

Scaling AWS Redshift Concurrency with Postgres

By Elliott Cordo, Will Liu, Paul Singman

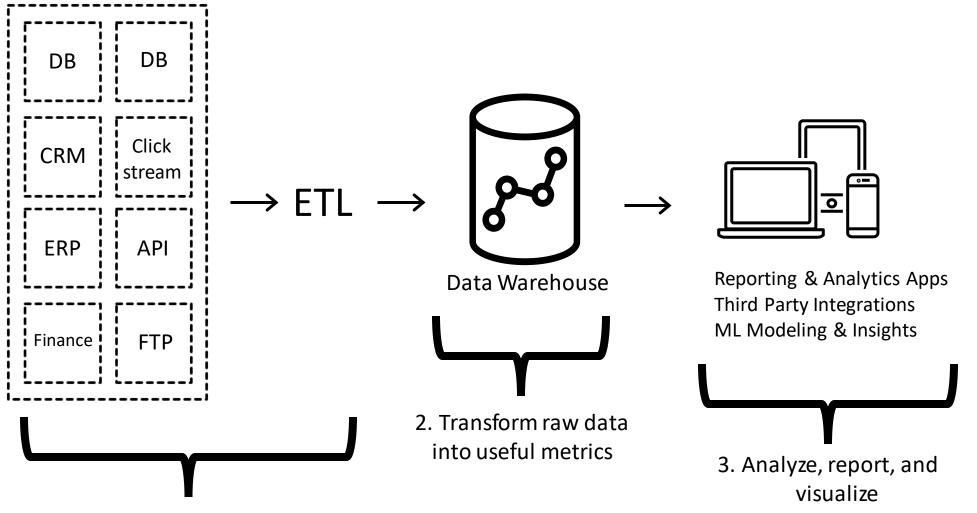
EQUINCX

Integrated luxury and lifestyle company with offerings centered on movement, nutrition, and regeneration



we operate more than 200 locations within every major city across the country in addition to London and Canada

Analytics Overview



1. Extract data from source systems

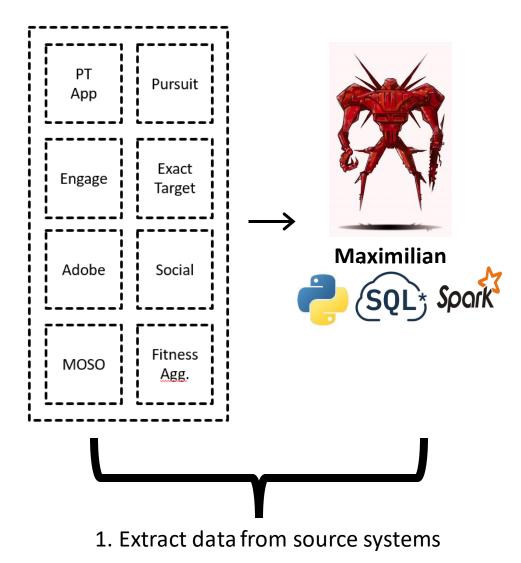
Why A Data Warehouse?

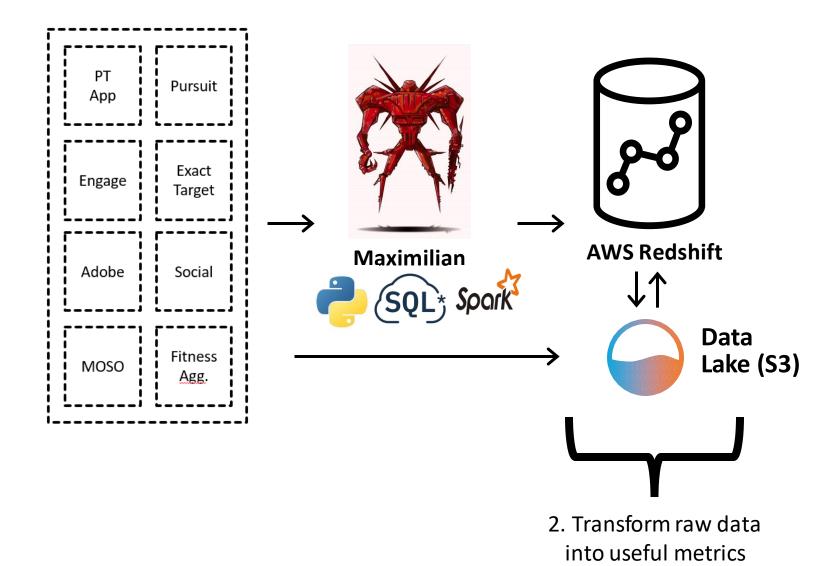
- Prevent data discrepancies
- Lower employee learning curve
- Avoid duplicating logic in multiple systems
- Isolate production DBs from analytic workloads

