

#### Timeseries data at scale for the masses

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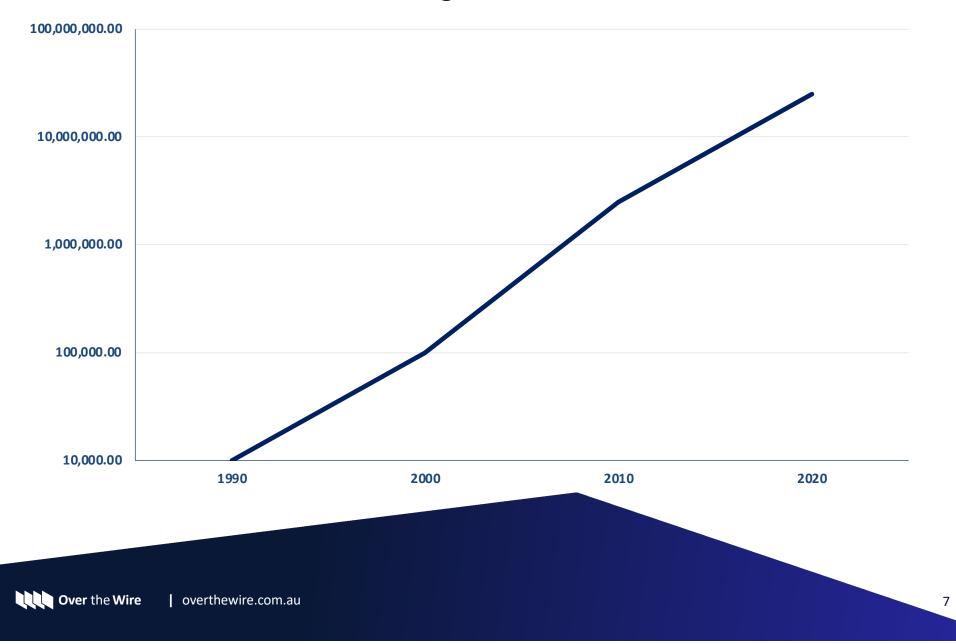
# **Timeseries Data**

Growth over the ages

1990	University Network 15 Hosts Performance metrics => 10k records per day
2000	ISP Network Dialup user base Radius sessions => 100k records per day
2010	Hosting Infrastructure Infrastructure metrics VM metrics => 2.5m records per day
Now	Telco Network Packet loss & Latency => 25m records per day per pod

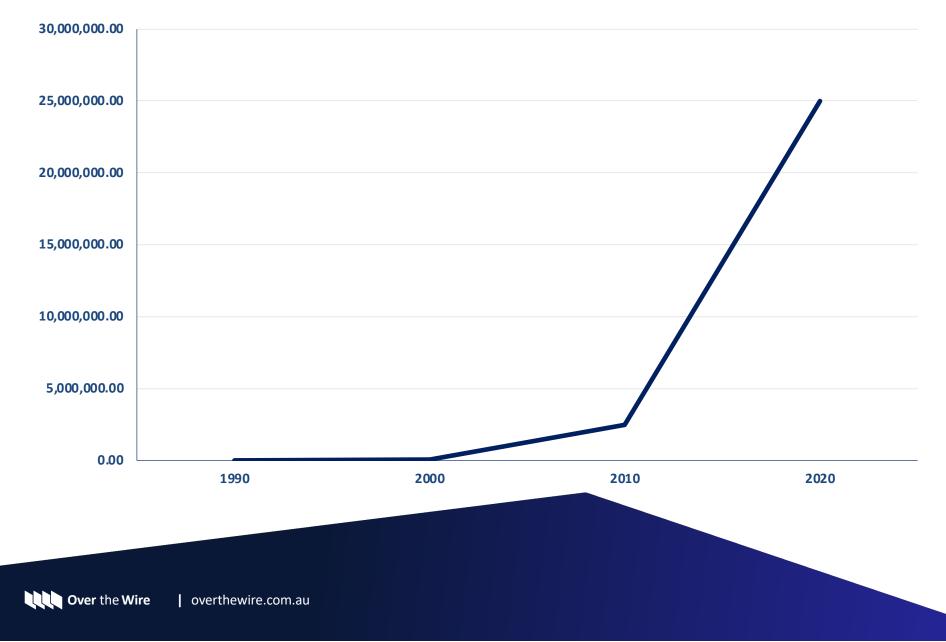
# **Data Growth**

Log scale



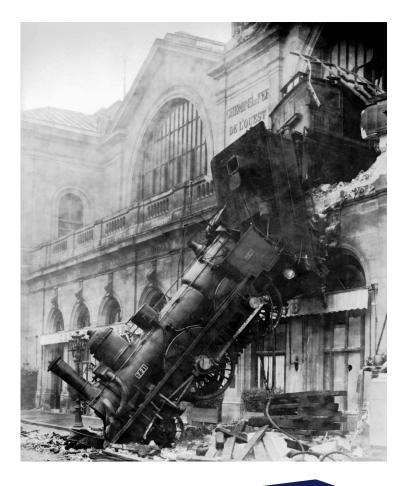
# **Data Growth**

Linear scale



# **Data Growth**

#### Train wreck scale





# **Dealing with Data**

The decade of NoSQL

- "Web Scale"
- Non Tabular
  - ✓ More Flexible
  - ✓ Higher Performance
- Designed to scale-out or cluster
- So many forms to choose from
- Just blame Google 😂

