



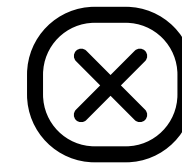
R CONFERENCE

ORGANIZED BY MALAYSIAN R USER GROUP (MYRUG)

Dr Tan Yan Bin

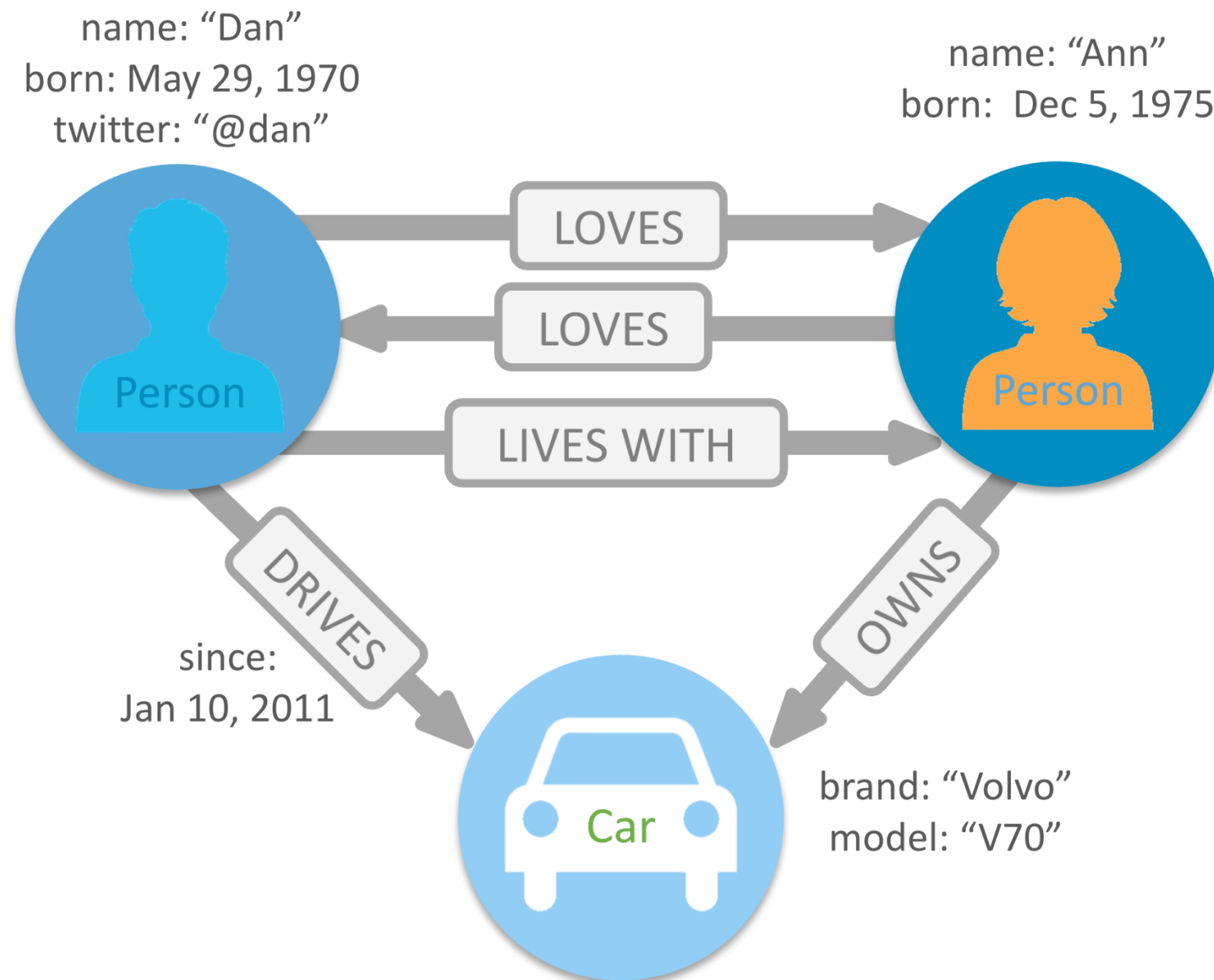
- Data Engineer ; Data Science trainer
- Specializations :
 - graph databases (Neo4j, TigerGraph)
 - SIEM (Splunk)
 - dashboards and reporting
 - machine learning