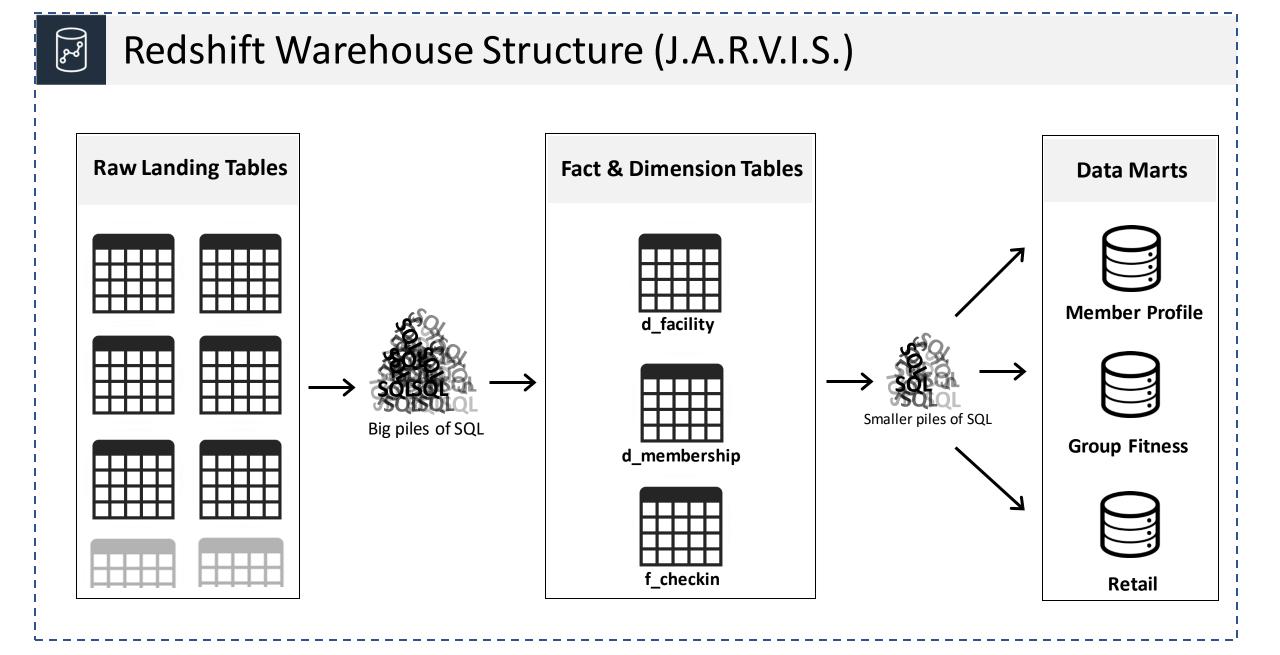
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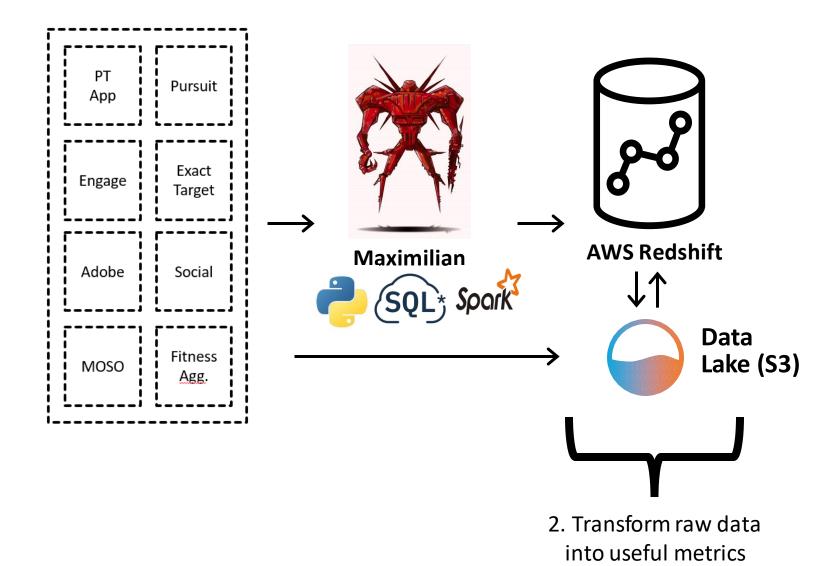
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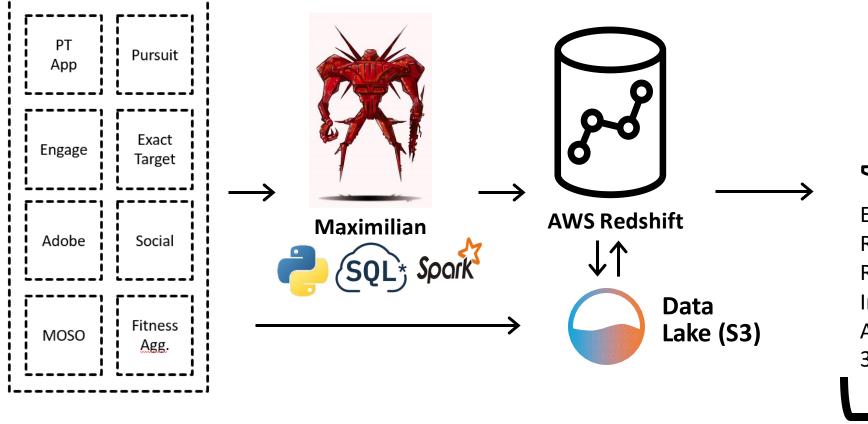
Presents a single point of failure so we created a data replication failover procedure







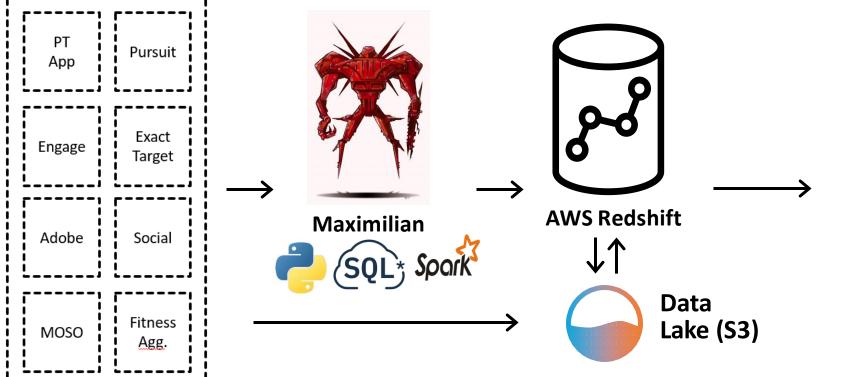






Equinox Apps Reporting/Dashboards Recommender Systems Internal APIs Ad-hoc Analysis 3rd Party Data Integrations

