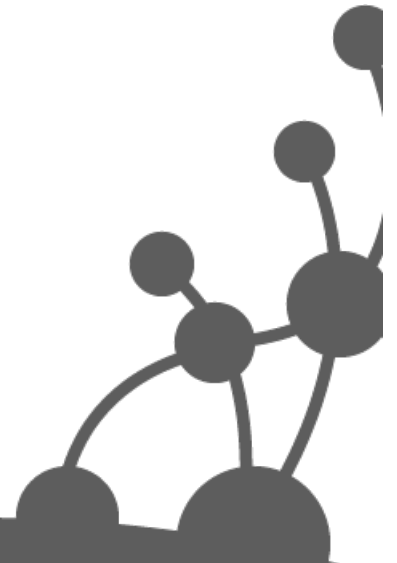


# A Little Graph Theory for the Busy Developer

Dr. Jim Webber

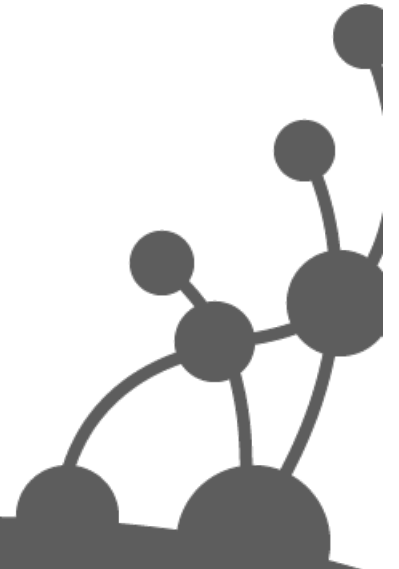
Chief Scientist, Neo Technology

@jimwebber

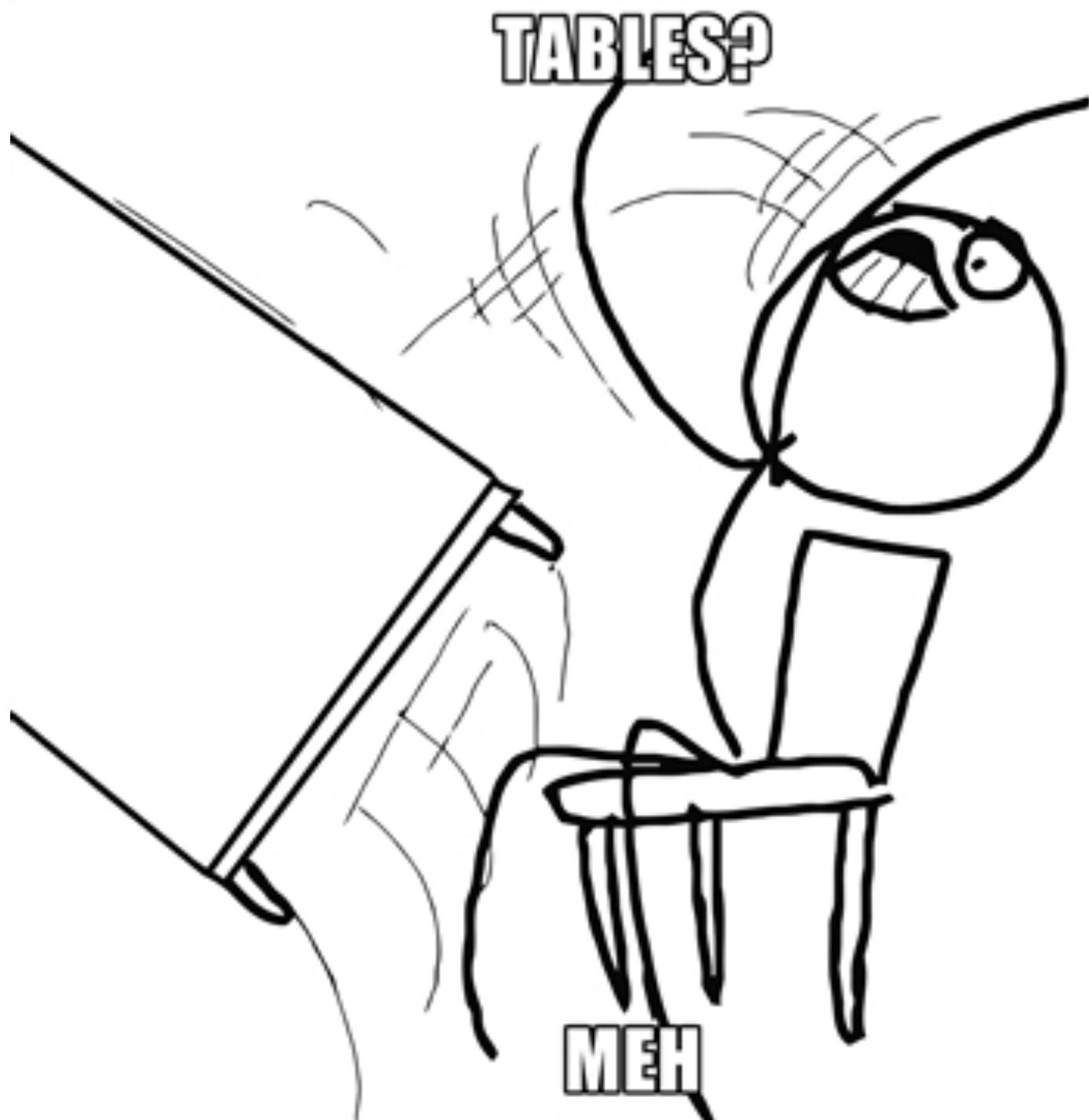


# Roadmap

- Imprisoned data
- Labeled Property Graph Model
  - And some cultural imperialism
- Graph theory
  - South East London
  - World War I
- Graph matching
  - Beer, nappies and Xbox
- End













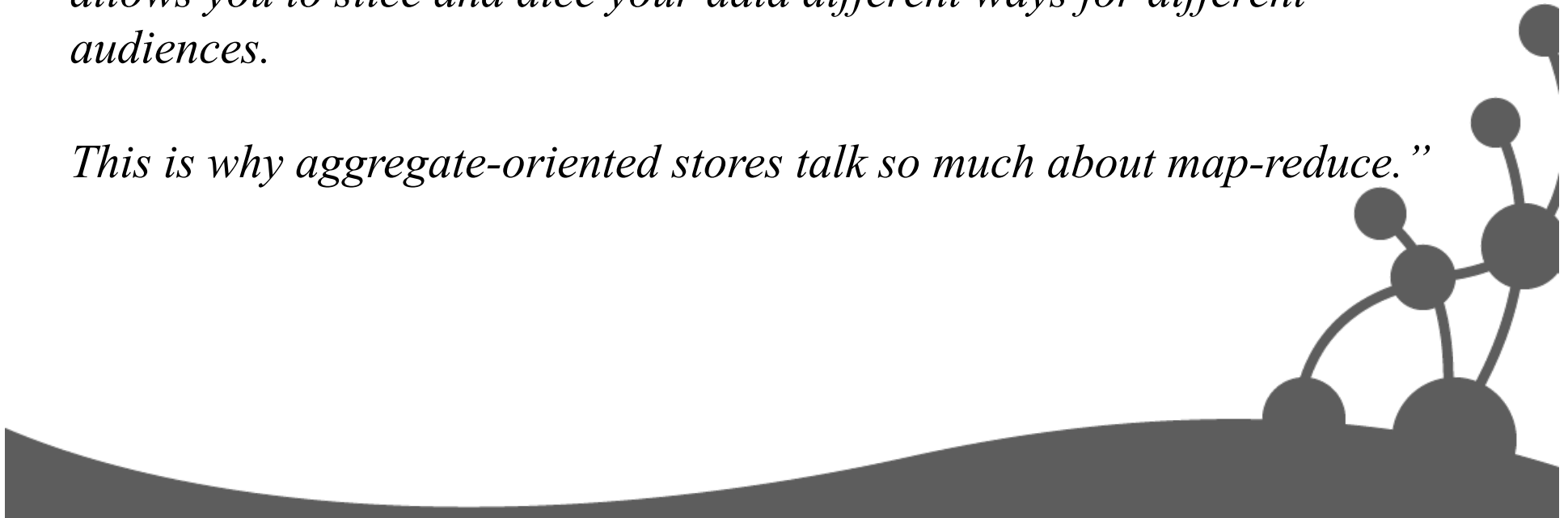


# Aggregate-Oriented Data

<http://martinfowler.com/bliki/AggregateOrientedDatabase.html>

*“There is a significant downside - the whole approach works really well when data access is aligned with the aggregates, but what if you want to look at the data in a different way? Order entry naturally stores orders as aggregates, but analyzing product sales cuts across the aggregate structure. The advantage of not using an aggregate structure in the database is that it allows you to slice and dice your data different ways for different audiences.*

