NIWA DAS database upgrade

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National Institute of Water and Atmspheric Research

New Zealand Crown Research Institute (CRI)

Government owned commercially funded environmental research institute, with science centres focused on:

Aquaculture Atmosphere Climate Coasts and Oceans Environmental Information Fisheries Freshwater and Estuaries Natural Hazards Pacific Rim



http://niwa.co.nz

Who am I?

Brent Wood

Joined NIWA (or its predecessor) in 1975

Fisheries field/seagoing technician

Data manager/database manager/database designer

Metadata manager (Geonetwork)

Open source GIS user (QGIS, PostGIS, Spatialite, GMT, Mapserver, Geoserver, R, ...)



This presentation describes a rebuild of the NIWA research vessel Data Acquisition System database

It now stores around 40 billion instrument/sensor readings captured since 1990

The 2021 rebuild addressed performance issues with the earlier design, and undertook to meet several user requirements (in addition to normal data management needs):

- rapid access to fine scale (1 min to 1 sec) temporal data (eg: diurnal analytics)
- rapid access to courser scale (1 hr to 1 day) data (eg: seasonal/annual analytics)
- near real time access to manage operational deployments (eg: underwater camera)



It describes: why we did it (as above) what we did how we did it with some example SQL queries

(Gets a bit technical WRT to SQL queries!)





DAS database:

NIWA Data Acquisition System database

Stores instrument/sensor readings, mostly from RV Tangaroa

Historically at 1 min intervals

Captured by various DAS software applications since 1990:

- original OS/2 DAS developed by Datacom
- FRC in-house DAS (Python storage optimised Mb/Gb, not Tb)
- Techsas (IFREMER commercial application)



User application: DASmap



Developed for NIWA in early/mid 2000's by Integrated Mapping (John McCombs, Christchurch)

Written using Mapserver & Python Mapscript as web application

Became too slow as the db grew over a decade or so, but still meets general user needs for previewing & extracting reading data

Alter this to use new database:

- modify embedded SQL's to query hstore table/column structure -
- new queries returning identical result as old queries on old tables



"World's most advanced Open Source Database" (or so they claim!)

PostgreSQL is generally regarded as a reasonable alternative to Oracle as a robust, secure enterprise level RDBMS

PostgreSQL is an ORDBMS, not just an RDBMS (supports object based non-relational data)

PostgreSQL has more features than ANY other RDBMS, and is fully ACID compliant (Atomicity, Consistency, Isolation, Durability)

Already widely used in NIWA



Postgres extensions – What are they:

Additions to core Postgres (often third party) to add focused and specific functionality

The extensions used in this case:

- PostGIS: adds OGC SFS spatial datatypes and functions/operators - probably the most complete and powerful spatial functionality of any RDBMS
- hstore: adds key/value datatype and functions to enter, access & decompose these - simpler than jsonb and meets requirements (core extension)
- TimescaleDB: adds timeseries functionality and auto partitioning - think of it as doing for timeseries data what PostGIS does for spatial data



Postgis:

Postgis provides an enhanced spatial data capability for the Postgres ORDBMS

Exceeds the specification for OGC SFS specification for SQL

Meets the ANSI MM/SQL specifications for SQL databases





hstore:

Core extension providing support for key/value storage in Postgres

What are key-value pairs?

A value stored in an hstore column is a list of keys, each with a value (so a pair) 1=>3.14159,2=>1.4142,3=>2.7183, ...

Why are they appropriate for this use case?