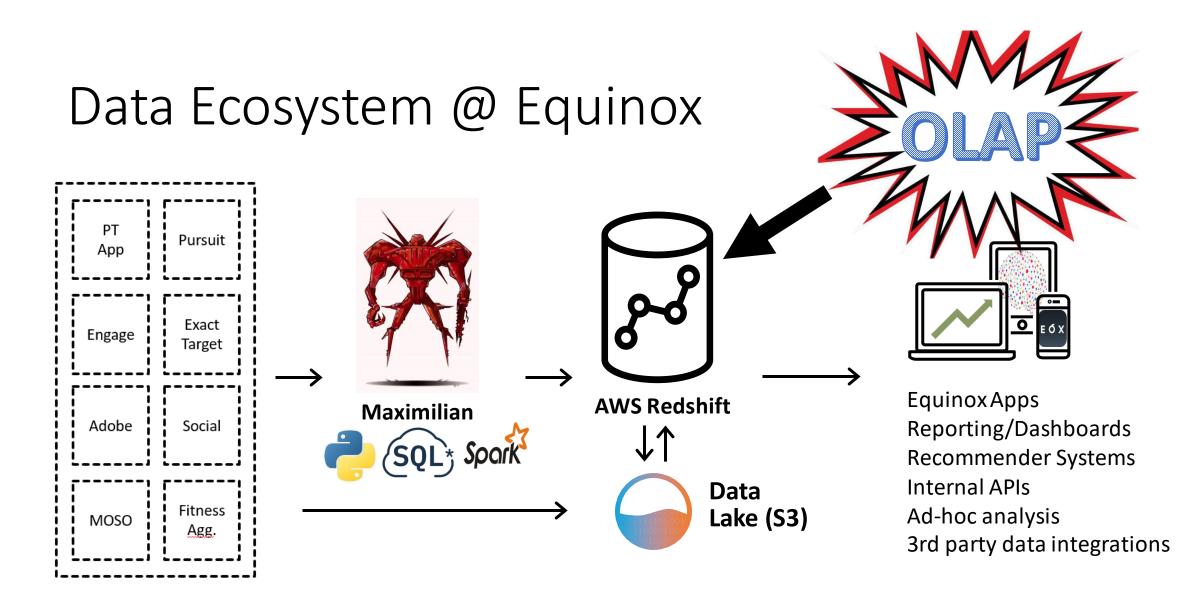
3. Analyze, report, and visualize





Equinox Apps Reporting/Dashboards Recommender Systems Internal APIs Ad-hoc Analysis 3rd Party Data Integrations

🏽 🖉 4. Data Monitoring & Quality











On-line Transactional Processing (OLTP) On-line Analytical Processing (OLAP)

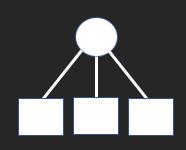
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- Column-oriented storage
- Immutable 1MB block storage
- Massively-parallel processing compute engine
- Native multi-node (leader + workers) architecture
- Distribution key & sort key table settings
- Workload Management queue settings

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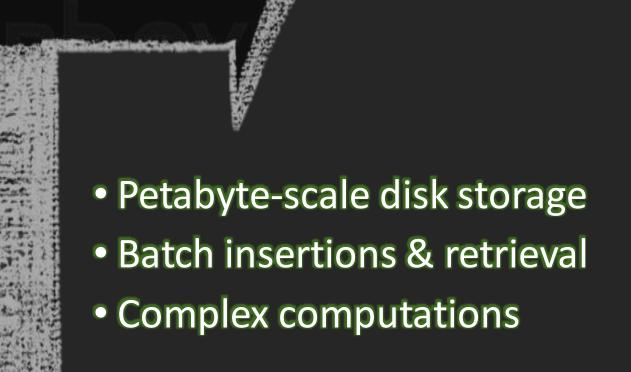


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Mandani Malan Salah S WEAKNESS STRENGTH



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High-frequency transactions
Concurrent user connections

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Petabyte-scale disk storage
Batch insertions & retrieval
Complex computations

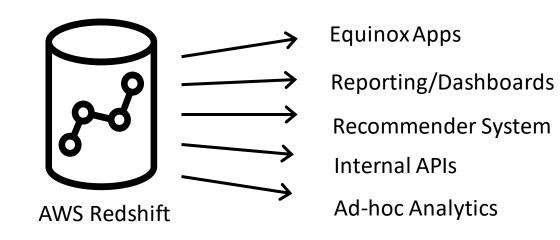
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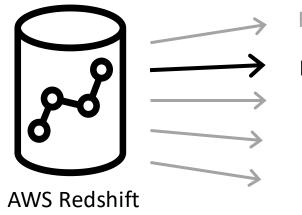
10x - 100x slower on simple SELECT500 connection limit (per cluster)50 connection limit (per user-defined queue)

Petabyte-scale disk storage
Batch insertions & retrieval
Complex computations

Warehouse Consumers



Warehouse Consumers



Equinox Apps

Reporting/Dashboards *

Recommender System

Internal APIs

Ad-hoc Analytics

Sales Review Details

From Date: 3/1/2019 To Date: 3/17/2019

Ashmentle, Emma

<u>3DayCXL</u>

Acct Code	Club	Member Id	Member Name	Member Status	Date	Payment Type	Class	Commission Status	Count
115	Park Avenue	1003931724	Moraloo, Podro	Cancelled	3/6/19	BM	Select Monthly	NO COMM	-1
115	Park Avenue	1003921211	Michael, Anthony	Cancelled	3/12/19	PIF	All Access Yearly	INC SIGNATURES, INCOMPLETE CONTACT INFO	-1
								Total Count:	-2

<u>3DayReactivate</u>

Acct						Payment		Commission	
Code	Club	Member Id	Member Name	Member Status	Date	Туре	Class	Status	Count
115	Park Avenue	1003931724	Moraloo, Pedre	Active	3/6/19	BM	All Access Monthly	DUE 1	1
								Total Count:	1