*This is why aggregate-oriented stores talk so much about map-reduce.”*









*complexity = f(size, connectedness, uniformity)*





## DENORMALISE

*Aggregate data into documents*

## RICHER MODEL

*Connected structured data*



*Simple data model*

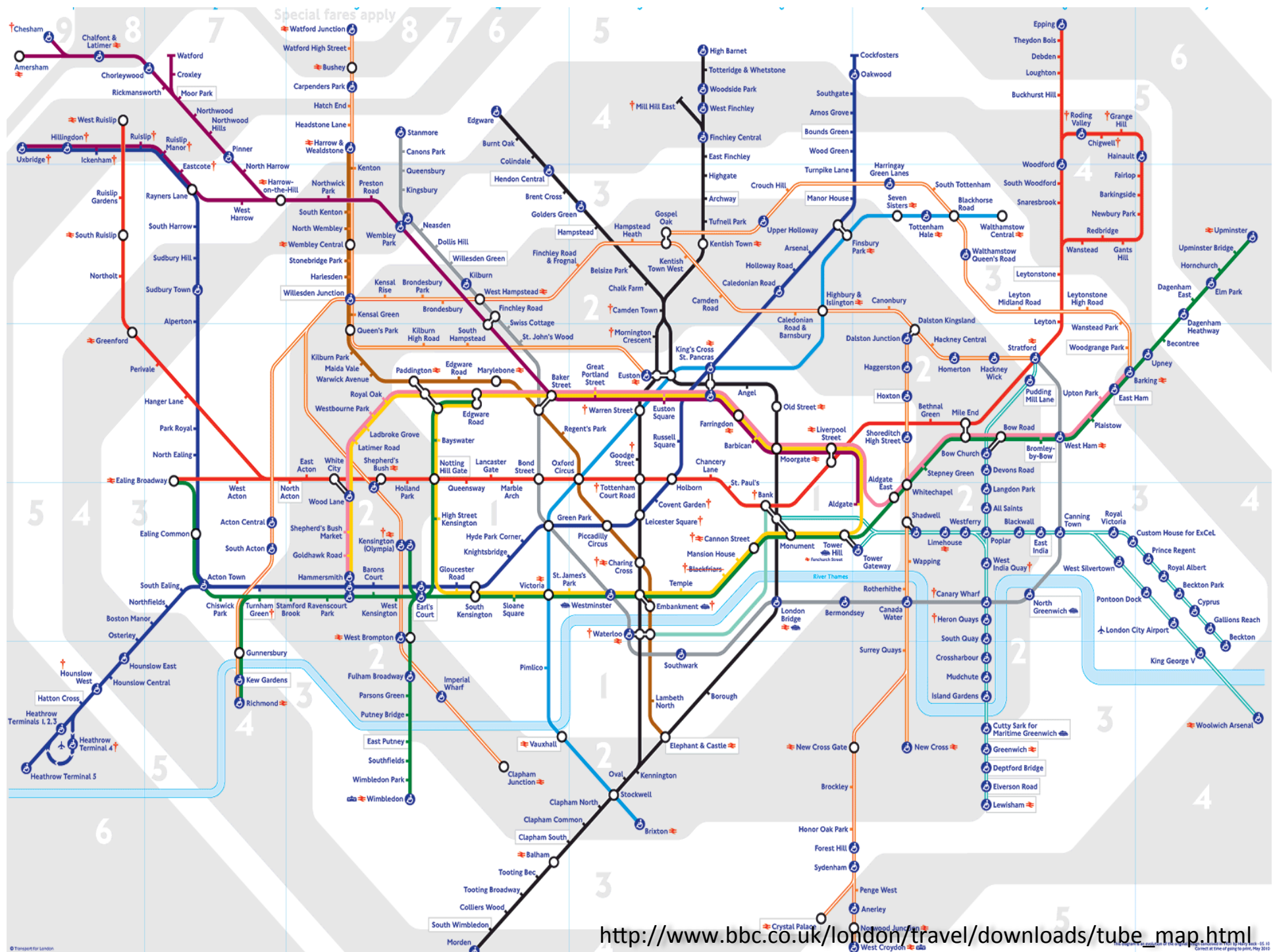
*Map-reduce friendly*

*Expressive power*

*Fast graph traversals*

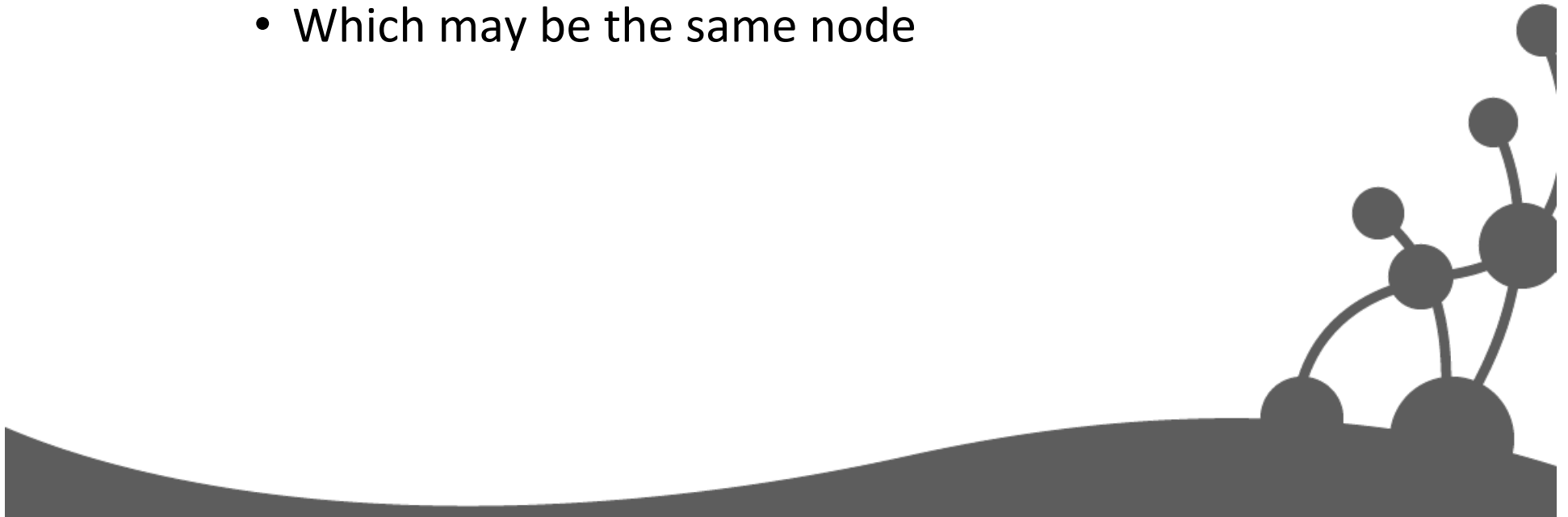


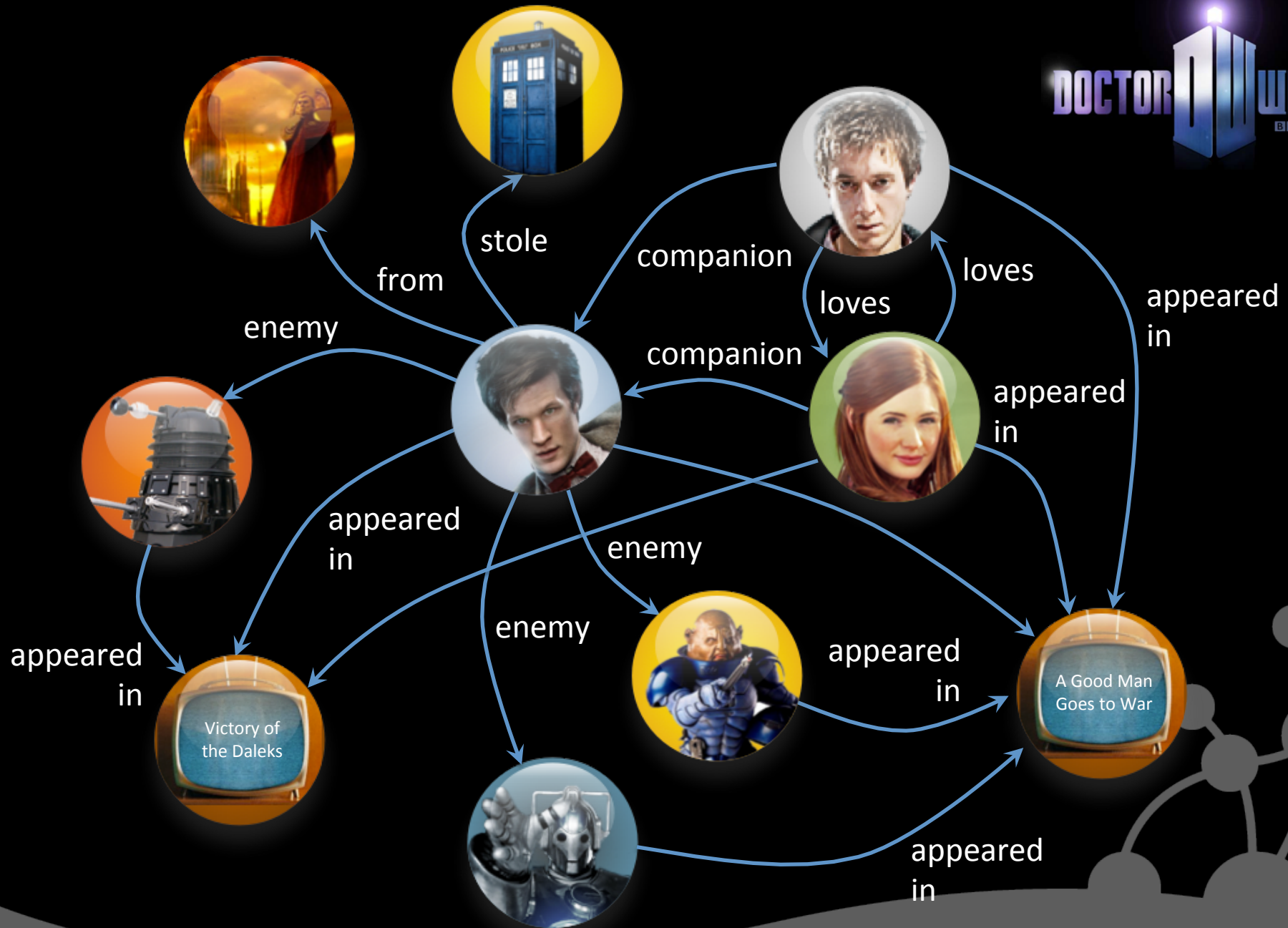


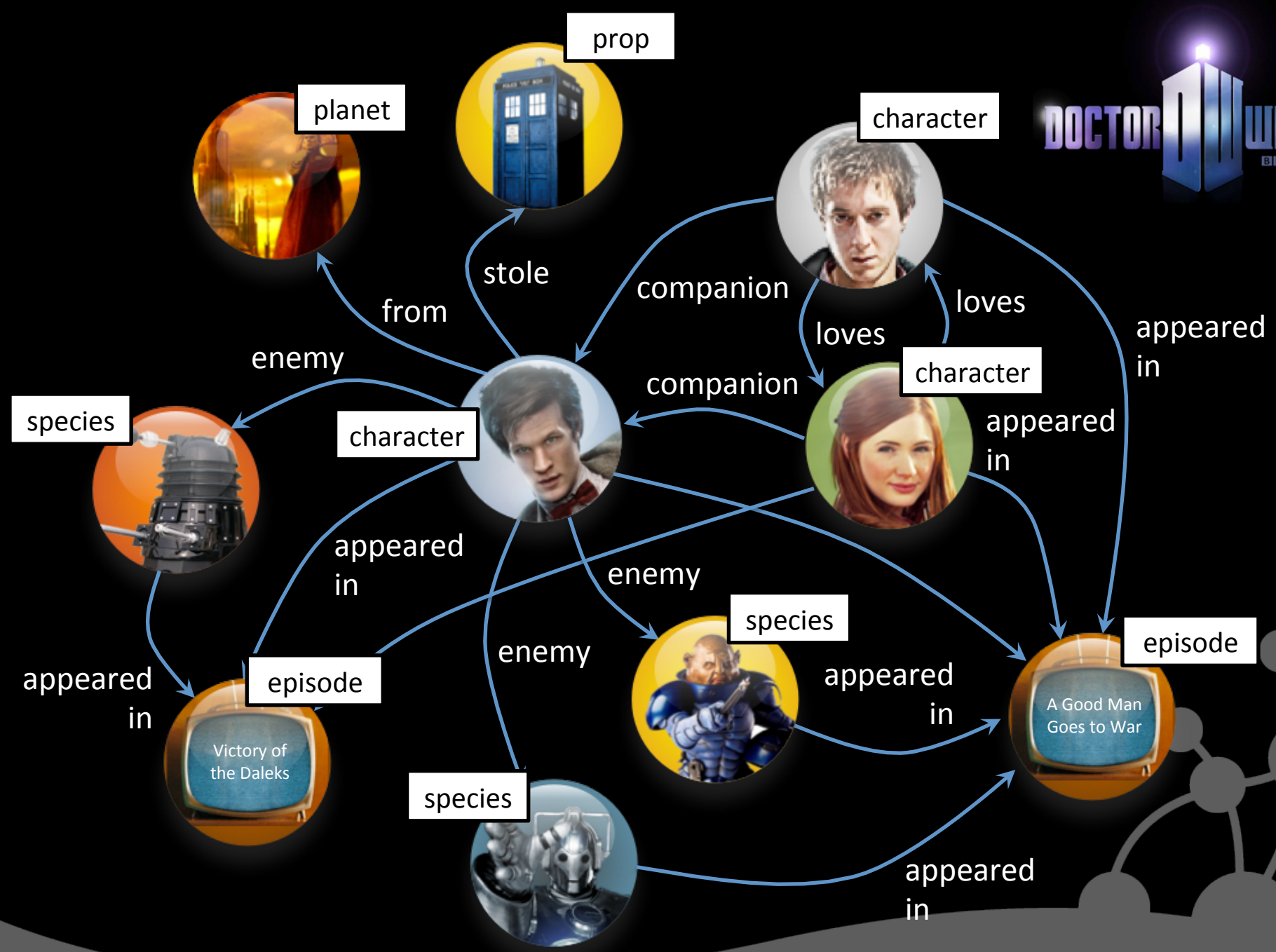


# Labeled Property graphs

- Property graph model:
  - Nodes with properties and labels
  - Named, directed relationships with properties
  - Relationships have exactly one start and end node
    - Which may be the same node







