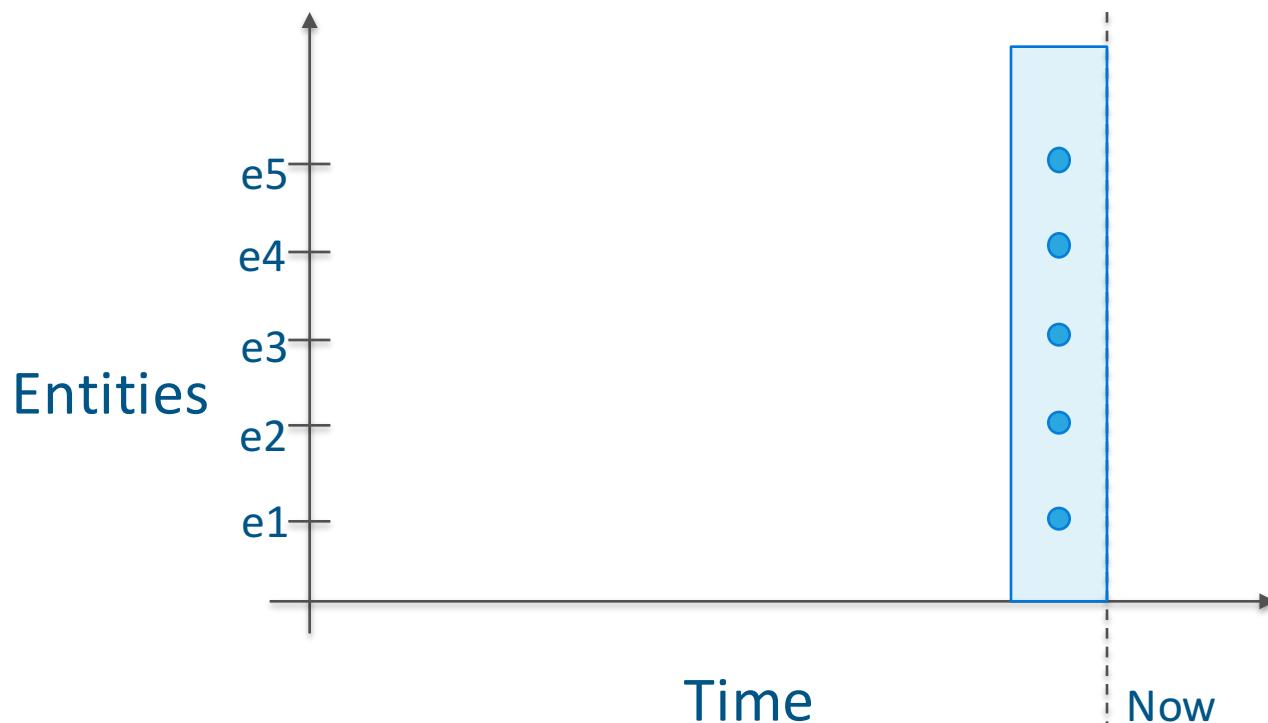


Let's dive into use case ...



Simple Entities

- Purely entity data, no relation between entities
 - Often from many different sources
 - Could be a well-done de-normalized RDBMS



Simple Entities : Schema

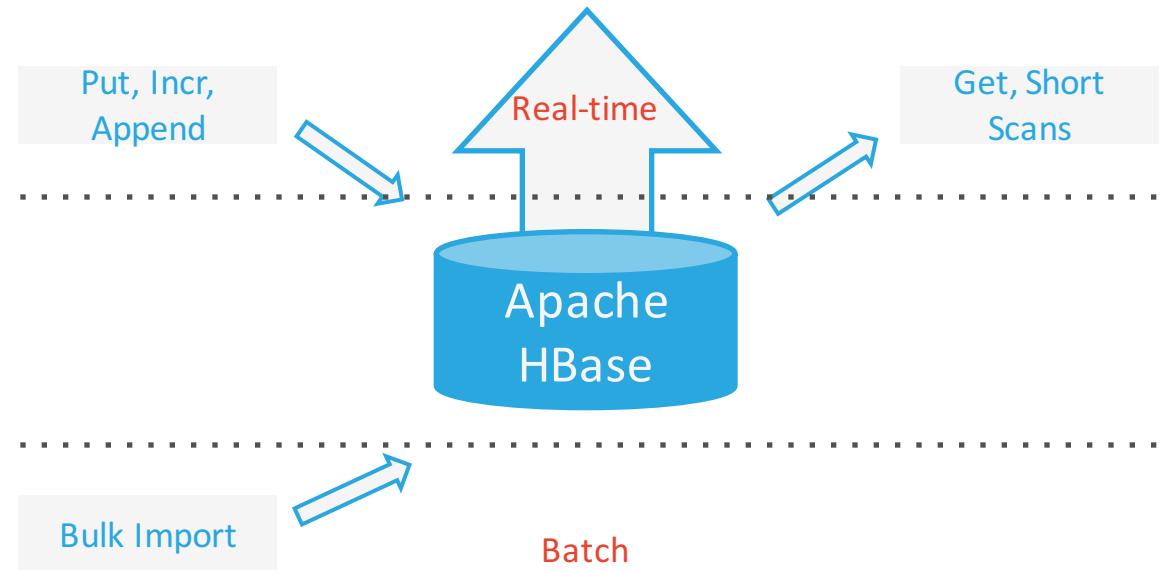
- One row per entity
- Row key => entity ID, or hash of entity ID
- Column => Property / field, possibly timestamp