<u>New</u>

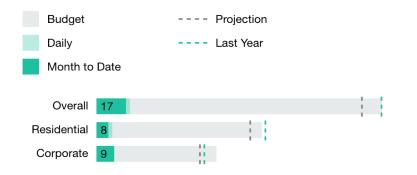
Acct						Payment		Commission	
Code	Club	Member Id	Member Name	Member Status	Date	Туре	Class	Status	Count
115	Park Avenue	1001742000	Maroinik, Pyan	Active	3/4/19	BM	All Access Monthly	DUE 3	1
115	Park Avenue	1003929636	Mestafa, Mayada	Active	3/4/19	BM	Select Monthly	DUE 1	1
115	Park Avenue	1003900694	Khan, Addil	Active	3/5/19	BM	All Access Monthly	INC CORPORATE	1
115	Park Avenue	1003932836	Witter, Priotine	Active	3/5/19	BM	Select Monthly	INC SIGNATURES	1
115	Park Avenue	1003932984	Buoa, Camantha	Active	3/5/19	BM	Select Monthly	DUE 1	1
115	Park Avenue	1001796524	Accola, Parala	Active	3/6/19	BM	Select Monthly	DUE 1	1
115	Park Avenue	1003931724	Meralee, Pedre	Active	3/6/19	BM	Select Monthly	NO COMM	1
115	Park Avenue	1003198113	Dimbaam, Natali e	Pending Start	3/7/19	BM	All Access Monthly	DUE 1	1
115	Park Avenue	1003630343	Menamera, Koli	Pending Start	3/7/19	BM	Select Monthly	DUE 3	1
	D 1 4	1000000000	- · · · · ·		0.7140		~ · · · · · · · · · · · · · · · · · · ·		l

ΕQUINOX

SALES

The progress bar charts show the daily actual, month to date and budget relationship of sales. The gray bars are the end of month budget as 100 percent, the green ones are the month to date actual in percentage of budget, with the exact value showing as label. The daily new actual value is marked out in lighter green color.

Green dot lines and gray dot lines are the projection and last year value as percentage of each category.

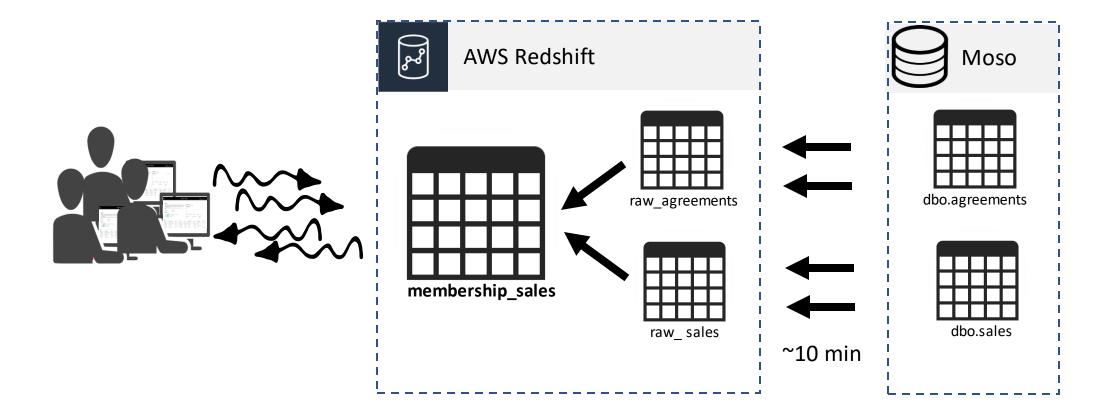




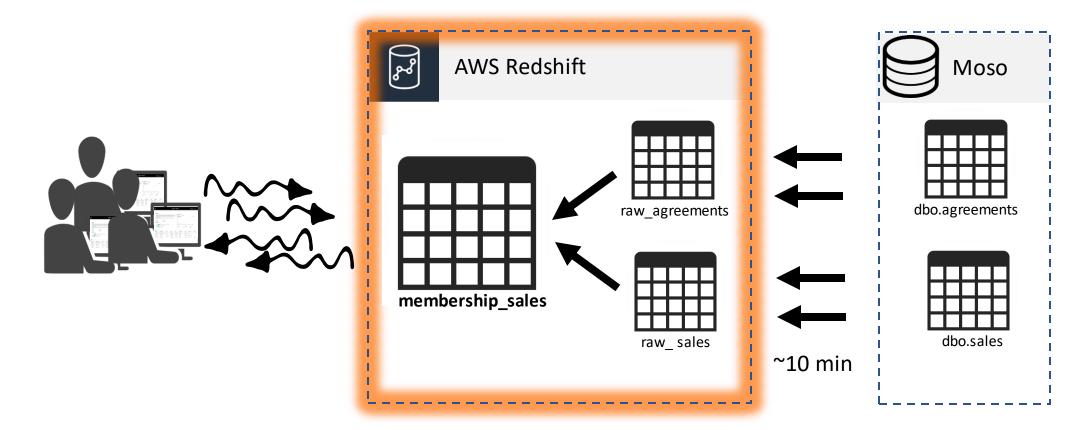
NEW

Member ID	Member Name	Status 🌲	Advisor	Date	Payment Type	Class	Commission \$\$	Equifit 🔶
1003858732	Maudolo del Bosque	Active	Tony Marchena	2019-03-02	Perpetual	All Access	DUE 1	
1003936006	Brian Plummer	Active	Ryan Sanders	2019-03-06	Perpetual	All Access	DUE 1	
1003932537	Eliana Arian	Pending Start	Ryan Sanders	2019-03-07	Perpetual	All Access	INC CORPORATE	
1003202293	Sarpreet Khera	Active	Ryan Sanders	2019-03-03	Perpetual	All Access	DUE 2	Confirmed

