

# The magic of network graphs: How P Shipy anables exploration of complex petwork

How R Shiny enables exploration of complex networks

Jane Li & Chris Samson Risk Analytics and Data Science | Immigration New Zealand R Exchange, May 2025





Why graphs

ETL process

Demo

App schematic

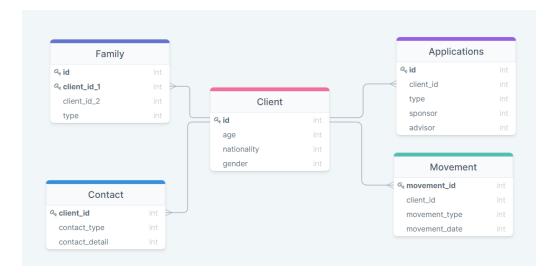
Deployment

Lessons learned

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#### A paradigm shift in how to think about data

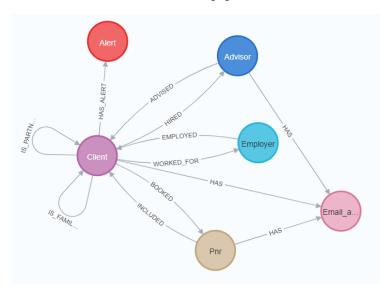
#### **Static approach**



Discrete data, as represented in a relational database

Tables and keys

#### **Connected approach**



Data represented as network in a "graph"-structure

Nodes and Links (entities and relations)



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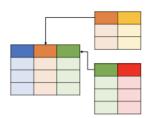
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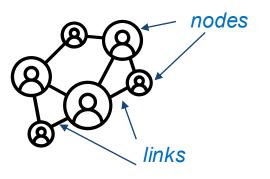
### Why Network Analytics and Graphs?

- Relational Databases: don't handle relationships well!
  - ✓ Not well adapted to interconnected semi-structure
  - ✓ Relationship queries **require painful SQL & joins**
  - ✓ Practically limited to three degrees of separation



A **graph** is a collection of **vertices** (**nodes**) representing entities and **edges** representing the **links** (**relationships**) among them.

- Graphs are optimised for data problems which involve many-to-many relationships:
  - ✓ Navigate deep hierarchies
  - ✓ Find **hidden connections** between distant items
  - ✓ Discover inter-relationships between items
  - ✓ Identifying key players and detecting communities





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### ETL process and Graph logic

#### Data collection



#### Data processing



#### Graph creation

- Using SQL connection to INZ server
- Package: DBI

- Data cleaning and deduplication
- Dataframes for nodes and links
- Properties for different types of node and link
  - "Client" Table

ID	Firs Name	Last Name	Gender	DoB	
102184	Marcelino	Russell	М	4/01/199	3
•••					4
WV22145615					
	> Node type "Client"			Marcelino R	Client_number: 102184 First_Name: Marcelino Last_Name: Russell Gender: M Date_of_Birth: 4/01/1993 age: 32

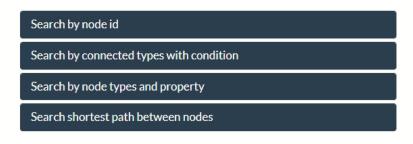
- Nodes and links saved as parquet files to use in Databricks
- Graph from nodes and links igraph::graph\_from\_data\_frame

#### Main R packages:

- igraph
- visNetwork
- shiny

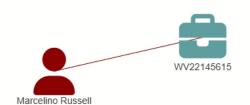


#### Network Analytics App





Client 102184, 1 hop(s)









App schematic Deployment Lessons learned Why graphs ETL process Demo **IN-CONFIDENCE Back End** Front End **Network Load Network Graph** Function **Network Replica** Block if > 100JS Nodes Staged Query Results Input Data **Download Buttons** (Variable) **Node Positions Remove Node Button Nodes And** Selected Node **Network Data Expand Node Button** Links **Graph Query** Icon **Search Buttons** Style Sheet **Functions Definitions** 



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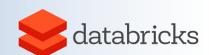
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### Deployment: App

#### Option

#### Local R







Third party managed cloud environment (e.g., posit)

Self-managed cloud app

(e.g., Azure App Services

Pros

- Easy to distribute
- Zero infrastructure required
- Good version control
- Notebook environment for development
- Easy set up with Snowpark Container Services
- Good access privilege control
- Good version control
- Often easy to use
- Environment for development & version control
- Fully modular and controllable

Cons

- A little technical to launch
- Version control not possible
- Users need admin rights
- Need to launch compute to activate
- Requires a containerized version of app
- No native notebook environment (Posit app available)
- An additional service to fit into instructed and security setup
- Vary in quality and features
- Requires skillset to set up and maintain
- Typically requires creation of a containerized version of app



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### Deployment: Data

Option	Pros	Cons	
File with App	<ul><li>Easy to distribute</li><li>Zero infrastructure required</li></ul>	<ul><li>Only works with small scale data</li><li>Difficult to update</li></ul>	
SQL	<ul> <li>Common database</li> <li>Can easily be integrated with existing cloud and on-premises architecture</li> </ul>	<ul> <li>Inefficient to query, particularly when traversing multiple edges</li> <li>Cannot be effectively used for graph specific algorithms (i.e. shortest paths)</li> <li>Does not scale well and cannot be effectively partitioned</li> </ul>	
Graph DB  ArangoDB	<ul> <li>Gold standard for storing graph data</li> <li>Efficient to query</li> <li>[Most] support graph algorithms</li> </ul>	<ul> <li>Almost always stand-alone tools that need to be procured and integrated into infrastructure</li> <li>No standard query language</li> </ul>	



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Why graphs ETL process

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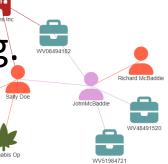
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#### Some lessons learned

- Engage with stakeholders early.
- Things grow fast.
- Sometimes small things make the largest difference.
- Deployment is a large effort.
- Building an app is only the beginning.





## THANK YOU!