Neo4j in R

Graph Fundamentals



Node

- an entity in the graph

Relationship

- connection between nodes

Node Label

- grouping of similar nodes, e.g. Person, Car

Properties

- description (key-value pairs) of a node or relationship, e.g. name, born, brand, model

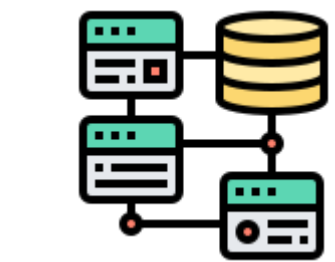
Why graph database?

- Nodes and relationships instead of table
- Flexible schema
- Deal with connected data, i.e. the 'connection' itself brings meaning
- Traverse multiple hops
- Discover patterns or hidden connections

Typical use cases :

- social media network, fraud detection, recommendation system, supply chain, ...

Conceptual comparison



RDBMS

Table



Node Label

Row



Node

Column



Properties

JOINS



Relationships



Graph
DB

Neo4j Cypher

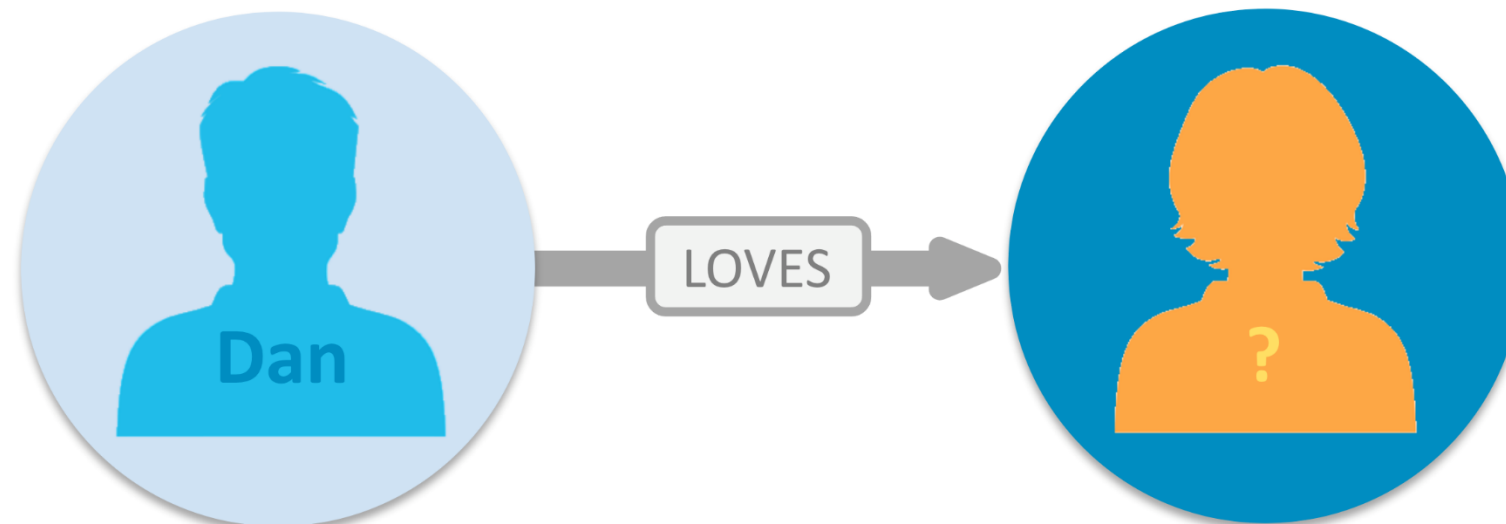


Diagram illustrating the components of the Cypher query:

- NODE**: The nodes in the query pattern.
- Relationship**: The relationship between the nodes.
- NODE**: The nodes in the query pattern.