Sales Reporting Architecture



Sales Reporting Architecture Problem



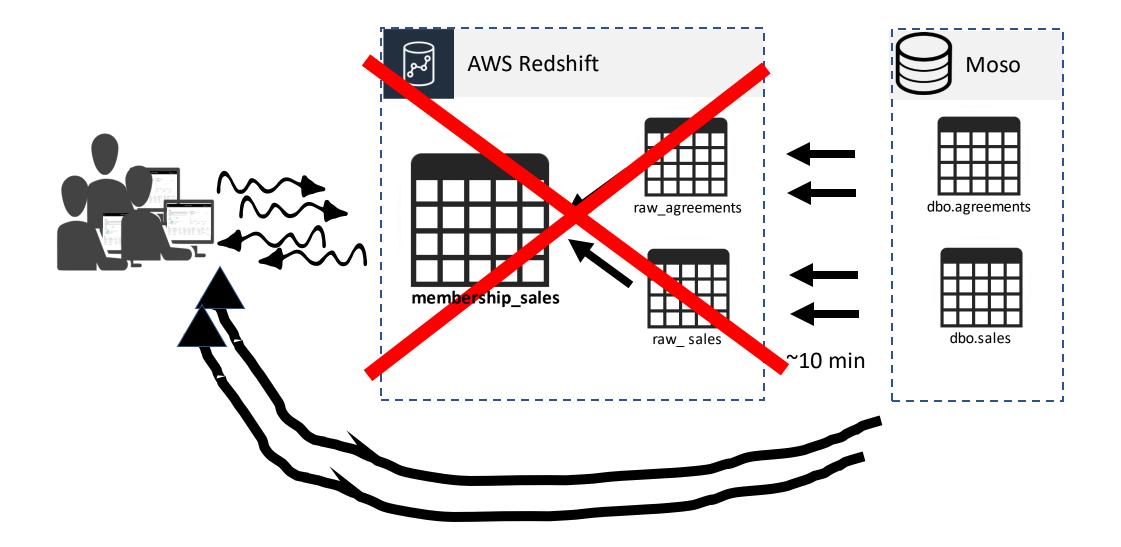
10x - 100x slower on simple SELECT500 connection limit (per cluster)50 connection limit (per user-defined queue)

Potential Solutions

Potential Solutions

1. Pull from source DB

Sales Reporting Architecture



1. Pull from source DB

Pros

• Source DB is OLTP

Cons

- Sales logic is complex!
- Burdens prod DB

1. Pull from source DB

Pros

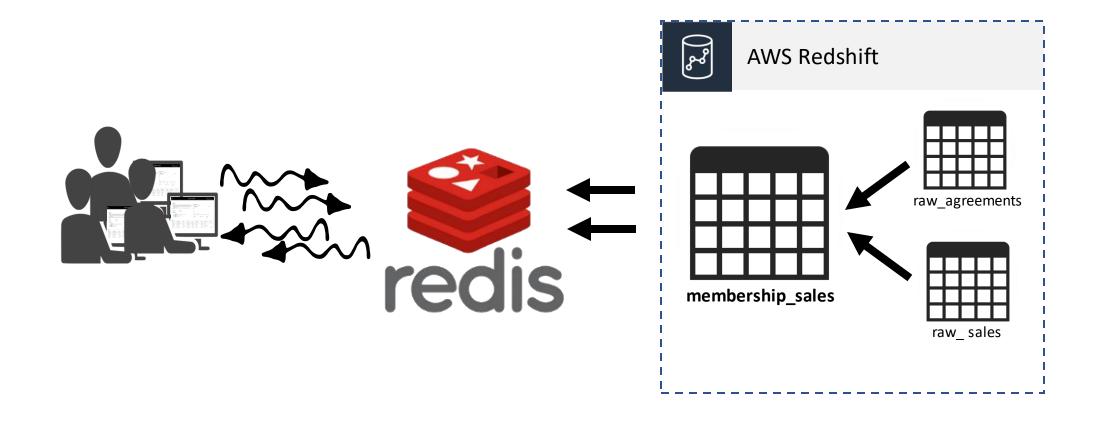
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Sales Reporting Architecture



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• Very fast performance

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- Non-relational data structure
- Keys must be created for every data view

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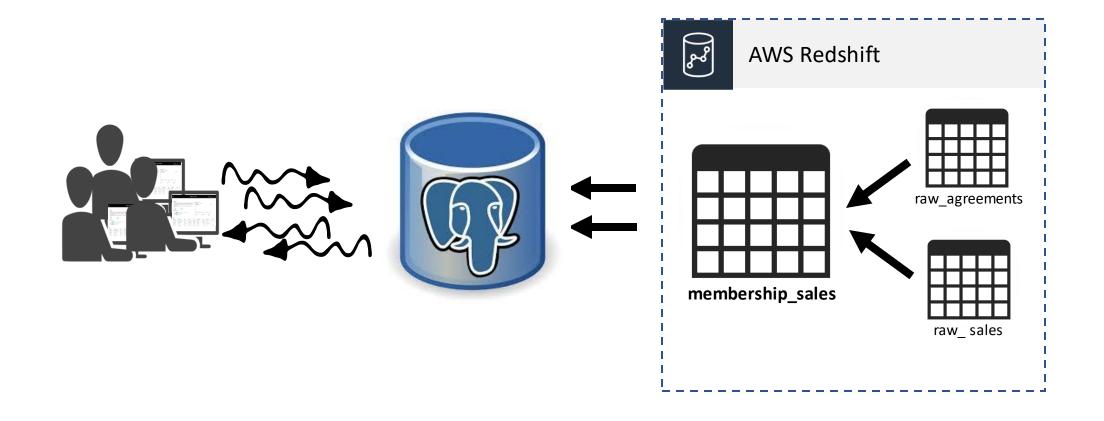
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Pros

• Maintain relational format

Cons

• Additional ETL step

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Copy to another DB

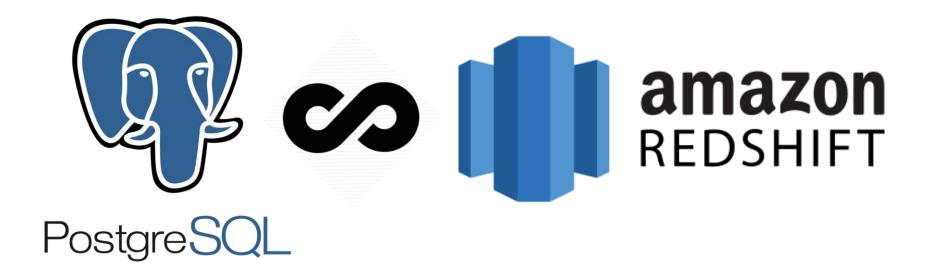
Pros

Maintain relational format

Cons

Additional ETL step

PostgreSQL Foreign Data Wrapper



PostgreSQL Foreign Data Wrapper



JOIN Amazon Redshift AND Amazon RDS PostgreSQL WITH dblink by Tony Oldow | on 03 Mile 2016 | in Amazon RDS, Amazon Redshift, AWS elig Data | Permatinik | 🖝 Comments | 🖛 Share