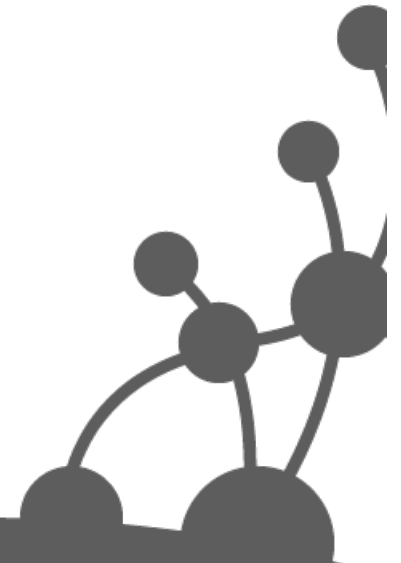




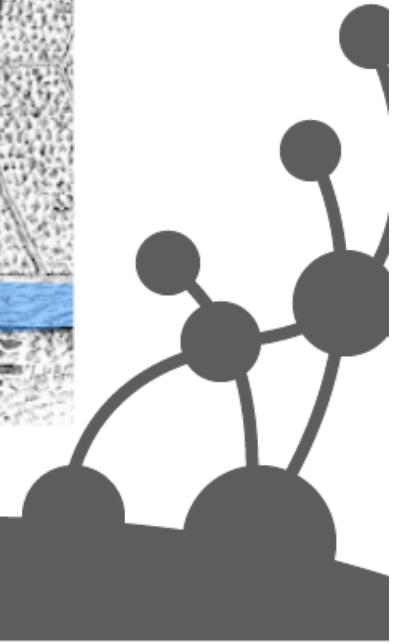
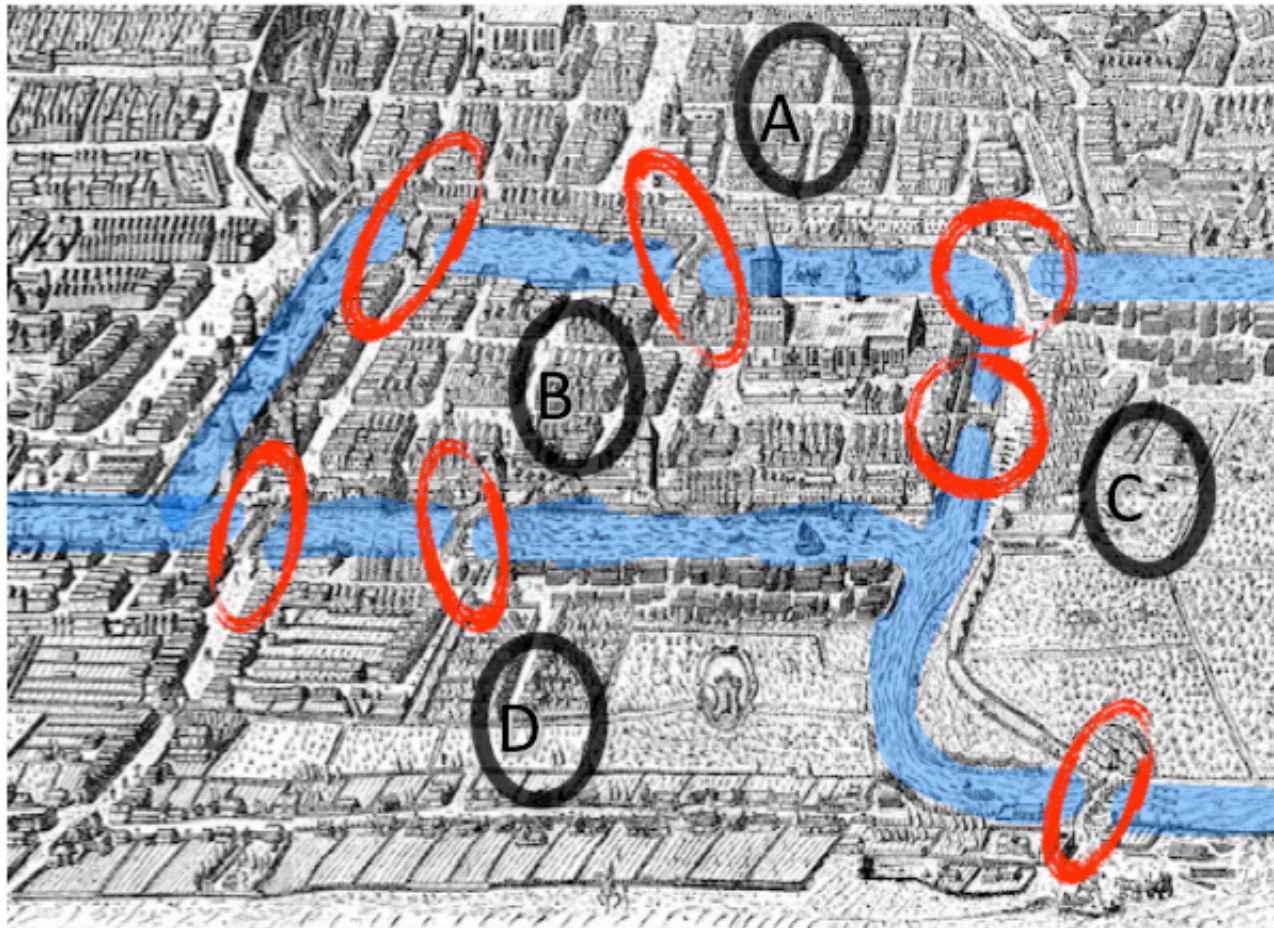


## Meet Leonhard Euler

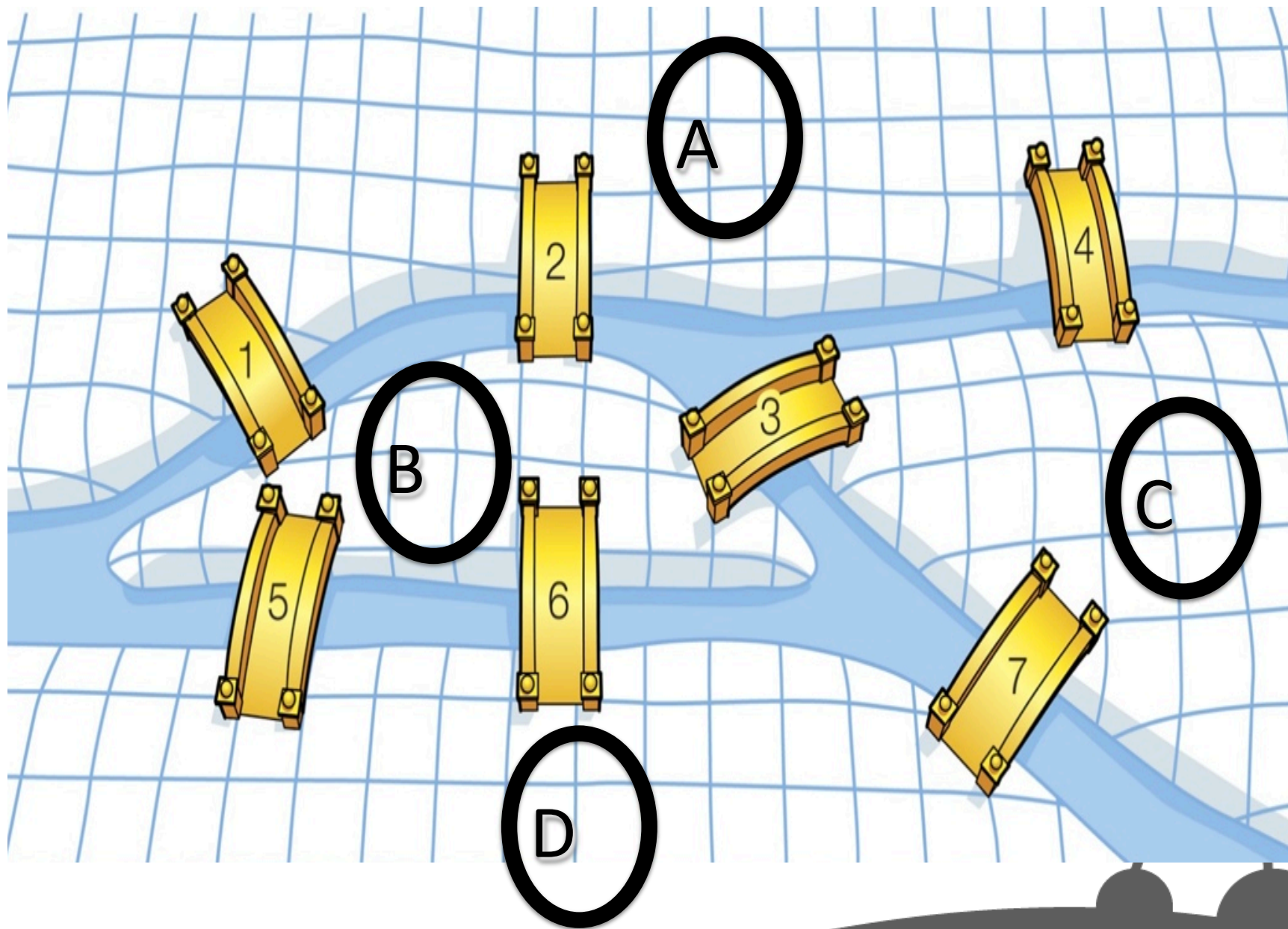
- Swiss mathematician
- Inventor of Graph Theory (1736)



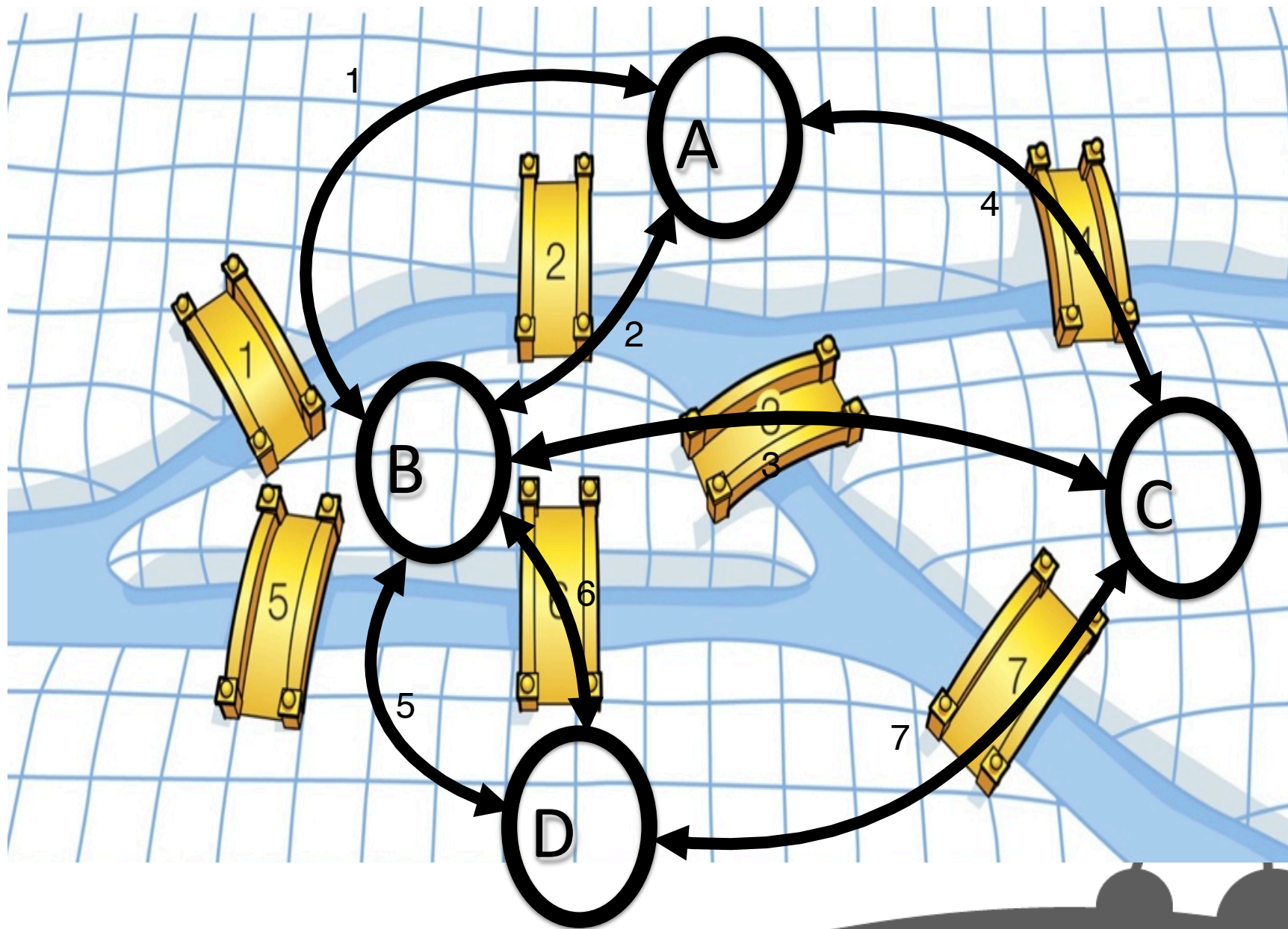
# Königsberg (Prussia) - 1736

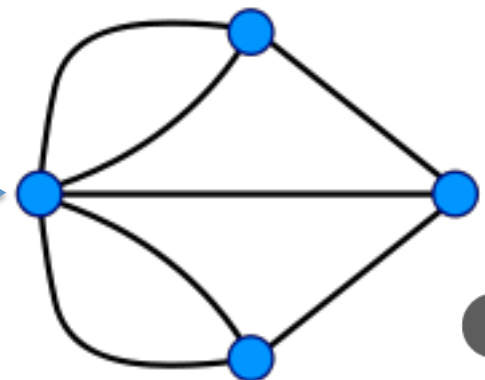
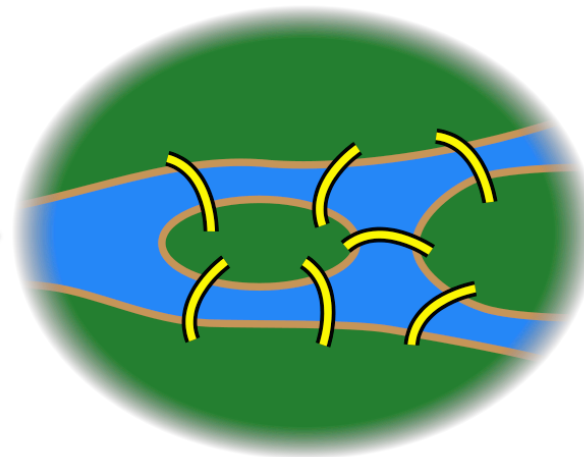
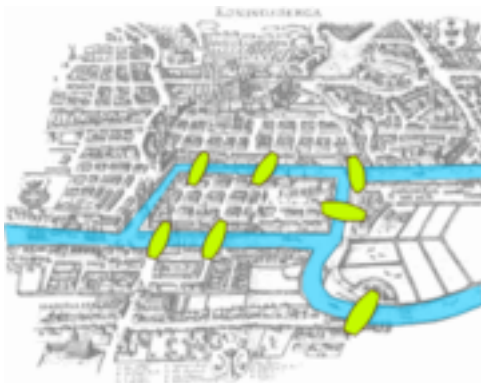