Query: `MATCH (:Person { name:"Dan" }) -[:LOVES]-> (whom) RETURN whom`

- LABEL**: The label `Person`.
- PROPERTY**: The property `name` and its value `"Dan"`.
- VARIABLE**: The variable `whom`.



Connecting to R

Available libraries :

- [neo2R](#)
- [neo4jshell](#)
- [neo4r](#)



library(neo2R)

```
graph <- startGraph("http://localhost:7474",  
  username="neo4j",  
  password="1234567890",  
  importPath="C:/Users/Yvaine Tan/.Neo4jDesktop/relate-data/dbmss/dbms-8f046786-f028-4b8b-857b-5b3679e851f0/Import")
```

```
cypher(graph, 'MATCH (n) DETACH DELETE n;')  
cypher(graph, 'CALL apoc.schema.assert({},{});')
```

```
constraints <- c("CREATE CONSTRAINT ClientConstraint IF NOT EXISTS FOR (p:Client) REQUIRE p.id IS UNIQUE;",  
  "CREATE CONSTRAINT EmailConstraint IF NOT EXISTS FOR (p:Email) REQUIRE p.email IS UNIQUE;",  
  "CREATE CONSTRAINT PhoneConstraint IF NOT EXISTS FOR (p:Phone) REQUIRE p.phoneNumber IS UNIQUE;",  
  "CREATE CONSTRAINT SSNConstraint IF NOT EXISTS FOR (p:SSN) REQUIRE p.ssn IS UNIQUE;",  
  "CREATE CONSTRAINT MerchantConstraint IF NOT EXISTS FOR (p:Merchant) REQUIRE p.id IS UNIQUE;",  
  "CREATE CONSTRAINT BankConstraint IF NOT EXISTS FOR (p:Bank) REQUIRE p.id IS UNIQUE;",  
  "CREATE CONSTRAINT TransactionConstraint IF NOT EXISTS FOR (p:Transaction) REQUIRE p.globalStep IS UNIQUE;",  
  "CREATE CONSTRAINT DebitConstraint IF NOT EXISTS FOR (p:Transaction) REQUIRE p.globalStep IS UNIQUE;",  
  "CREATE CONSTRAINT CashInConstraint IF NOT EXISTS FOR (p:CashIn) REQUIRE p.globalStep IS UNIQUE;",  
  "CREATE CONSTRAINT CashOutConstraint IF NOT EXISTS FOR (p:CashOut) REQUIRE p.globalStep IS UNIQUE;",  
  "CREATE CONSTRAINT TransferConstraint IF NOT EXISTS FOR (p:Transfer) REQUIRE p.globalStep IS UNIQUE;",  
  "CREATE CONSTRAINT PaymentConstraint IF NOT EXISTS FOR (p:Payment) REQUIRE p.globalStep IS UNIQUE;",  
  "CREATE INDEX ClientNameIndex IF NOT EXISTS FOR (n:Client) ON (n.name)")
```

```
for (c in constraints) {  
  cypher(graph,c)  
}
```

```
clients <- read.csv("https://raw.githubusercontent.com/neo4j-field/graph-summit-apac-2023/main/data/clients.csv")
```

```
load_clients <- 'MERGE (c:Client { id: row.ID })  
  SET c.name = row.NAME  
  MERGE (p:Phone { phoneNumber: row.PHONENUMBER })  
  MERGE (c)-[:HAS_PHONE]->(p)  
  MERGE (s:SSN { ssn: row.SSN })  
  MERGE (c)-[:HAS_SSN]->(s)  
  MERGE (e:Email { email: row.EMAIL })  
  MERGE (c)-[:HAS_EMAIL]->(e);'  
import_from_df(graph=graph, cql=load_clients,toImport=clients)
```



Graph Data Science

Neo4j GDS

. . . . use graph algorithms, i.e. set of instructions that traverse across the graph to analyse relationships and patterns in connected data.