Simple Entities : Example

OCLC : Online Computer Library Center

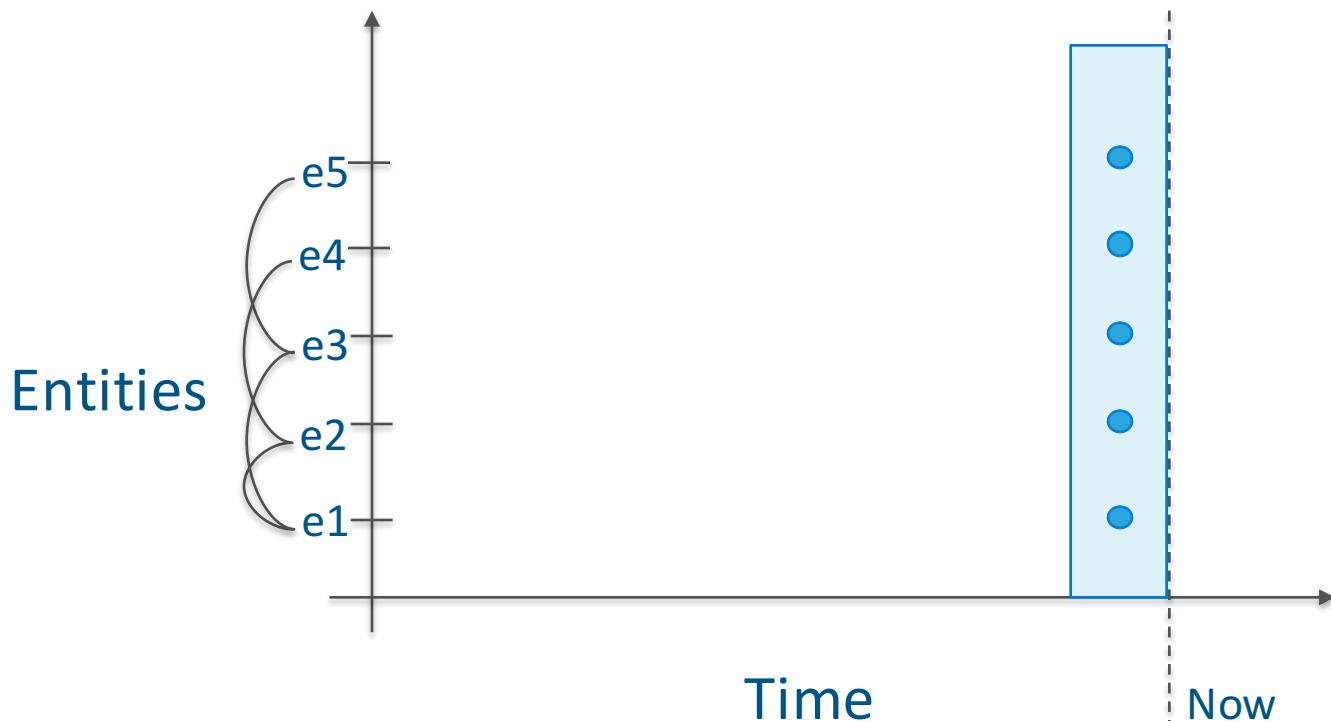
Workloads:

- Lookup books → Real time read
- Add new book one at a time, update information about existing books, issue books → Real-time write
- New library joins the group, import its data → Batch write



Linked Entities (Graph Data)

- Entity are linked to form a graph



Linked Entities (Graph Data) : Schema

- One row per Node (Entity)
- Row key => Node ID (Entity ID)
- One Column Family to store edges =>
“Relationship:OtherNodeID”
- Value => Meta data about relationship
- Second column family to store other information about node

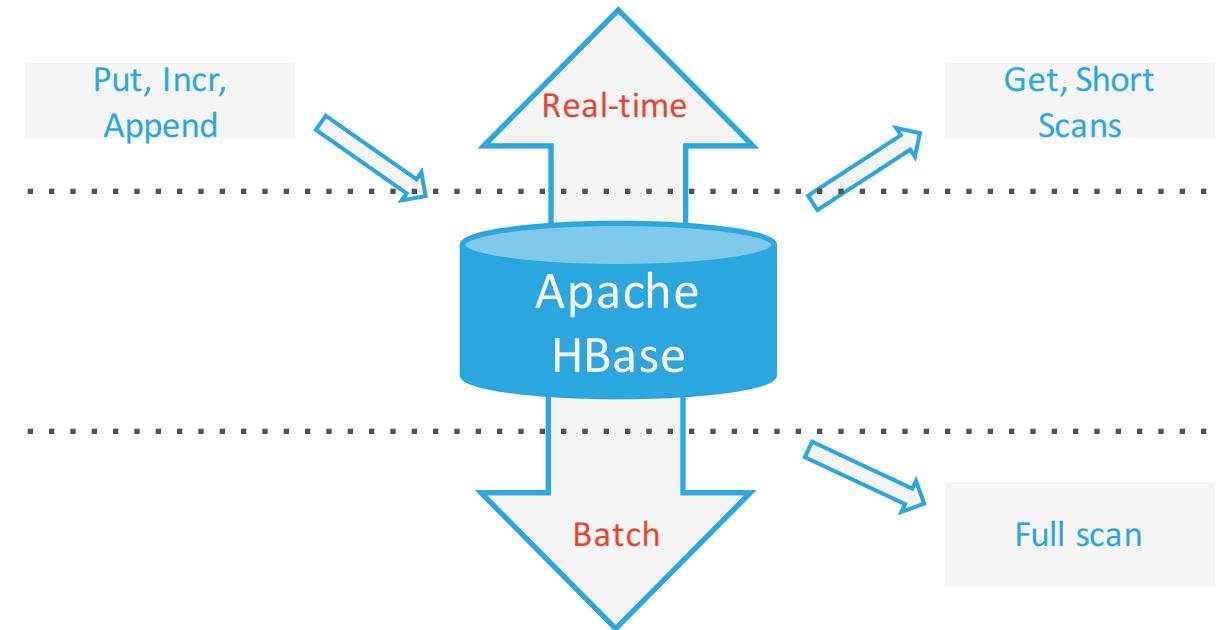
Linked Entities (Graph Data) : Example

Social Network (Facebook)