# DULWICH SUPERMARKET & OFF LICENCE

ENGLISH , TURKISH , GREEK , MEDITERRANEAN FOOD

18

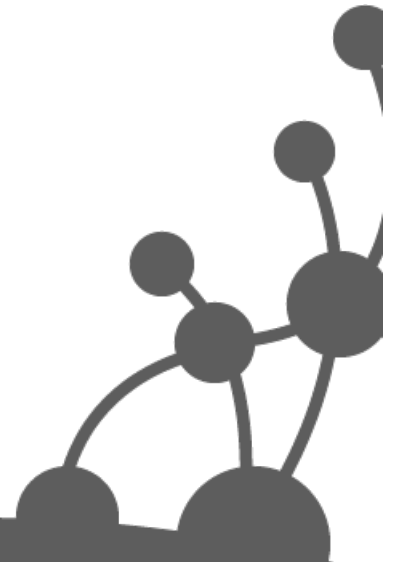
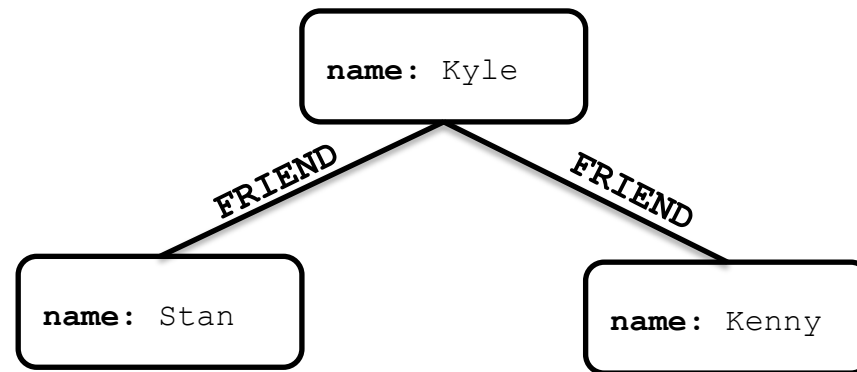
DELICATESSEN & ORGANICS

TEL : 020 8299 2214

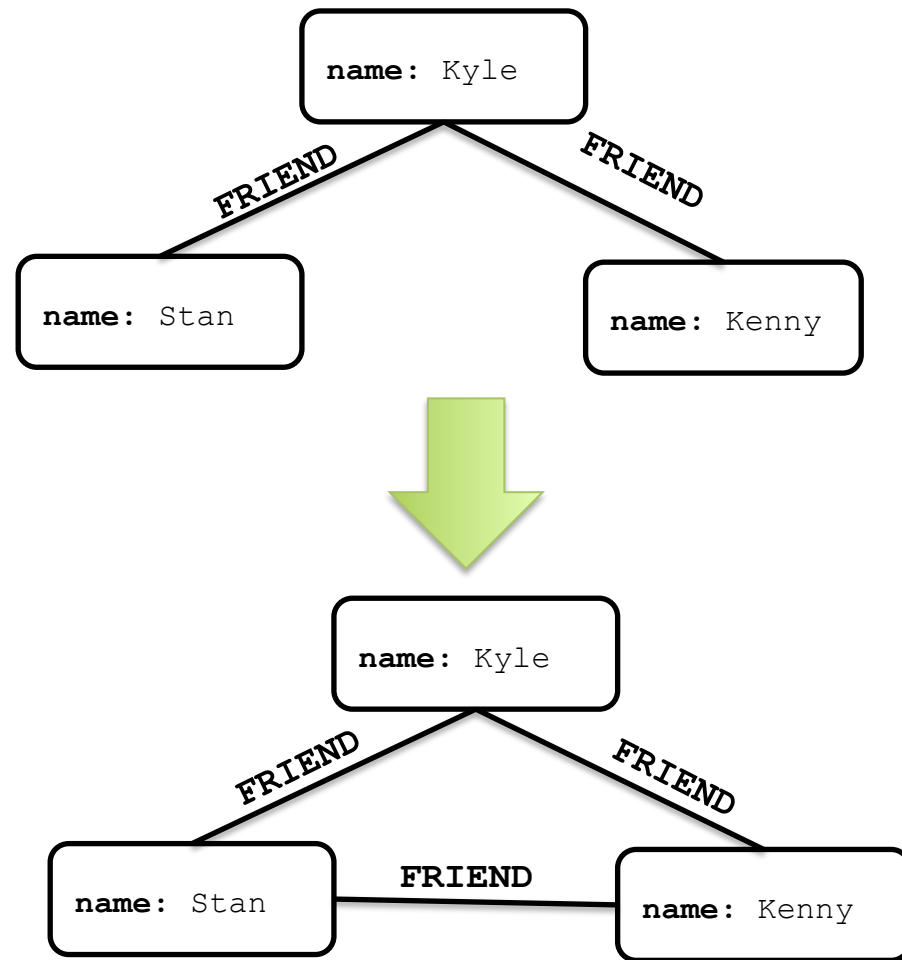
**FRESHLY CUT SANDWICHES · BILTONG**



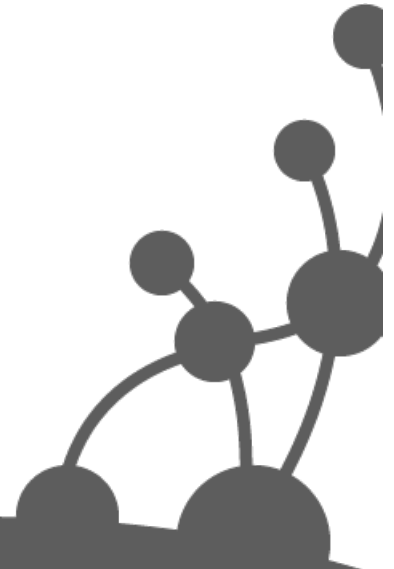
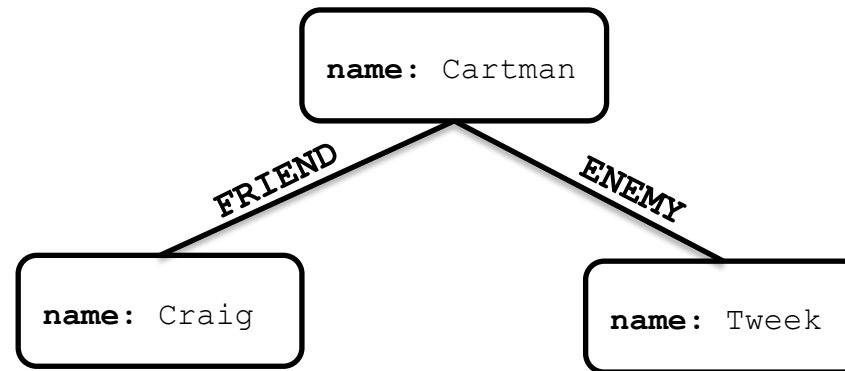
# Triadic Closure