# **A NoSQL For Any Occasion**

#### Key Value Store

E.g. Memcache Simple and high performance. Perfect for local caches

#### Structured Store

E.g. Redis A Key Value Store with typed data. Also great for local caches

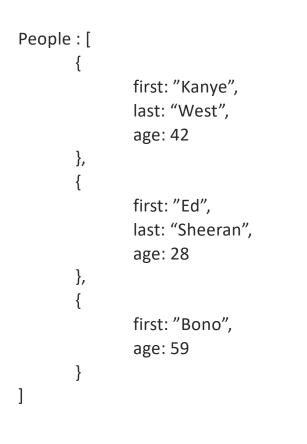
#### Document Database

E.g. MongoDB Key Value where the value is a document, and each document may have a different structure.

#### Graph Database

E.g. Neo4J If your world looks like a graph then this is the business !!

### Tabular data by any other name is still tabular



First	Last	Age
Kanye	West	42
Ed	Sheeran	28
Bono		59



### My Data Wish List

- 1. Tabular data
- 2. SQL Interface
- 3. APIs for every language I can think of
- 4. Grow in data depth without performance hit
- 5. Grow in data width without performance hit

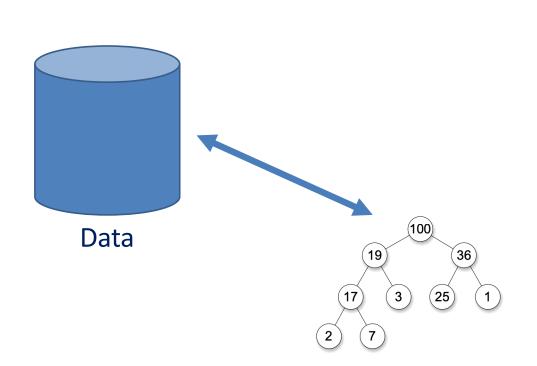
Deliver the x10 scale promise without sacrificing performance, flexibility, or ease of use

## TimescaleDB

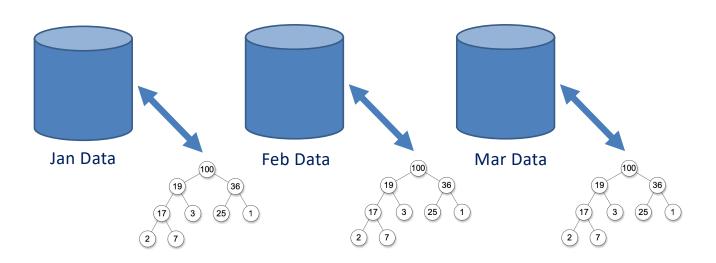
A PostgreSQL extension



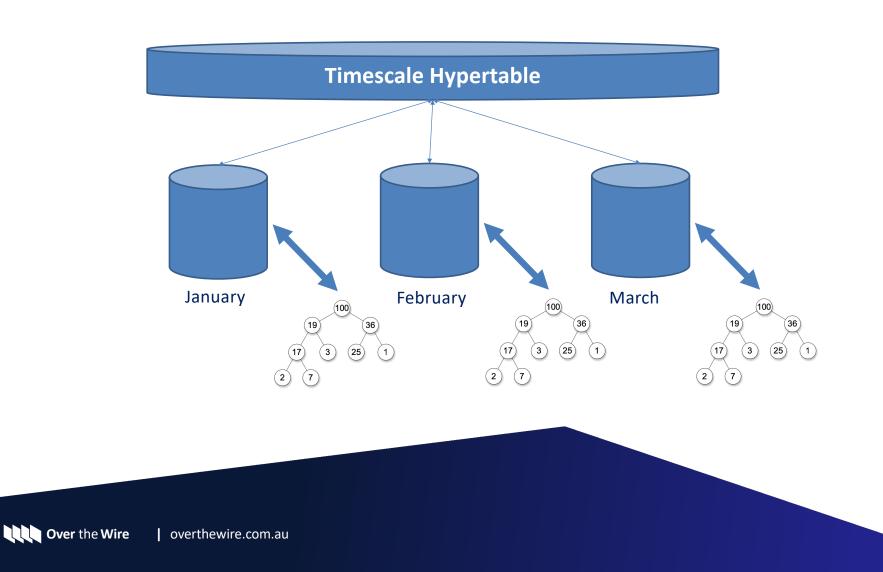
- Specifically written for time series data at scale
- Plugs seamlessly into the Postgres query planner
- Hides the complexity from the developer. Standard SQL
- Works with any tool that works with PostgreSQL
- No retraining if your teams already use PostgreSQL
- Opensource or Commercial