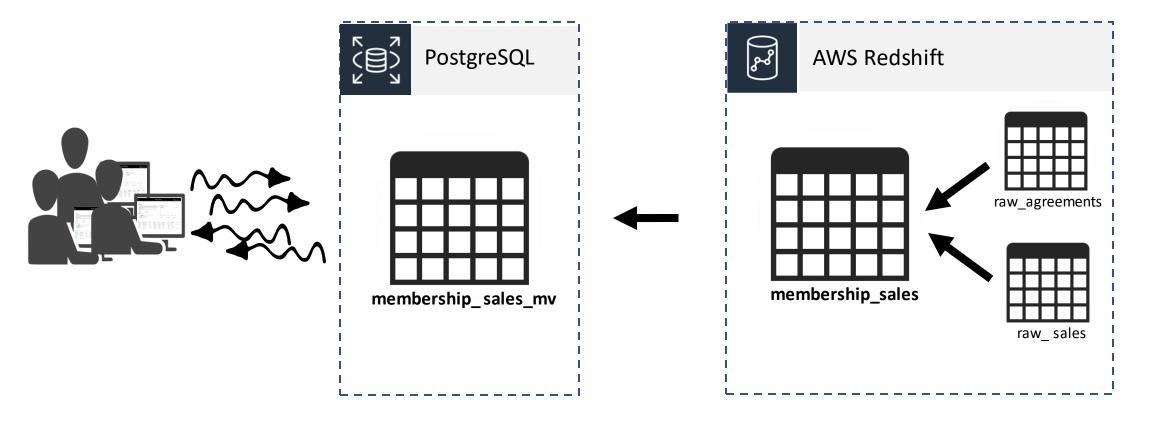
Tony Gibbs is a Solutions Architect with AWS

(Update: This blog post has been translated into Sepanese)

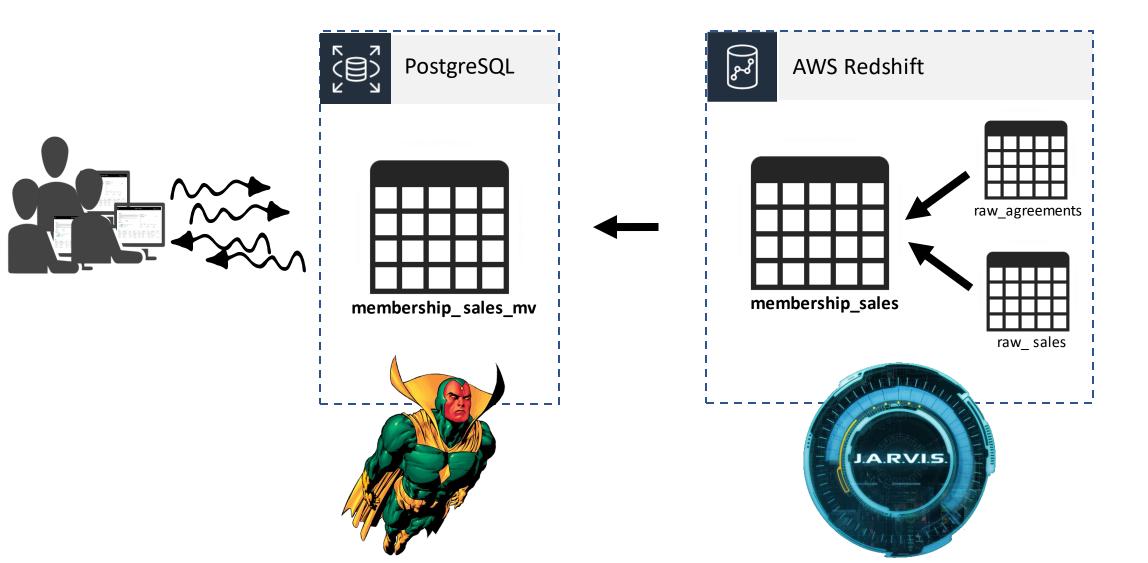
When it comes to choosing a SQL-based database in AWS, there are many options. Sometimes it can be difficult to know which one to choose. For example, when would you use Amazon Aurora instead of Amazon RDS PostgreSQL or Amazon Redshift? To answer this question, you must first understand the nature of the data workload and then evaluate other factors such as the quantity of data and query access patterns.

The design and capabilities of the different AWS services mean that each service has different strengths and excels at different workloads. This leads to trying to pick the right tool for the job, which can result in tradeoffs. But sometimes

Proposed Sales Reporting Architecture



Proposed Sales Reporting Architecture



Environment Set-up

Create Redshift cluster

Create PostgreSQL server (9.5+)

- RDS recommended
- For self-managed, install Postgres contrib package:
 - sudo yum install postgresql10-contrib.x86_64

Networking (AWS)

- Co-locate in same Availability Zone
- Configure Security Group

Creating the Link

--1 enable the required extensions CREATE EXTENSION postgres_fdw; CREATE EXTENSION dblink;

--2 create the external server CREATE SERVER jarvis FOREIGN DATA WRAPPER postgres_fdw OPTIONS (host 'REDSHIFT_ENDPOINT', port '5439',

dbname 'REDSHIFT_DB_NAME', sslmode 'require');

--3 save redshift login to this external server CREATE USER MAPPING FOR PG_USERNAME SERVER Jarvis OPTIONS (user 'RS_USERNAME', password 'RS_PASSWORD');