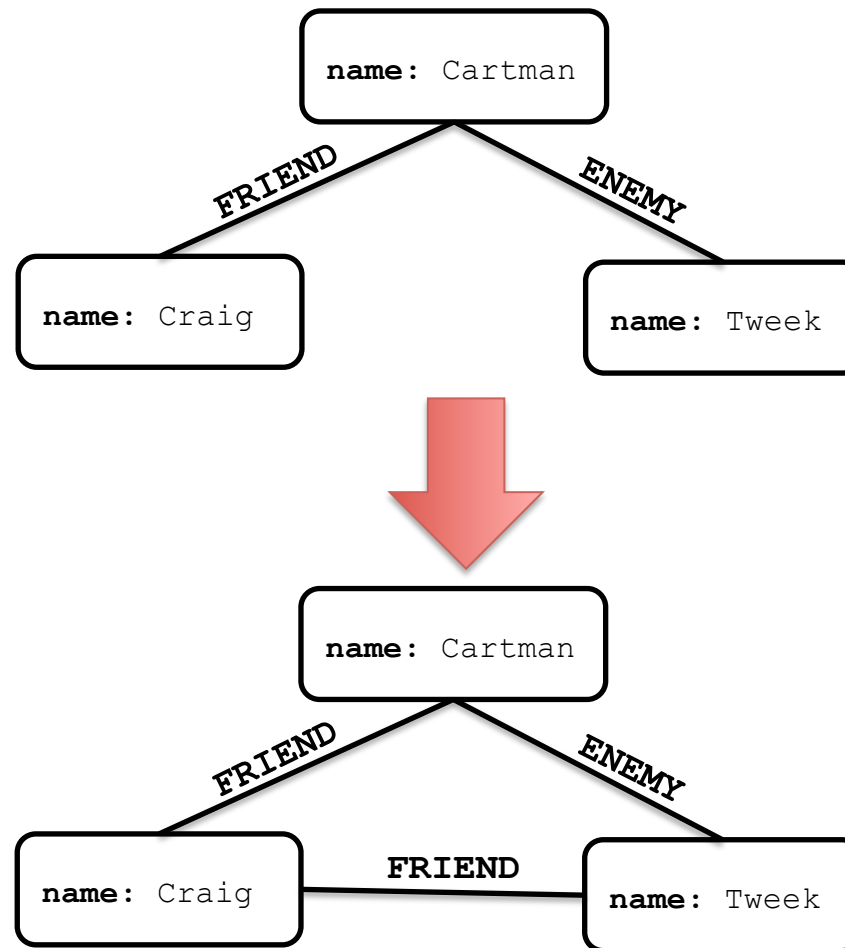
# Triadic Closure



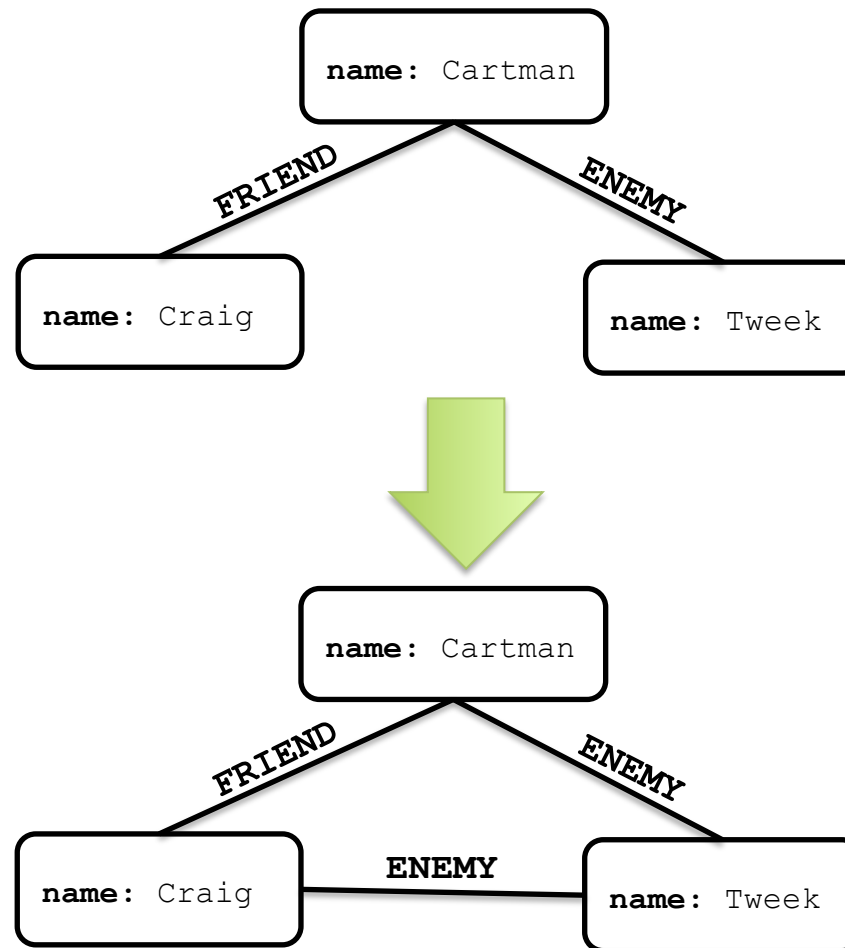
# Structural Balance



# Structural Balance

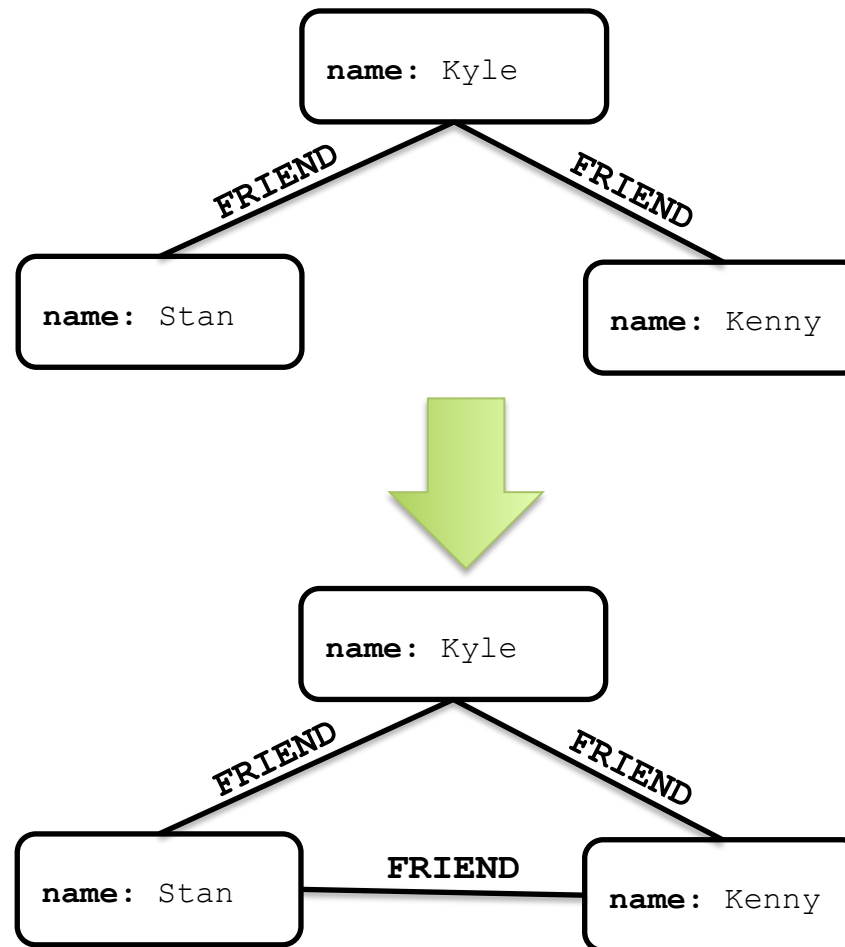


# Structural Balance



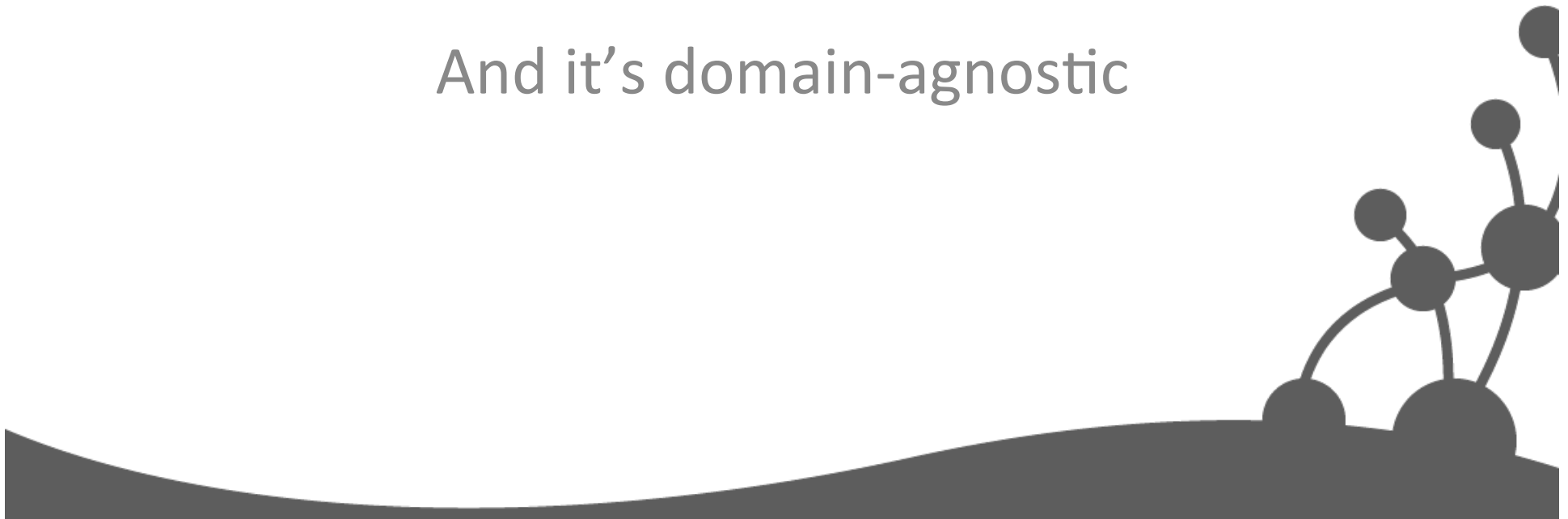


# Structural Balance

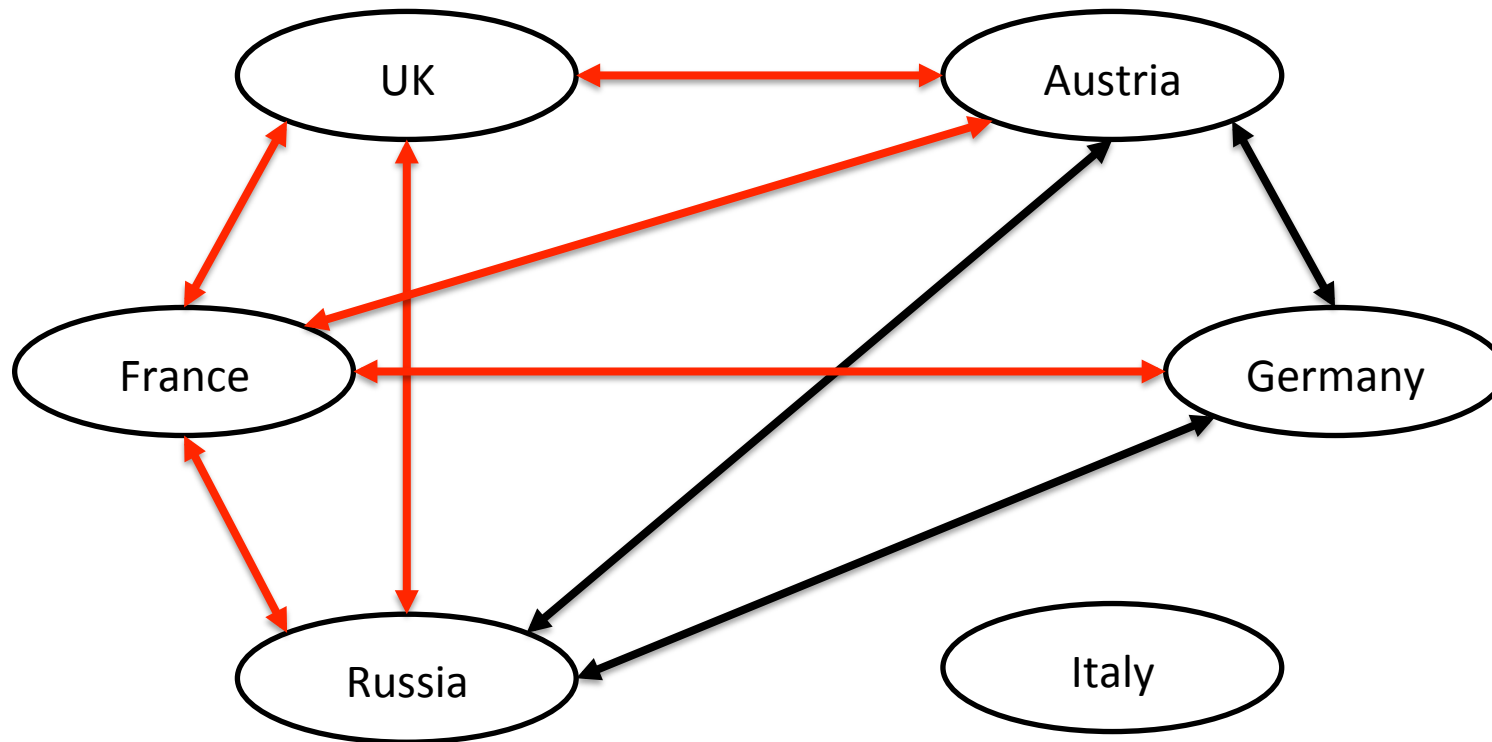


# Structural Balance is a *key* predictive technique

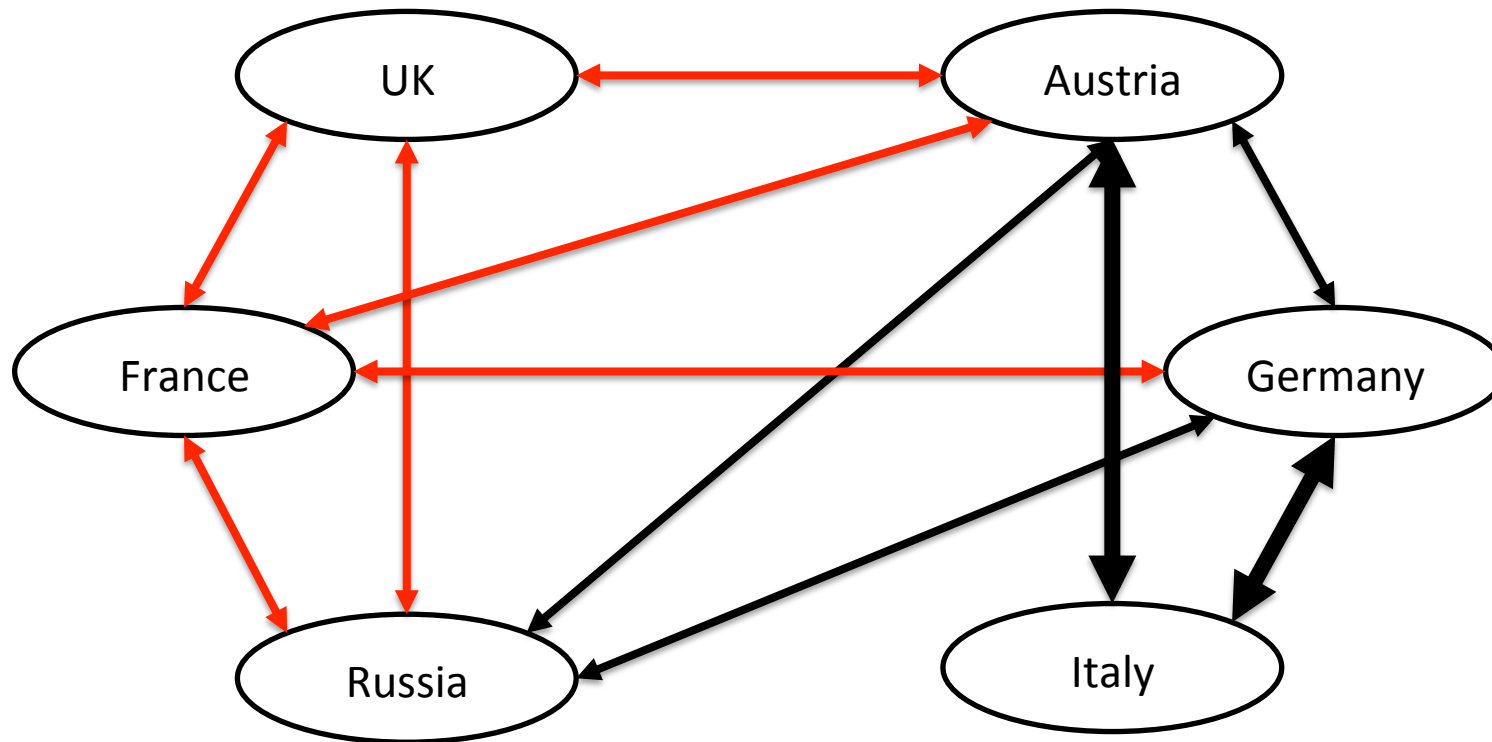
And it's domain-agnostic



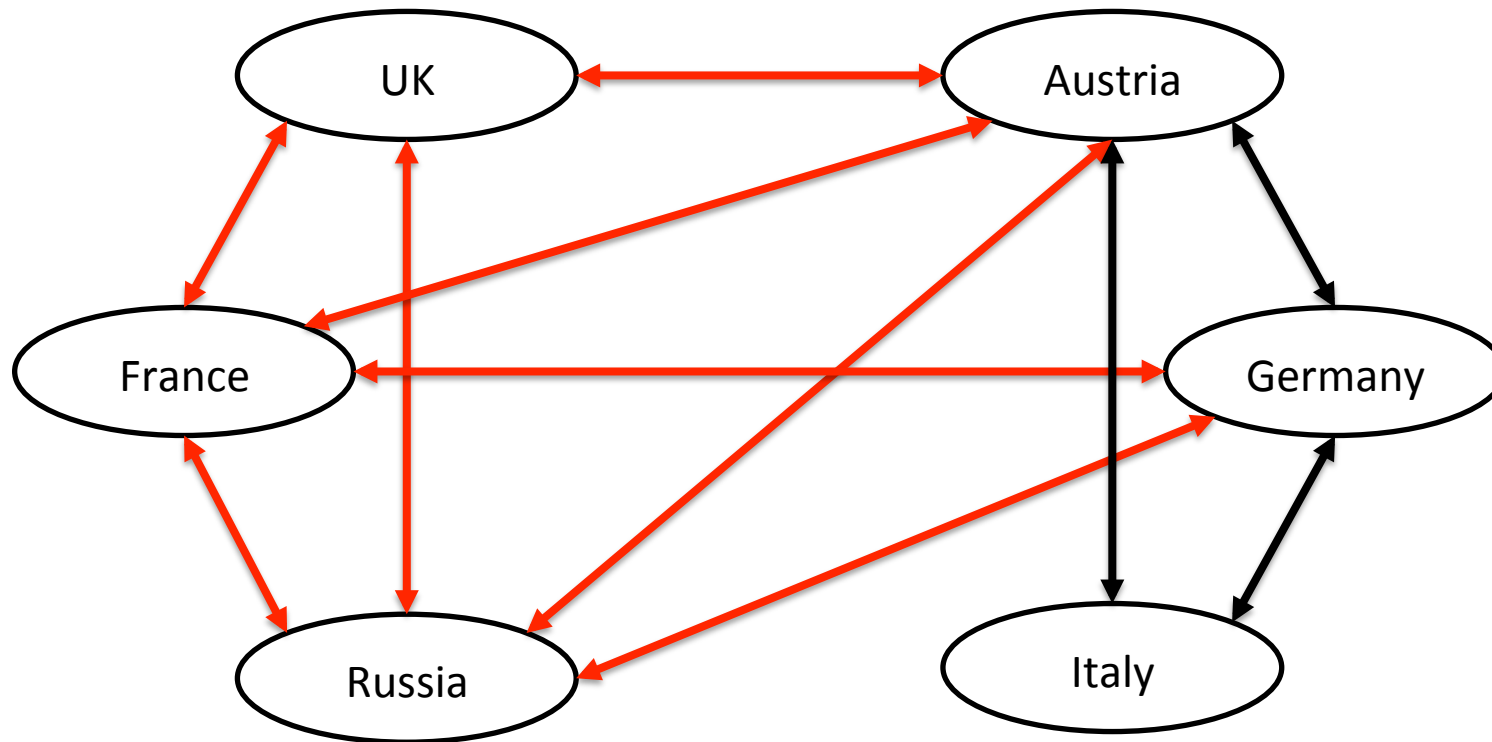
# Allies and Enemies



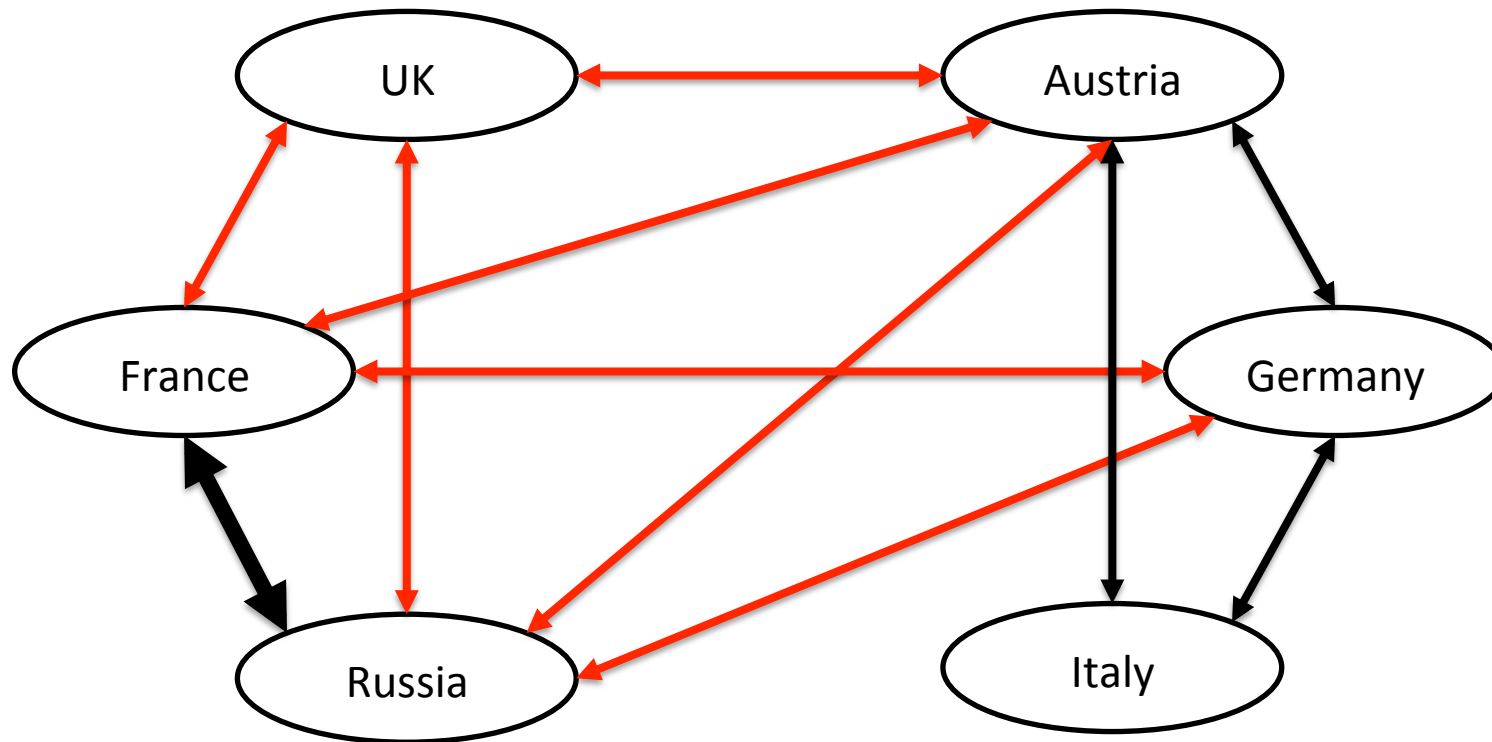
# Allies and Enemies



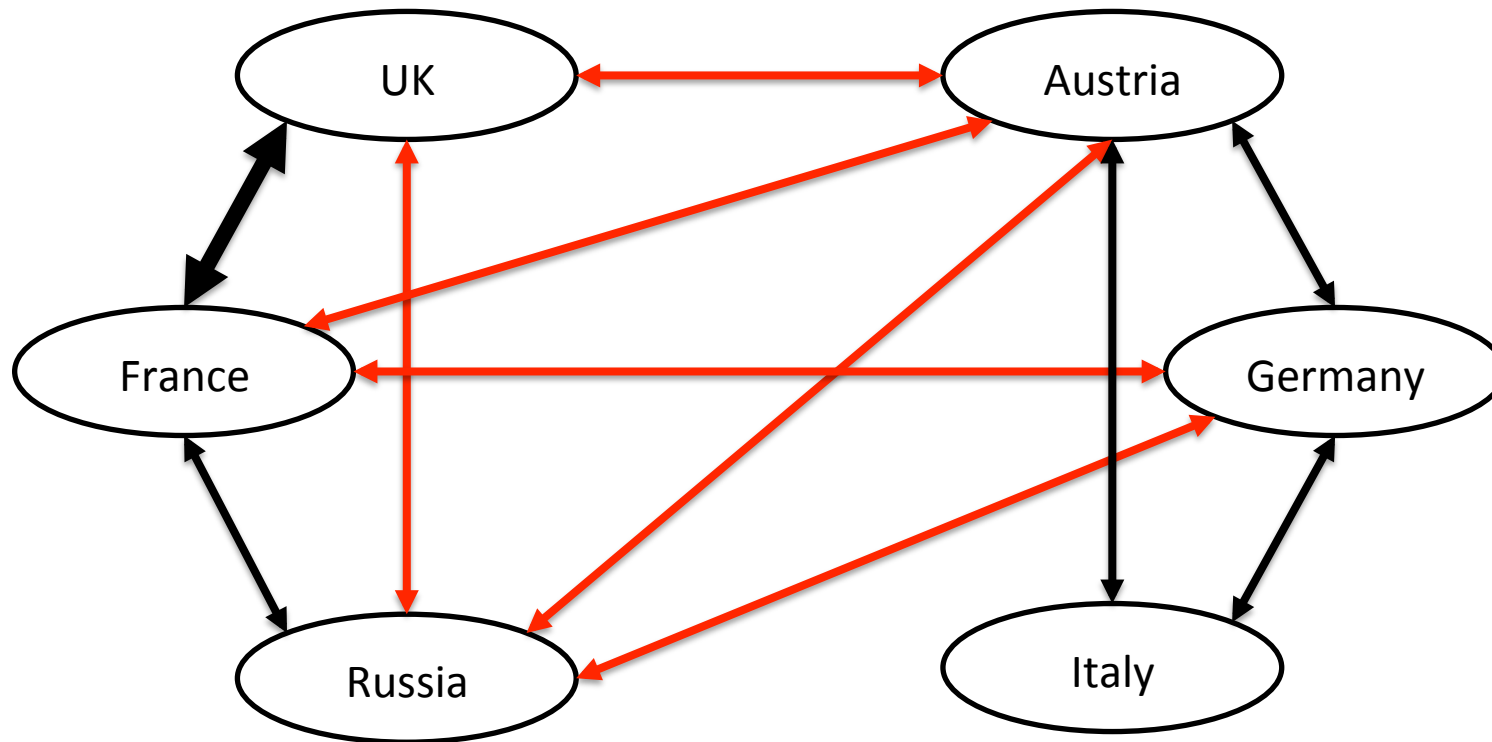