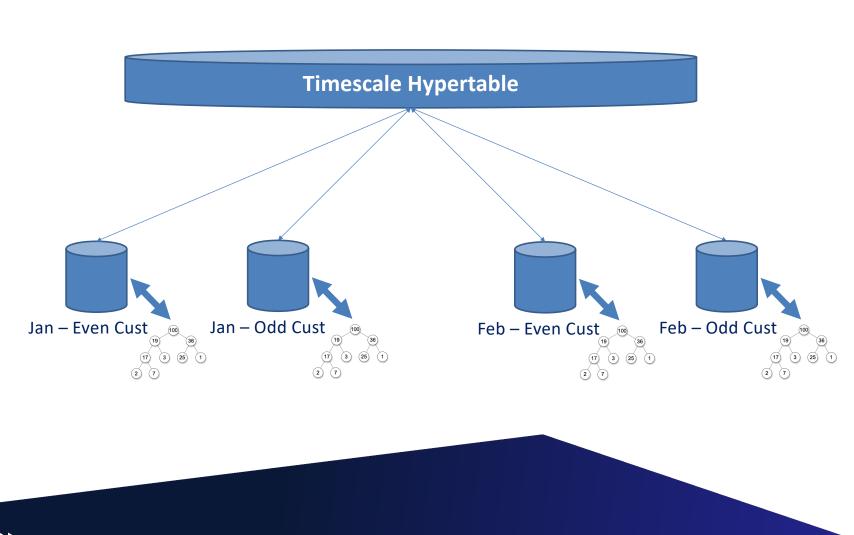












## Hypertables in action

- "Depth" of data should not impact performance. If you keep 10 years of data it will just consume storage until someone queries it.
- Lookup performance requires indices to be loaded. Current indices will be in RAM as you will be writing to the time series in the current period.
- Lookups are commonly grouped. E.g. most customers look at their monitoring for the current month. Locality of lookups keeps indices cached.
- Concurrent writes and reads are handled efficiently.

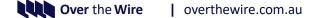
### Hypertables in action

#### **Example Benchmark**

Voip CDR records (very wide data records)3 Million records per day6 months data (i.e. approx. 550 million records)

#### Test 1

No concurrent data inserts Query 1 : Get 1<sup>st</sup> 50 of 1000 records for a client in a given month => Data returned in under 2 seconds Query 2 : Get 2nd 50 records same client and same month => Data returned in 15 msec



#### Hypertables in action

#### Test 2

Base insert rate of 5,000 records per minute Query 1 : Get 1<sup>st</sup> 50 of 1000 records for a client in a given month => Data returned in under 2 seconds Query 2 : Get 2nd 50 records same client and same month => Data returned in 15 msec Test 3

Base insert rate of 50,000 records per minute
Query 1 : Get 1<sup>st</sup> 50 of 1000 records for a client in a given month
=> Data returned in under 3 seconds
Query 2 : Get 2nd 50 records same client and same month
=> Data returned in 25 msec

## Scale out

"Unlimited" data width

#### **Scale-out Option 1**

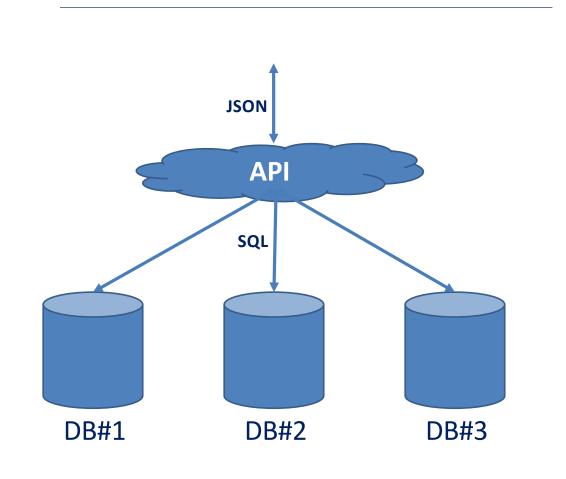
Using TimescaleDB Features

- Clustering in beta
- Uses combination of data nodes and access nodes
- Access nodes route queries to data node

#### **Scale-out Option 2**

Simple solution using "home grown sharding"

## Scale out via "API Sharding"





## Thanks !

Any questions ?

(remember there's coffee outside  $\odot$  )