Running queries on PostgreSQL

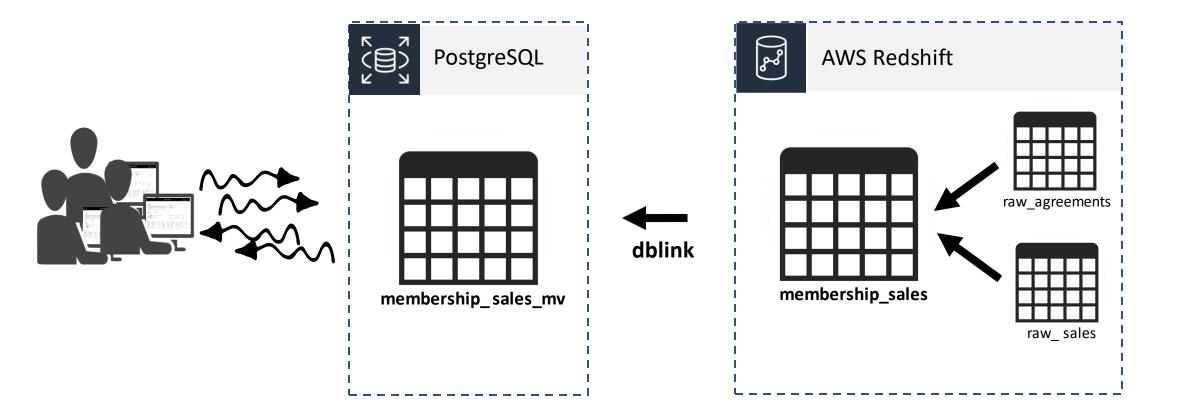
SELECT * FROM dblink('jarvis', \$REDSHIFT\$ SELECT member_sales_id, member_id, sales_action, sales_action_date FROM rs_landing.raw_sales \$REDSHIFT\$) AS sales_actions (member_sales_id varchar(50), member_id varchar(50), sales_action varchar(50), sales_action_date date);

Leveraging a Materialized View

```
CREATE MATERIALIZED VIEW pg.membership sales copy AS
  SELECT *
  FROM dblink('jarvis', $REDSHIFT$
  SELECT
   member sales id,
   member id,
   sales action,
   sales action date
  FROM
   rs.membership_sales $REDSHIFT$
) AS membership_sales_copy (
       member_sales_id varchar(50),
       member_id varchar(50),
       sales_action varchar(50),
       sales action date date
);
```

REFRESH materialized VIEW pg.membership_sales_copy;

Sales Reporting Architecture



Materialized View Roadblock

membership_sales_mv contract_id batch_id timestamp 1111 101 2019-03-10 2222 101 2019-03-10 3333 101 2019-03-10 4444 101 2019-03-10 5555 102 2019-03-21	PostgreSQL	rshin s	alos my		(مگ	Ļ
2222 101 2019-03-10 3333 101 2019-03-10 4444 101 2019-03-10			_			
3333 101 2019-03-10 4444 101 2019-03-10 Image: Control of the	1111	101	2019-03-10			
3333 101 2019-03-10 4444 101 2019-03-10	2222	101	2019-03-10			
~10 mins	3333	101	2019-03-10	Of records		
	4444	101	2019-03-10	~10 mins		
5555 102 2019-03-21				10 111113		
	5555	102	2019-03-21			
6666 102 2019-03-21	6666	102	2019-03-21			

AWS Redshift

membership_sales

contract_id	batch_id	timestamp
1111	101	2019-03-10
2222	101	2019-03-10
3333	101	2019-03-10
4444	101	2019-03-10
5555	102	2019-03-21
6666	102	2019-03-21

Change Data Capture for Large Tables

	ostgreS	GQL							کی م	AWS Re	edshift				
memb	NEW bership	_sales											memb	pership	_sales
contract_id	batch_id	timestamp						i.					contract_id	batch_id	timestamp
							1	1							
1111	101	2019-03-10					- I	1	<u> </u>				1111	101	2019-03-10
	101 101	2019-03-10 2019-03-10								IEW			1111 2222	101	2019-03-10
2222										IEW	-		2222	101	2019-03-10
2222 3333	101	2019-03-10	n	nember	ship_s	ales_mv			nember	ship_s	ales_cdo	2	2222 3333	101 101	2019-03-10 2019-03-10
2222 3333 4444	101 101	2019-03-10 2019-03-10										:	2222 3333 4444	101 101 101	2019-03-10 2019-03-10 2019-03-10
2222 3333 4444 	101 101 101	2019-03-10 2019-03-10 2019-03-10	_	contract_id	batch_id	timestamp			contract_id	batch_id	timestamp		2222 3333 4444 	101 101 101 	2019-03-10 2019-03-10 2019-03-10
11111 2222 3333 4444 5555 5555	101 101 101 	2019-03-10 2019-03-10 2019-03-10 	_				dblink						2222 3333 4444	101 101 101	2019-03-10 2019-03-10 2019-03-10

--Step 1: Create staging table in Redshift with last few hours of sales actions --CREATE TABLE rs_landing.stage_sales_action DELETE FROM rs.membership sales cdc

INSERT INTO rs.membership_sales_cdc SELECT member_sales_id, member_id, sales_action, sales_action_date FROM rs.membership_sales WHERE date >= ' \$[?from_date]';

--Step 2: Refresh materialized view in Postgres REFRESH materialized VIEW pg.membership_sales_mv;

--Step 3: Upsert logic to populate final table in Postgres from materialized view --temp table to hold last batch DROP TABLE IF EXISTS cdc_sales; CREATE TEMP TABLE cdc_sales AS SELECT * FROM pg.membership_sales_mv;

--update changed records, member_sales_id as the key to identify a unique record UPDATE pg.membership_sales ms SET sa.member_id = s.member_id, ms.sales_action = s.sales_action, ms.sales_action_date = s.sales_action_date FROM cdc_sales s WHERE s.member sales id = ms.member sales id;

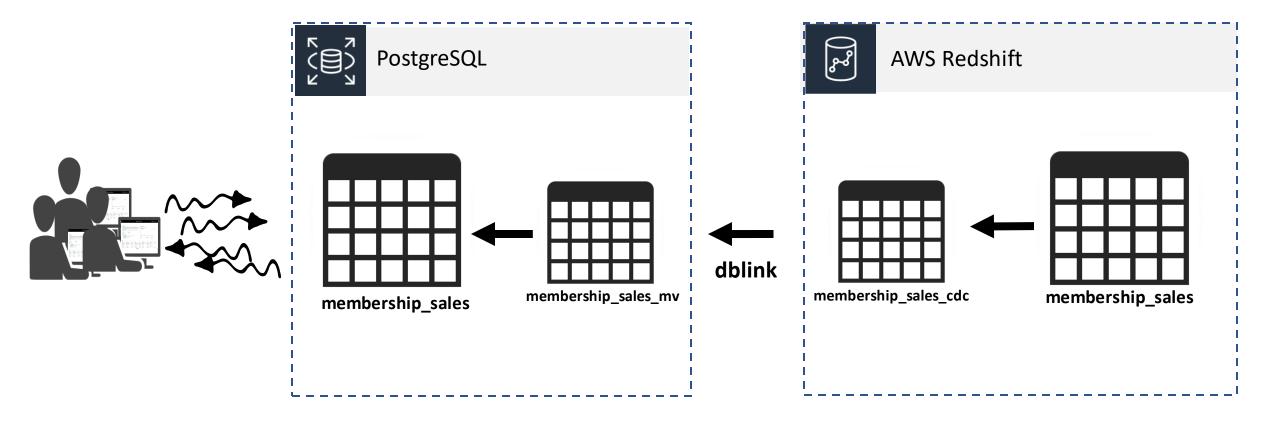
--delete the records we just updated from temp table DELETE FROM cdc_sales s USING pg.membership_sales ms WHERE s.member_sales_id = ms.member_sales_id;