# Allies and Enemies



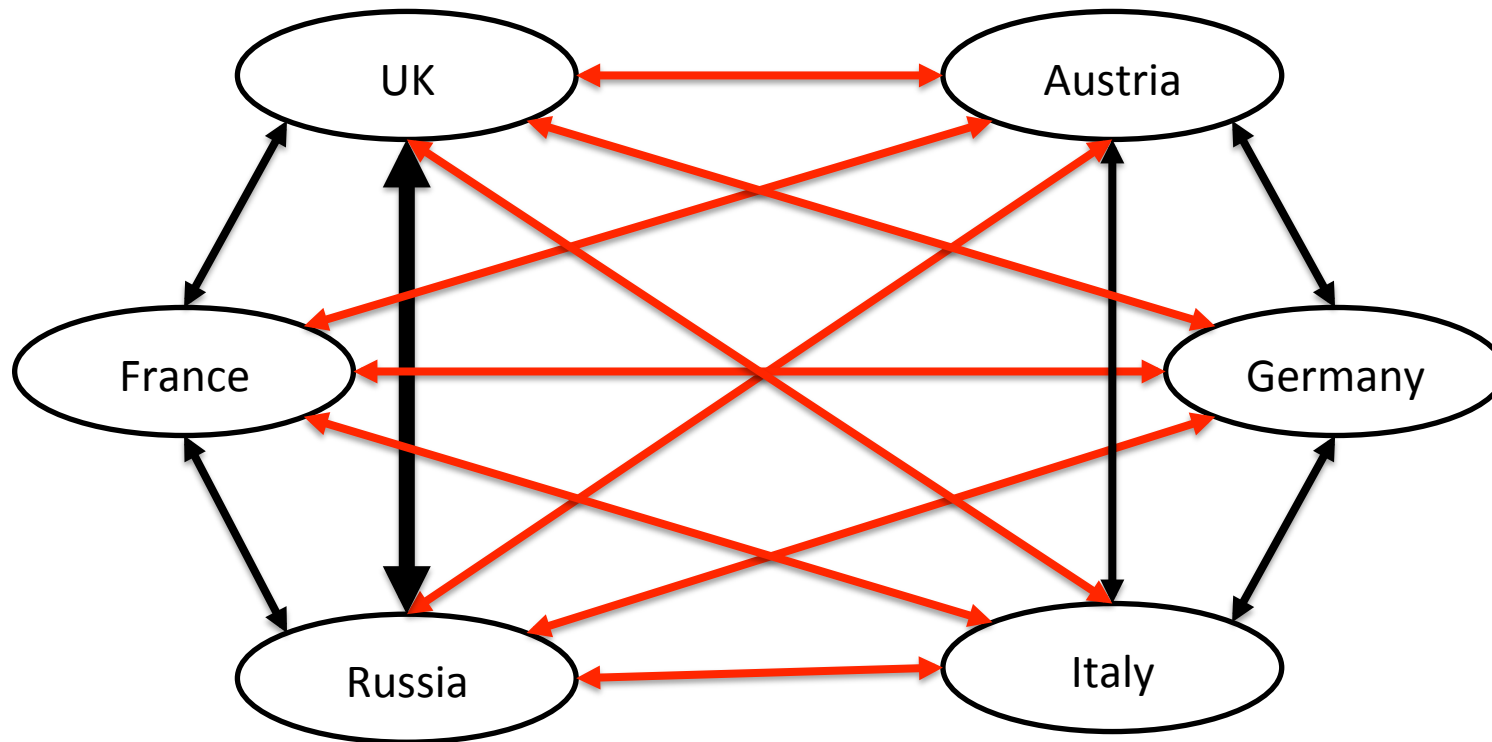
# Allies and Enemies



# Allies and Enemies



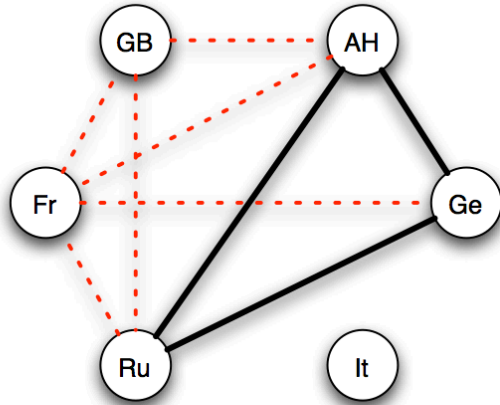
# Allies and Enemies



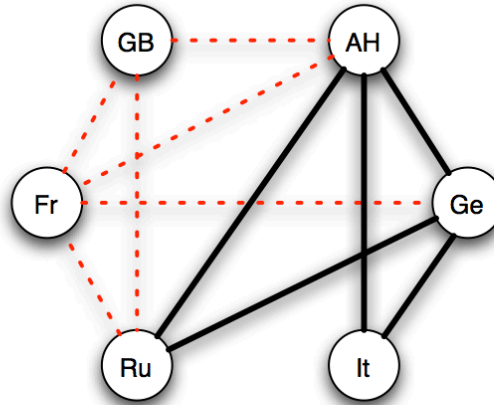


# Predicting WWI

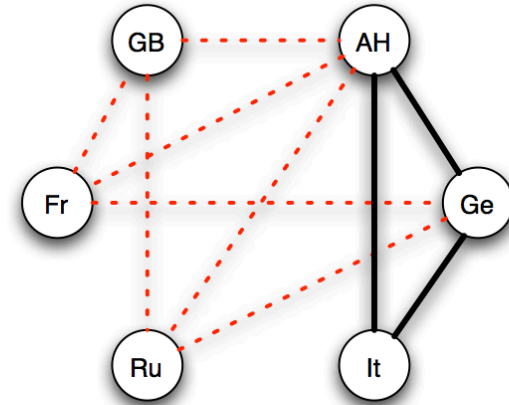
[Easley and Kleinberg]



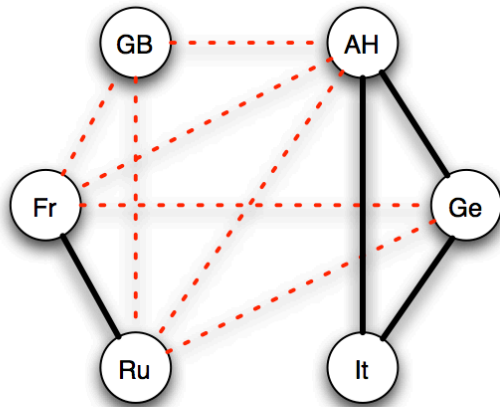
(a) *Three Emperors' League 1872–81*



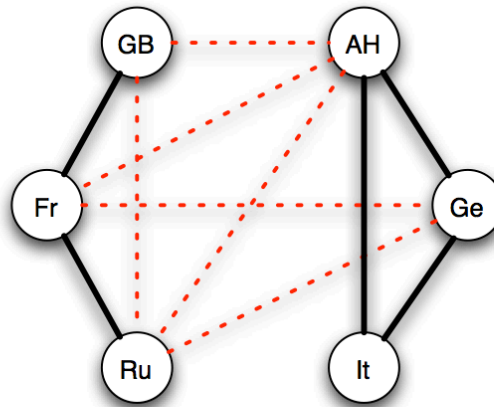
(b) *Triple Alliance 1882*



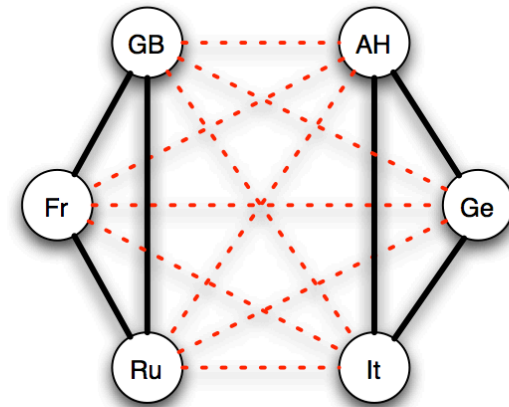
(c) *German-Russian Lapse 1890*



(d) *French-Russian Alliance 1891–94*