Graph Algorithm Family



**Pathfinding &
Search**



**Centrality &
Importance**



**Community
Detection**



**Supervised
Machine Learning**



**Heuristic Link
Prediction**



Similarity

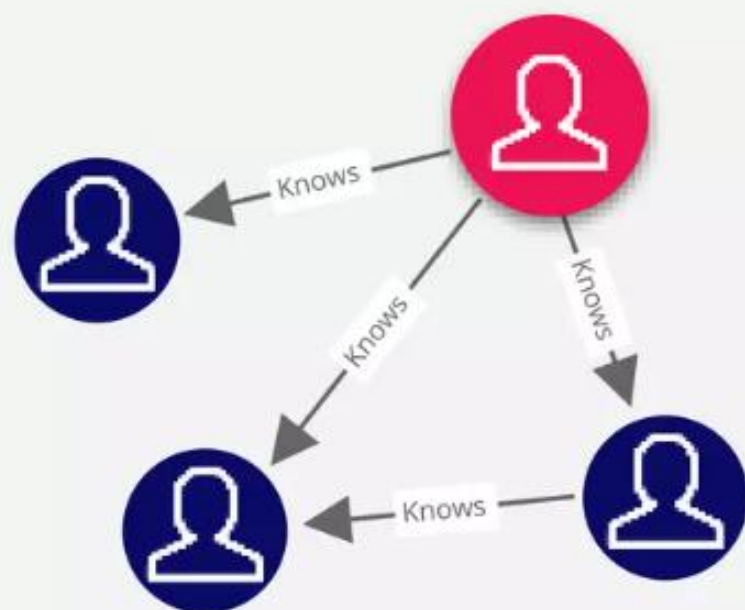


**Graph
Embeddings**



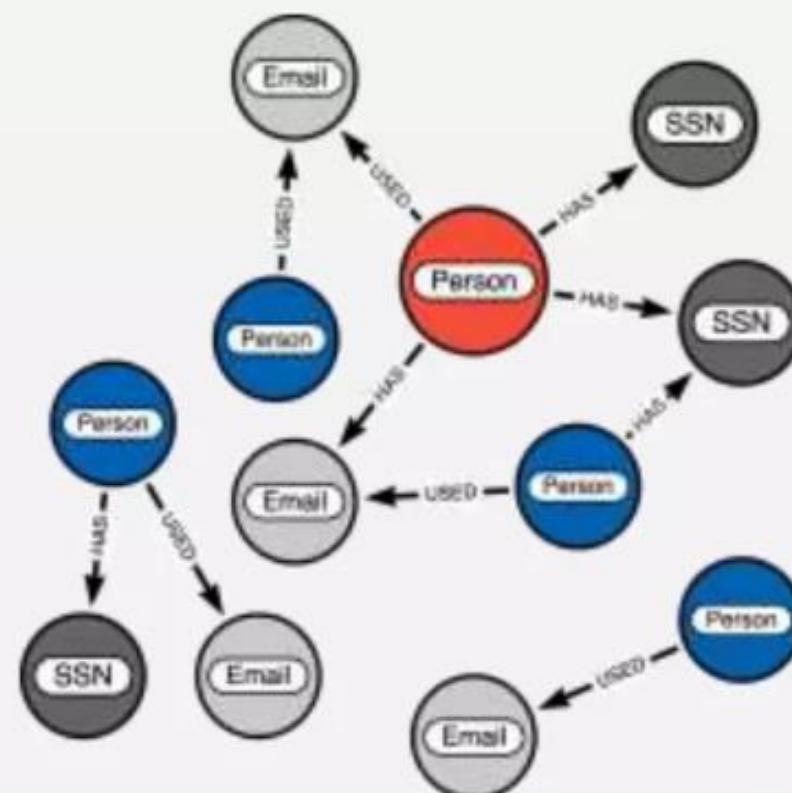
...and more

Graph Structure Improves Data Science Outcomes



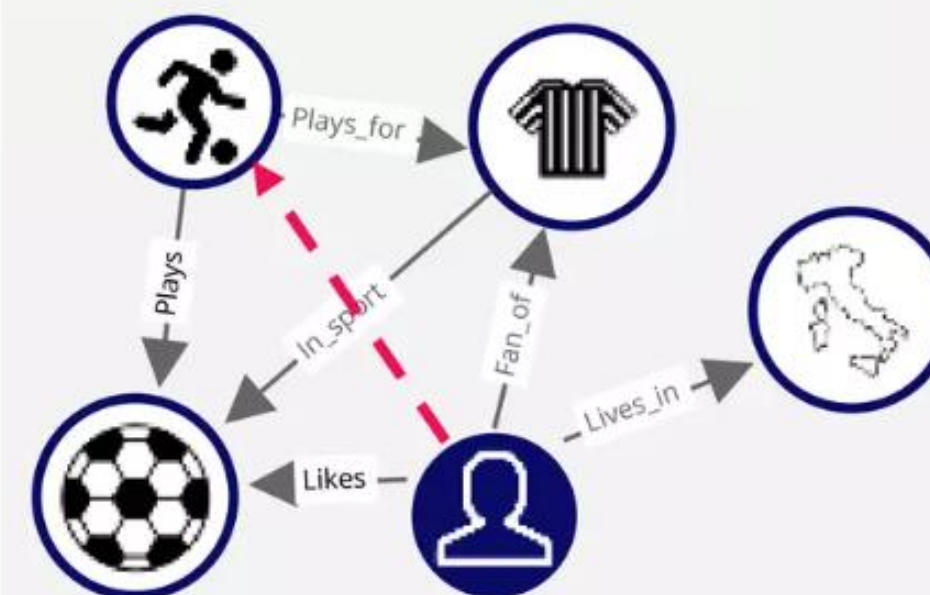
What's important? Prioritization

Who has the most connections?
Who has the highest page rank?
Who is an influencer?



What's unusual? Anomaly & Fraud Detection