--insert new records not found in membership_sales INSERT INTO pg.membership_sales SELECT * FROM cdc sales;

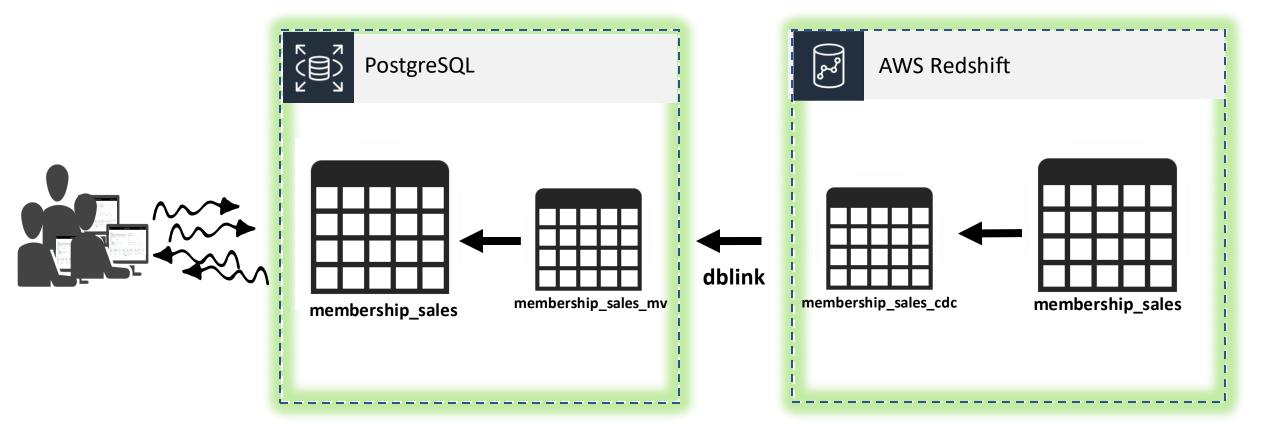
--drop temp table DROP TABLE cdc_sales; Redshift

PostgreSQL

Sales Reporting Architecture



Sales Reporting Architecture Solution



Learnings & Notes

- Minimal maintenance on Postgres instance
- Won't reflect source deletions
- Limited to a few tables
- Flexible for schema evolution



Looking to the Future...

- Make use of foreign table
- Front-end scaling with read-replicas
- Extensible to other datastores
- Event-based streaming architecture

More to Explore

Appendix F. Additional Supplied Modules

Table of Contents

F.1. adminpack

F.2. amcheck

F.2.1. Functions

F.2.2. Using amcheck effectively

F.2.3. Repairing corruption

F.3. auth_delay

F.3.1. Configuration Parameters

F.3.2. Author

F.4. auto_explain

F.4.1. Configuration Parameters

F.4.2. Example

F.4.3. Author

F.5. bloom

F.5.1. Parameters

F.5.2. Examples

F.5.3. Operator Class Interface

F.5.4. Limitations

F.5.5. Authors

F.6. btree_gin

F.6.1. Example Usage

F.6.2. Authors

F.7. btree_gist

F.7.1. Example Usage

F.7.2. Authors

F.8. chkpass

F.8.1. Author

F.9. citext

Generic SQL Database Wrappers

Data Source	Туре	License	Code	Install	Doc	
ODBC	Native		github 🔒			CartoD
JDBC	Native		github 🔒			
JDBC2	Native		github 🔒			
SQL_Alchemy	Multicorn 🔒	PostgreSQL	GitHub 🔒	PGXN 🔒	documentation 🔒	Can be used to acc
VirtDB	Native	GPL	GitHub 🔒			A generic

Specific SQL Database Wrappers

Data Source	Туре	License	Code	Install	Doc	
PostgreSQL	Native	PostgreSQL	git.postgresql.org 🔒		documentation 🔒	
Oracle	Native	PostgreSQL	github 🔒	PGXN 🔒	website 🗗	
MySQL 🔒	Native		github 🔒	PGXN 🔒	example 🔒	
Informix	Native	PostgreSQL	github 🔒			
Firebird	Native		github 🔒	PGXN 🔒		
SQLite	Native		github 🔒			
SQLite 🔒	Native	PostgreSQL	github 🔒	PGXN 🔒	README 🔒	An FDV
Sybase / MS SQL Server	Native		github 🔒	PGXN 🔒		
MonetDB 🔒	Native		github 🔒			

NoSQL Database Wrappers

Data Source	Туре	License	Code	Install	Doc
BigTable or HBase 🔒	Native Rust Binding (RPGFFI)	MIT	Github 🔒		
Cassandra 🚱	Multicorn 🔒	MIT	Github 🔒	Rankactive 🔒	
Cassandra2	Native	МІТ	Github 🔒		
Cassandra 🛃	Multicorn 🔒	PostgreSQL	Github 🔒		
ClickHouse 🔒	Multicorn 🔒	BSD	Github 🔒		READM

Q&A

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- React Native Engineer
- Sr. React Native Engineer
- API Engineer
- SDET Java Architect
- SDET Javascript Architect
- SDET Java
- SDET Javascript
- Sr. UX Researcher

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