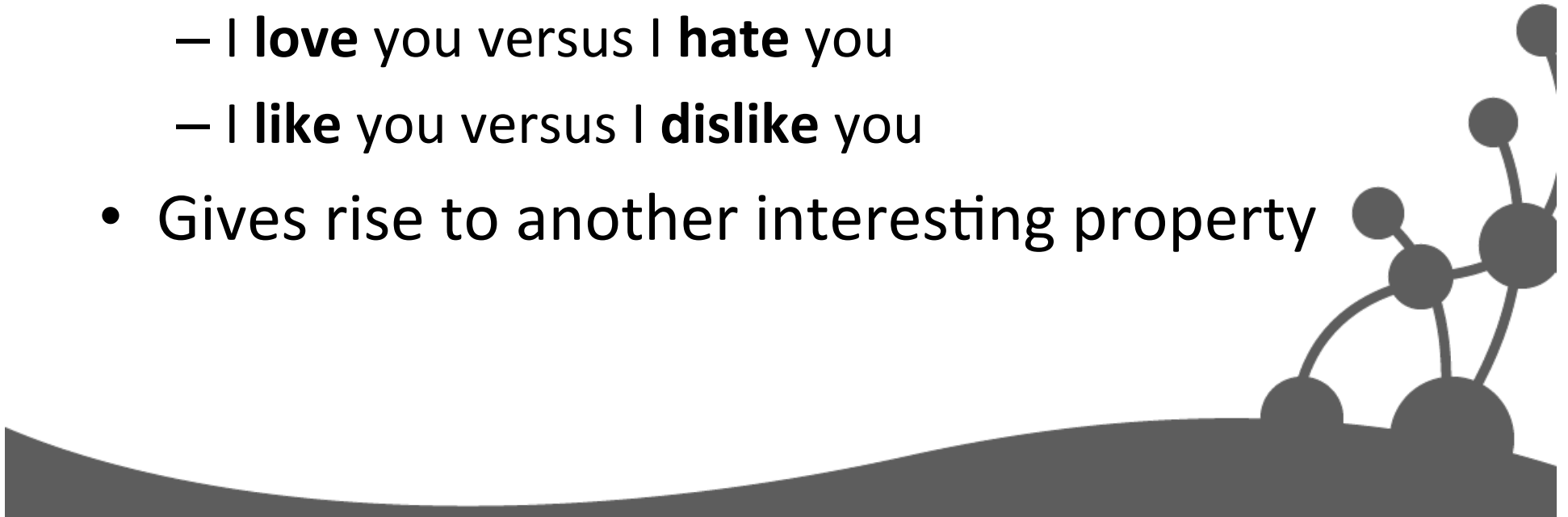
(e) *Entente Cordiale 1904*



(f) *British Russian Alliance 1907*

# Strength

- Relationships can have strength as well as sentiment
- This gives us another dimension to consider in our triangles
  - I **love** you versus I **hate** you
  - I **like** you versus I **dislike** you
- Gives rise to another interesting property



# Strong Triadic Closure Property

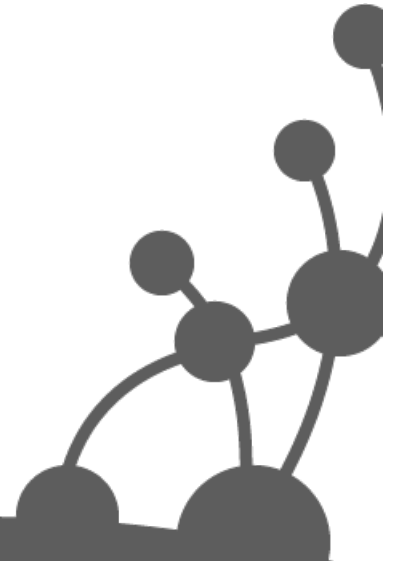
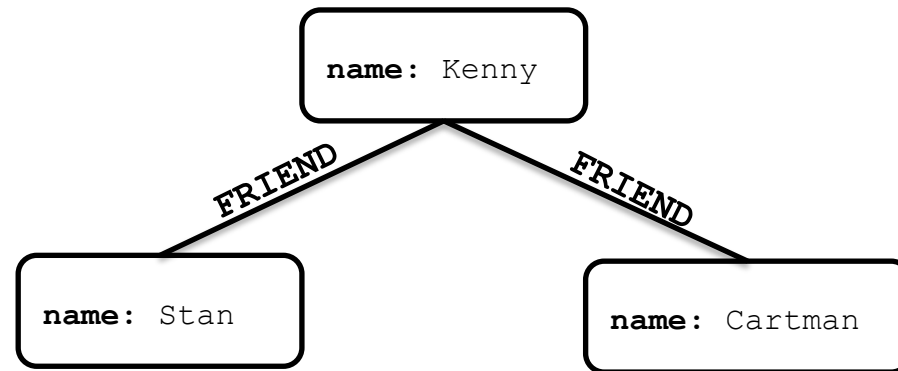
*It if a node has strong relationships to two neighbours, then these neighbours must have at least a weak relationship between them.*