The "key" value is an identifier for a sensor on an instrument (instrument metadata is managed in various other database tables)

The "value" is the sensor value for the timestamp (in another column) in the record

Thus a database column stores all sensor readings for a given timestamp using a structured but non-normalised No-SQL approach within an SQL database



TimescaleDB:

Provides storage, query and analytical tools for timeseries data in a Postgres database

Fundamentally changes your Postgres database –

provides auto-partitioning of timeseries tables (hypertables) for highly optimised performance with never ending streams of accumulating timeseries data

adds tools to aggregate, count, analyse records (quickly!)

we use the free open source version in a local Postgres database, but it is also available on AWS, Azure & Google Cloud.

http://www.timescale.com



Old database: Postgres + Postgis

- Readings normalised (robust, RDBMS "best practice" design), one timestamped reading per record
- 1 minute interval
- Manual one year partitioning
- 600,000,000 readings = 600,000,000 records
- DASmap gui took >3 minutes to load 1 voyage

New database: Postgres + Postgis/Timescaledb/hstore

- Readings as hstore key/value pairs (non-relational approach), one timestamp per record many readings/record, but individually accessible
- 1 second interval
- Automated 1 week chunks in Timescaledb hypertable
- ~40,000,000,000 readings = ~70,000,000 records
- DASmap gui takes <10 seconds to load 1 voyage



Basic table structures:

Old reading table:

(manually maintained yearly partition, 1 min readings) **t_reading** timer timestamp primary key, sensor int pkey + foreign key on t_sensor (key), value int

New reading table:

(*Timescaledb hypertable – auto partition weekly, 1 sec readings*) **t_reading_hstore_sec** timer timestamp primary key values hstore hstore



1 min LIFO extract SQL:

get 1 day of 1 minute LIFO style readings from 1 sec table:

SELECT DISTINCT date_trunc('minute',_timer)+interval '1 minute' as timer, key::int, first_value(value) as last_value OVER (PARTITION BY key, date_trunc('minute',_timer) order by _timer desc) FROM (SELECT timer as _timer, (each(values_sec)).* FROM t_reading_hstore_sec WHERE timer between '2013-04-15 00:00:00' and '2013-04-15 23:59:59') as x ORDER BY key::int, timer;



SELECT DISTINCT date_trunc('minute',_timer)+interval '1 minute' as timer, key::int, first_value(value) as last_value OVER (PARTITION BY key, date_trunc('minute',_timer) ORDER BY _timer desc) FROM (SELECT timer as _timer, (each(values_sec)).* FROM t_reading_hstore_sec WHERE timer between '2013-04-15 00:00:00' and '2013-04-15 23:59:59') as x ORDER BY key::int, timer;

Any guesses how long it takes (from 40b readings)?

To get the values for 1 minute only (just change "between" value): ???

To get the values for 1 day (1440 minutes): ???



SELECT DISTINCT date_trunc('minute',_timer)+interval '1 minute' as timer, key::int, first_value(value) as last_value OVER (PARTITION BY key, date_trunc('minute',_timer) ORDER BY_timer desc) FROM (SELECT timer as _timer, (each(values_sec)).* FROM t_reading_hstore_sec WHERE timer between '2013-04-15 00:00:00' and '2013-04-15 23:59:59') as x ORDER BY key::int, timer;

Any guesses how long it takes (from 40b readings)?

To get the values for 1 minute only (just change "between" value): <30 ms

To get the values for 1 day (1440 minutes): ~55 sec (~7hrs for a year)

SELECT DISTINCT date_trunc('minute',_timer)+interval '1 minute' as timer, key::int, first_value(value) as last_value OVER (PARTITION BY key, date_trunc('minute',_timer) ORDER BY_timer desc) FROM (SELECT timer as _timer, (each(values_sec)).* FROM t_reading_hstore_sec

WHERE timer between '2013-04-15 00:00'

and '2013-04-15 23:59:59') as x

ORDER BY key::int, timer;

timer		key	last_value
2013-04-15 00:01:00 2013-04-15 00:02:00	- - 	27 27	17.799999237060547 17.700000762939453
 2013-04-15 01:20:00 2013-04-15 01:21:00		63 63	1385.6600341796875 1394.30004882815



SQL to get 1 minute lats/longs for a trip (7 weeks) from hstore key/value pairs:

Use COALESCE to get first non-NULL value as there are several GPS units with different keys over the years...