Where is a community forming?
What are the group dynamics?
What's unusual about this data?



What's next? Predictions

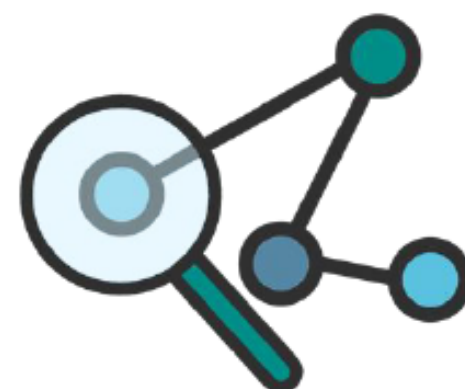
What's the most common path?
Who is in the same community?
What relationship will form?

Graph and Data Science

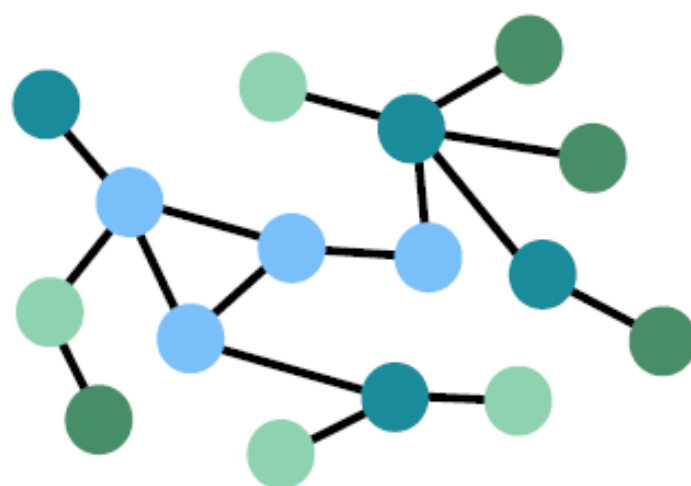
Graph Native Machine Learning



Graph Algorithms



Knowledge Graphs



Find the patterns you're looking for in connected data

Use unsupervised machine learning techniques to identify associations, anomalies, and trends.