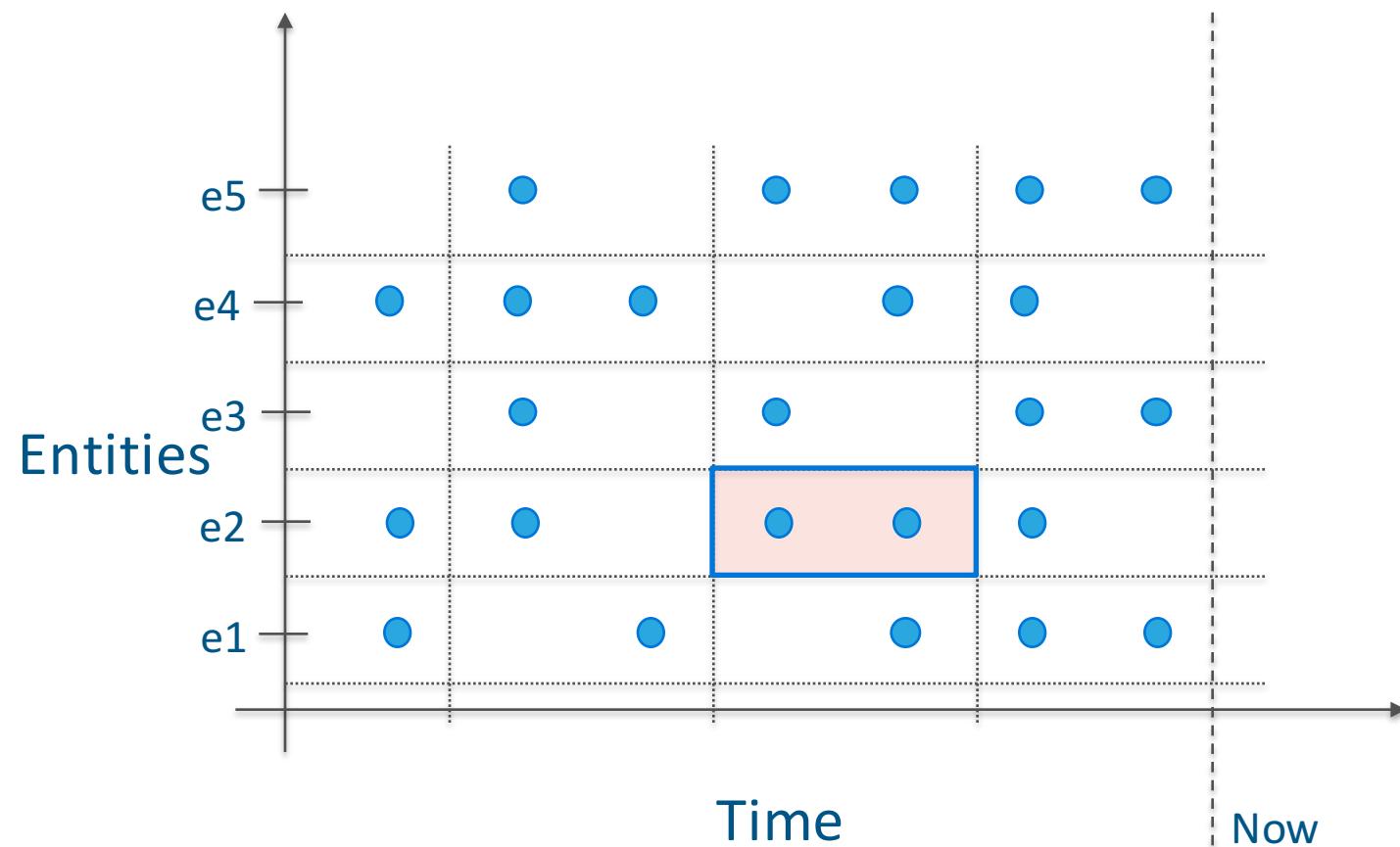
Workloads:

- Get any info about a user → Real time read
- Update any info about a user → Real time write
- Limited graph analysis (based on immediate friends) → Batch read



Time-coupled entities

- Events about entities
- Focus on entities first



Time-coupled entities : Schema

- Row = Entity's events in a time slice
- Row key = Entity ID + (time / k)
- Column Qualifier = timestamp

Time-coupled entities: Example

Messaging service



Primary Workload

- Sending a message, update metadata (read, star, move, delete) →

Real-time write

- Reading a message, get last N messages → Real-time read

HBase is great!

But there are some use cases which are better off without it...

Current HBase weak spots

- HBase architecture can handle a lot
 - Engineering tradeoffs optimize for some use cases
 - HBase can still do things it is not optimal for
 - Other systems are fundamentally more efficient for some workloads

A not so good use case: Large Blob Store

- Saving large objects >50 MB per cell
- Examples
 - Raw video storage in HBase
- Problems:
 - **Write amplification** when re-optimizing data for read (compactions on large unchanging data)
 - New: Medium Object (MOB) supported (lots of 100KB-10MB cells)

Another not good use case: Analytic archive

- Store data chronologically, time as primary index
- Schema
 - Row key: timestamp
 - Monotonically increasing row key
 - Column qualifiers: properties with data or counters
- Example
 - Machine logs organized **by timestamp** (causes write hot-spotting)

That's all folks!





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Questions ?

Sources

- A Survey of HBase Application Archetypes
 - Lars George, Jon Hsieh
 - <http://www.slideshare.net/HBaseCon/case-studies-session-7>
- OpenTSDB 2.0
 - Benoit Sigoure, Chris Larsen
 - <http://www.slideshare.net/HBaseCon/ecosystem-session-6>
- Hadoop and HBase: Motivations, Use cases and Trade-offs
 - Jon Hsieh
- Phoenix
 - <https://phoenix.apache.org>