SELECT timer, COALESCE (value -> '67', value -> '307', value -> '317', ") as lat, COALESCE (value -> '68', value -> '308', value -> '318', ") as lon FROM t_reading_hstore_min WHERE timer >= '20210101'::date AND timer <= '20210217'::date ORDER BY timer;

Execution time: ???



SQL to get one minute lats/longs for a trip (6 weeks) from hstore key/value pairs:

SELECT timer, COALESCE (value -> '67', value -> '307', value -> '317', '') as lat, COALESCE (value -> '68', value -> '308', value -> '318', '') as lon FROM t_reading_hstore_min WHERE timer >= '20210101'::date AND timer <= '20210217'::date ORDER BY timer;

Execution time: ~1 sec



SQL to get one minute lats/longs for a trip of 7 weeks) from hstore key/value pairs:

```
SELECT timer,
	COALESCE (value -> '67', value -> '307', value -> '317', ") as lat,
	COALESCE (value -> '68', value -> '308', value -> '318', ") as lon
FROM t_reading_hstore_min
WHERE timer >= '20210101'::date
	AND timer <= '20210217'::date
	ORDER BY timer;
```

timer	lat	l lon
2021-01-01 00:00:00 2021-01-01 00:01:00 2021-01-01 00:02:00	-41.311622 -41.311621 -41.311621	174.811812 174.811812 174.811812 174.811812





SQL to create a voyage trackline from lat & lon coordinates (yep - GIS at last!):

```
-- build linestring from raw hstore coords
-- use ST RemoveRepeatedPoints() to clean repeated points
-- use ST Simplify() to drop resolution to about 10m (remove redundant - nearby - vertices)
-- ST Simplify @ 0.00001=34000 vertices, @ 0.0001=10000 vertices
SELECT ST AsText(
     ST Simplify(
      ST RemoveRepeatedPoints(
       ST Makeline(
        ST SetSRID(
          ST MakePoint(
           COALESCE (value -> '68', value -> '308', value -> '318')::real,
           COALESCE (value -> '67', value -> '307', value -> '317')::real),
        4326)
       ORDER BY timer)
      ),0.0001)) as track
FROM t reading hstore min
WHERE timer >= '20210101'::date
   AND timer <= '20210217'::date;
```

Execution time: ???



SQL to create a voyage trackline from lat & lon coordinates:

```
-- build linestring from raw hstore coords
-- use ST RemoveRepeatedPoints() to clean repeated points
-- use ST Simplify() to drop resolution to about 10m (remove redundant - nearby - vertices)
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SELECT ST AsText(
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           COALESCE (value -> '68', value -> '308', value -> '318')::real,
           COALESCE (value -> '67', value -> '307', value -> '317')::real),
        4326)
       ORDER BY timer)
      ),0.0001)) as track
FROM t reading hstore min
WHERE timer >= '20210101'::date
   AND timer <= '20210217'::date;
```

Execution time: <1.3 sec



SQL to return instruments/sensors for a survey:

SELECT i.instrument_code||'_'||m.measurement_code as reading, m.measurement key FROM t measurement m. t instrument I WHERE m.measurement key::varchar in (SELECT distinct key FROM (SELECT skeys(value) as key FROM t reading hstore min WHERE timer = '20210101'::date AND timer <= '20210217'::date) as mytable) and m.instrument key = i.instrument key ORDER BY reading;



Execution time: ???