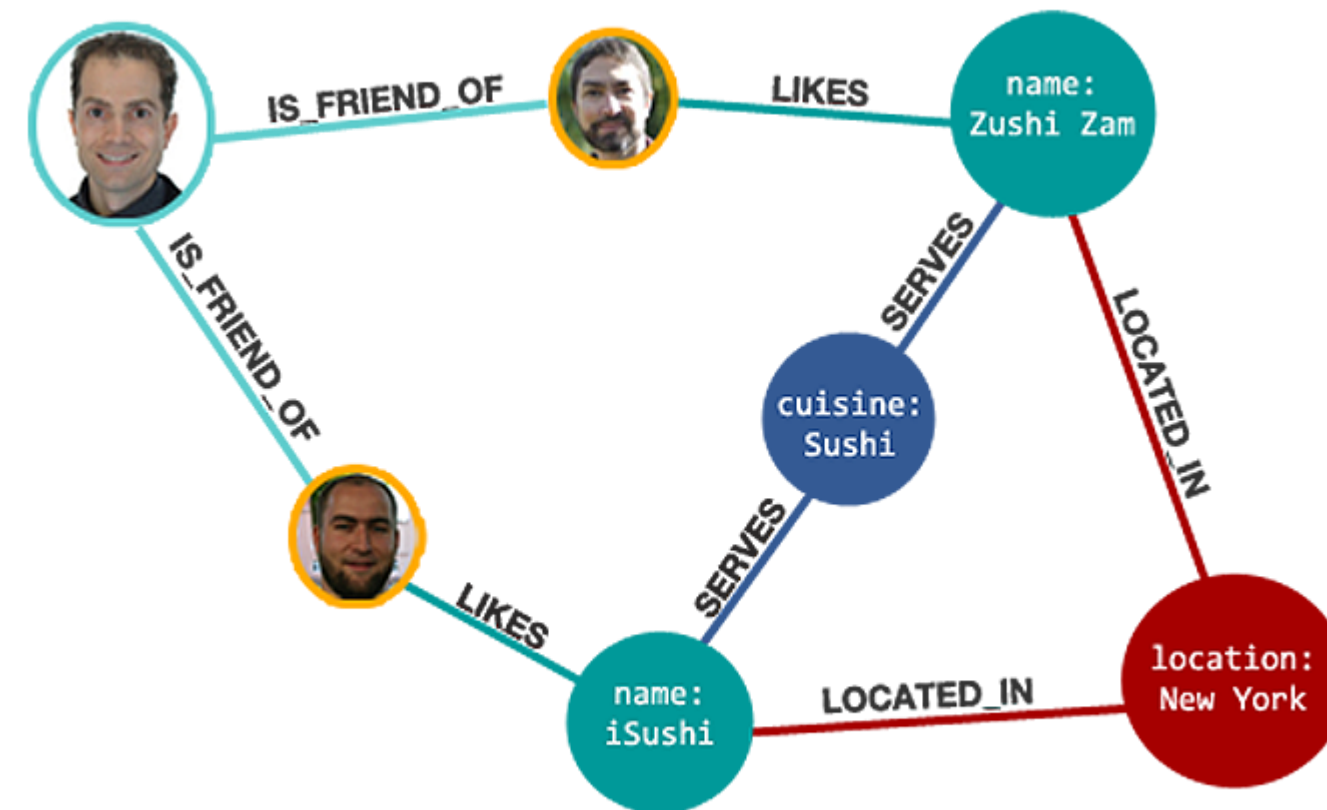
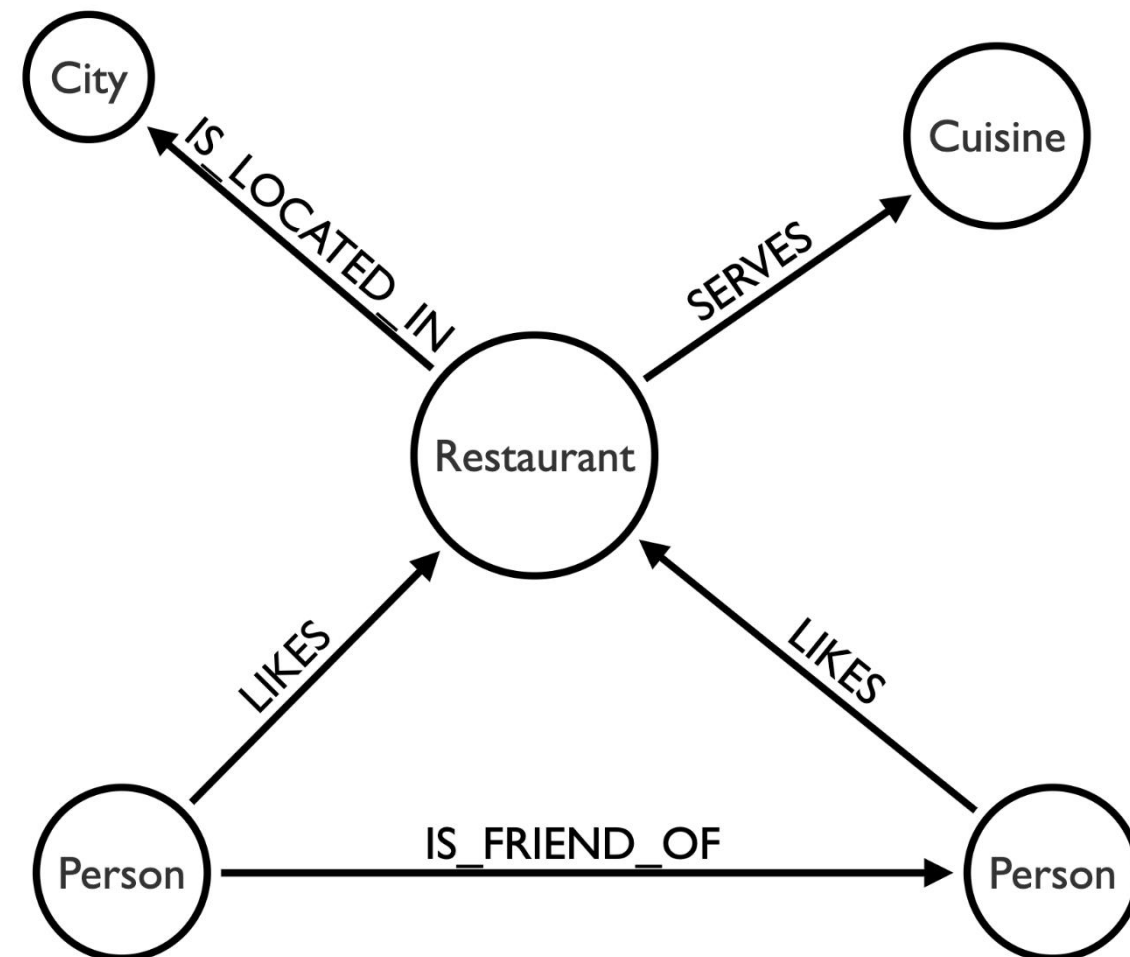
Use embeddings to learn the features in your graph that you don't even know are important yet.

Train in-graph supervised ML models to predict links, labels, and missing data.



Use case example :

Recommend Philip some sushi restaurants in New York that his friends like.



Algorithm :

1. Find Philip and his friends
2. Find restaurants that serve sushi in New York
3. Find restaurants that Philip's friends like



Read more

1. [neo4r User Guide](#)
2. [10 Things You Can Do with Cypher](#)
3. [Neo4j GraphGist – use cases and examples](#)



Thank you !

Q & A



MERGE (:R)-[:LOVES]->(: Neo4j)