SQL to return all sensors with data for a survey:

SELECT i.instrument_code||'_'||m.measurement_code as reading, m.measurement key FROM t measurement m. t instrument I WHERE m.measurement key::varchar in (SELECT distinct key FROM (SELECT skeys(value) as key FROM t reading hstore min WHERE timer = '20210101'::date AND timer <= '20210217'::date) as mytable) and m.instrument key = i.instrument key ORDER BY reading;



Execution time: <7ms

SQL to return all sensors with data for a survey:

SELECT i.instrument_code||'_'||m.measurement_code as reading, m.measurement_key FROM t_measurement m, t_instrument I WHERE m.measurement_key::varchar in (SELECT distinct key FROM (SELECT skeys(value) as key FROM t_reading_hstore_min WHERE timer = '20210101'::date AND timer <= '20210217'::date) as mytable) AND m.instrument_key = i.instrument_key ORDER BY reading;

reading	measurement_key
AWS_AirTemp	27
AWS_BarPressur	37
AWS_DewPoint	56
Eppley_PIR1_mV	267
Eppley_PIR1_Rdome	268
GYRO_P1HDT	95
Rainfall_RG_mm	473

. . .



SQL to return 1000 vessel and DTIS camera point locations from 1 min table

Use in QGIS to display camera & vessel positions in **near real time during deployment** (builds the point geometries on-the-fly from hstore key/value pairs in auto refresh map layer)

select timer. extract(epoch from timer) as id, h.value -> '307' as vess lat, h.value -> '308' as vess lon, h.value -> '256' as dtis lat, h.value -> '257' as dtis lon, (ST SetSRID(ST Makepoint((h.value -> '257')::decimal(10,7), (h.value -> '256')::decimal(10,7)),4326)) as dtis geom, (ST SetSRID(ST Makepoint((h.value -> '308')::decimal(10,7),(h.value -> '307')::decimal(10,7)),4326)) as geom from t_reading_hstore min h where (h.value -> '256') notnull and (h.value -> '256') != '0.0' and timer > '2018-05-14' --- or use now() - interval('1 day') order by timer asc limit 1000;

Execution time: ???



The transect data from the query from OGIS



and zoomed in



SQL to return vessel and DTIS camera (HiPAP) point locations from 1 min table

Use in QGIS to display camera & vessel positions in near real time (builds the point geometries on-the-fly from hstore key/value pairs)

select timer, extract(epoch from timer) as id, h.value -> '307' as vess lat, h.value -> '308' as vess lon, h.value -> '256' as dtis lat, h.value -> '257' as dtis lon, (ST SetSRID(ST Makepoint((h.value -> '257')::decimal(10,7), (h.value -> 256)::decimal(10,7),4326) as dtis geom, (ST SetSRID(ST Makepoint((h.value -> '308')::decimal(10,7), (h.value -> '307')::decimal(10,7)),4326)) as geom from t reading hstore min h where (h.value -> '256') notnull and (h.value -> '256') != '0.0' and timer > '2018-05-14' order by timer asc limit 1000;

Execution time: 50ms



The transect data from the query



Zoomed in



On disk data volumes:

Difficult to be accurate, we are still finding new indexes to build to improve performance (which take up space)...

It seems the new approach stores around 70x the number of readings in about 40% of the space used by the old database.

Summary:

Postgresql with Postgis, Timescaledb and hstore extensions works pretty much out of the box, is very fast, and very effective.

We have not (yet) had any instances where we could not reasonably easily and quickly extract the data we desired in the format we needed.

We are pleased and impressed with this suite, very fit for this purpose.

And situation normal – we have yet to complete the database documentation!



Types of sensors:

While most of the sensors are collecting environmental data, various vessel systems (engines, pumps, etc) also provide telemetry data which is monitored and captured by the DAS, and so ends up in the DAS database.

It is useful to have winch for an instrument deployed off that winch, and to have data such as load, oil pressures and temperatures from engines and pumps, not just from a science perspective, but from a vessel operation perspective.

Similarly – an acoustic scientist who finds noise in their data for a transect can see what else was happening at the time –

eg: engineers starting and testing a backup generator during a transect



This work was funded as a NIWA IT CAPEX project

My thanks to Andrea Mari and Simon Wood for getting this done, and within budget!!!



Thank you for attending!

Questions?