*[Wikipedia]*



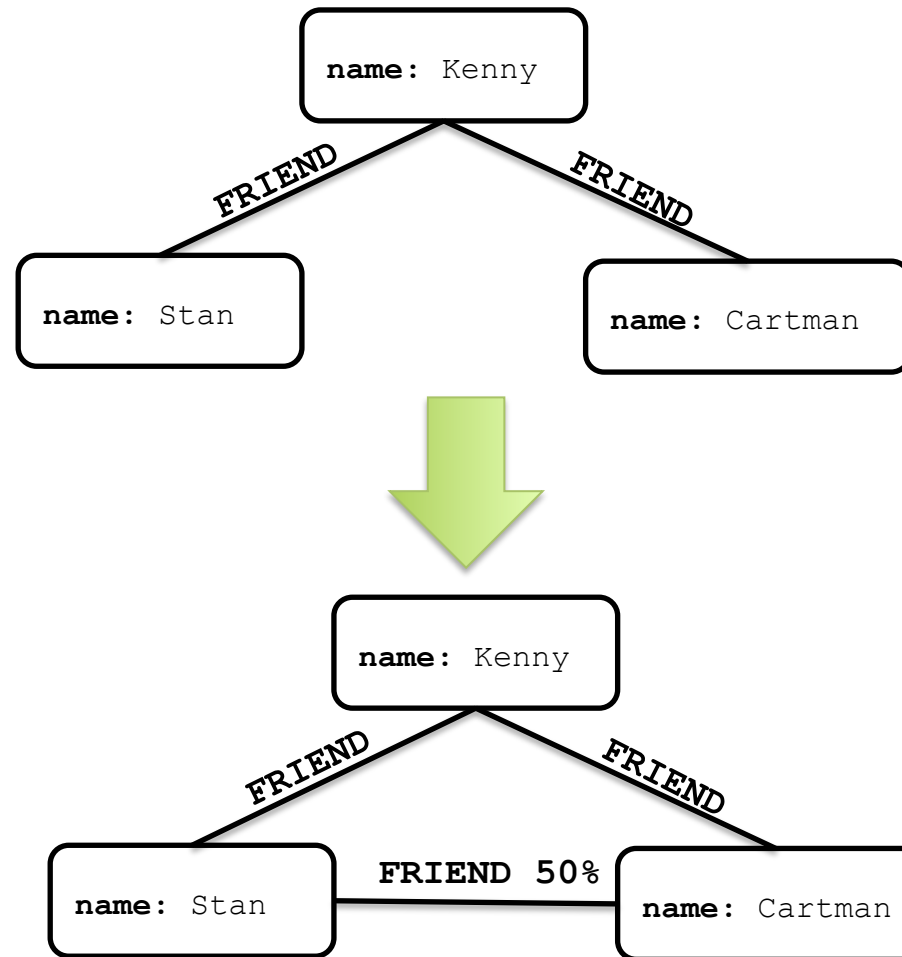
# Triadic Closure

(weak relationship)



# Triadic Closure

(weak relationship)



# Weak relationships

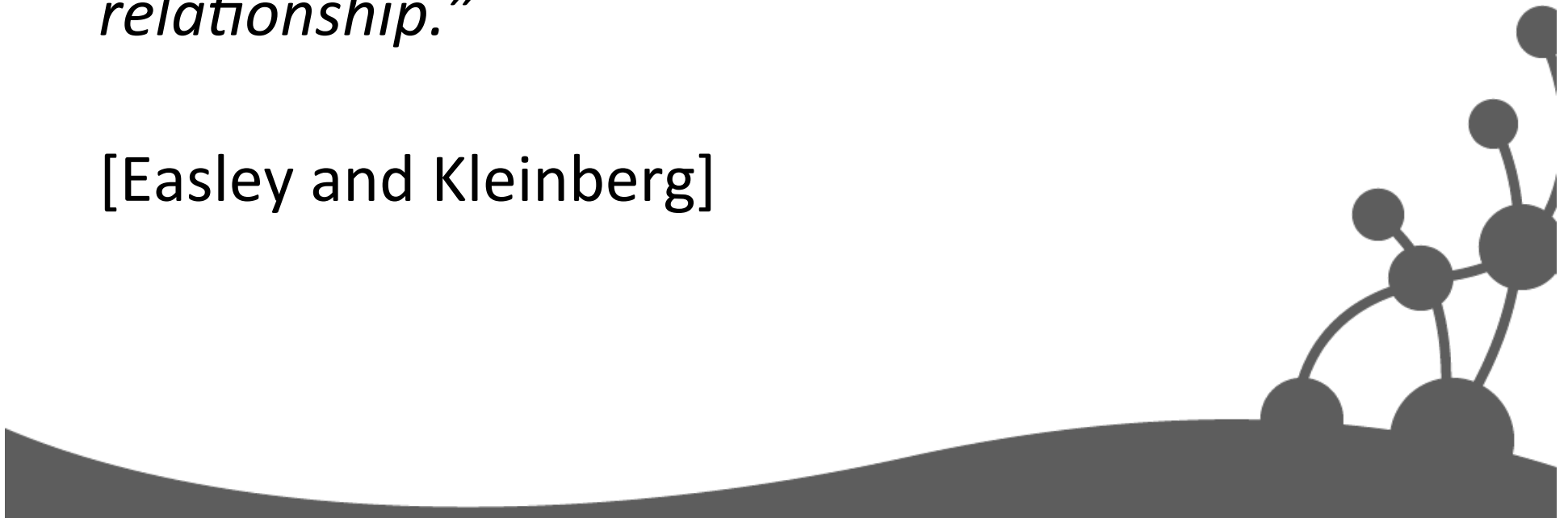
- Weak links play another **super-important** structural role in graph theory
- They bridge neighbourhoods
- Which is allows us to partition graphs



# Local Bridge Property

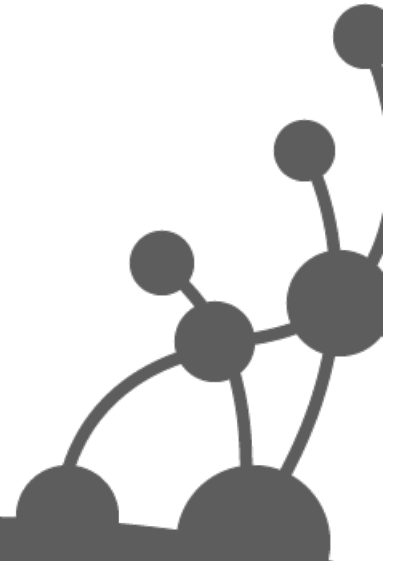
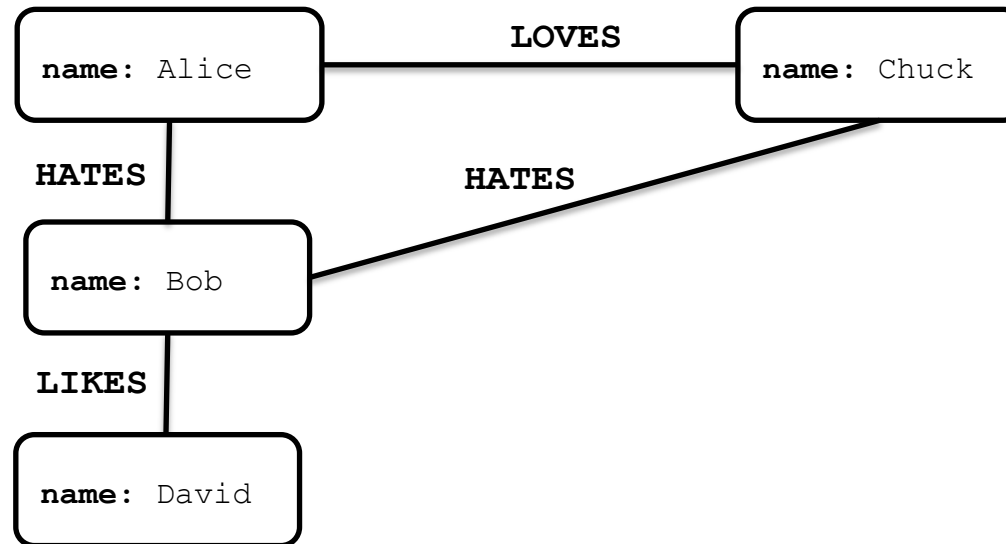
*“If a node **A** in a network satisfies the Strong Triadic Closure Property and is involved in at least two strong relationships, then any local bridge it is involved in must be a weak relationship.”*

[Easley and Kleinberg]

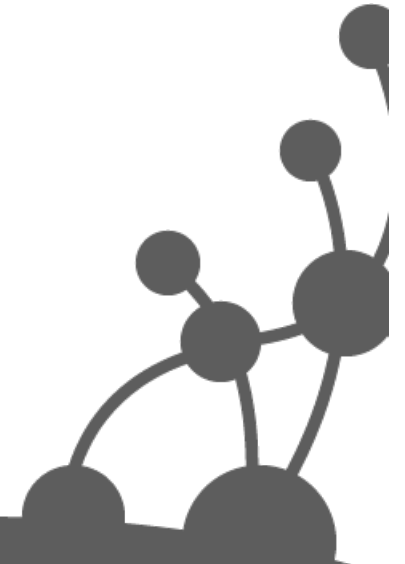
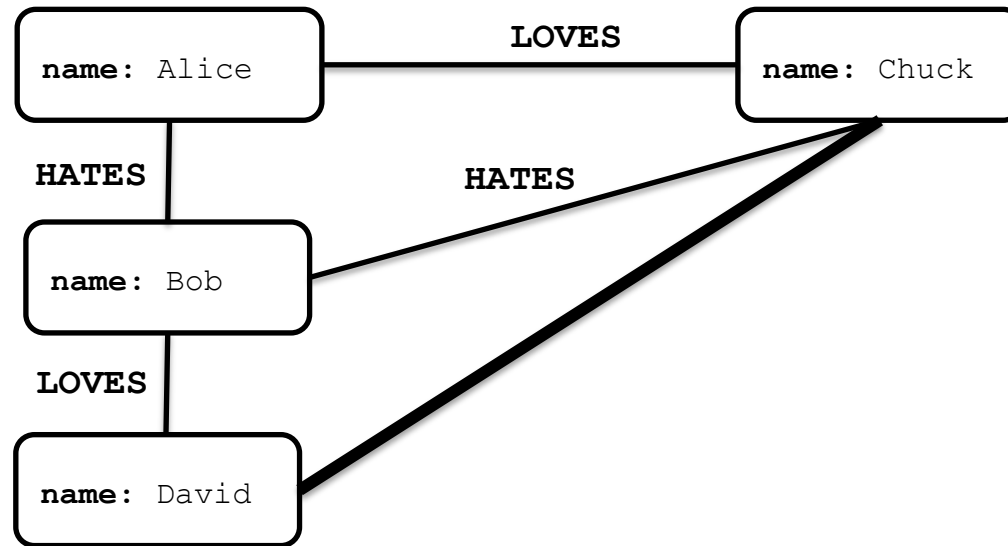




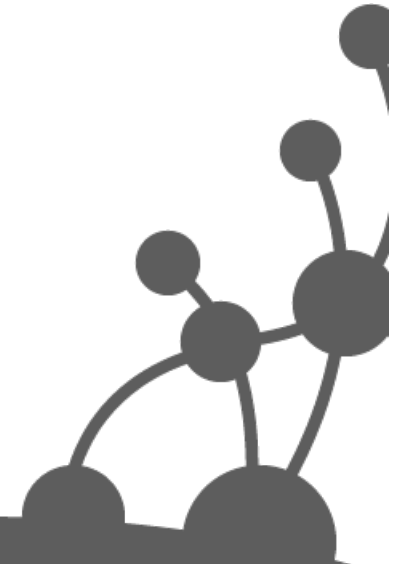
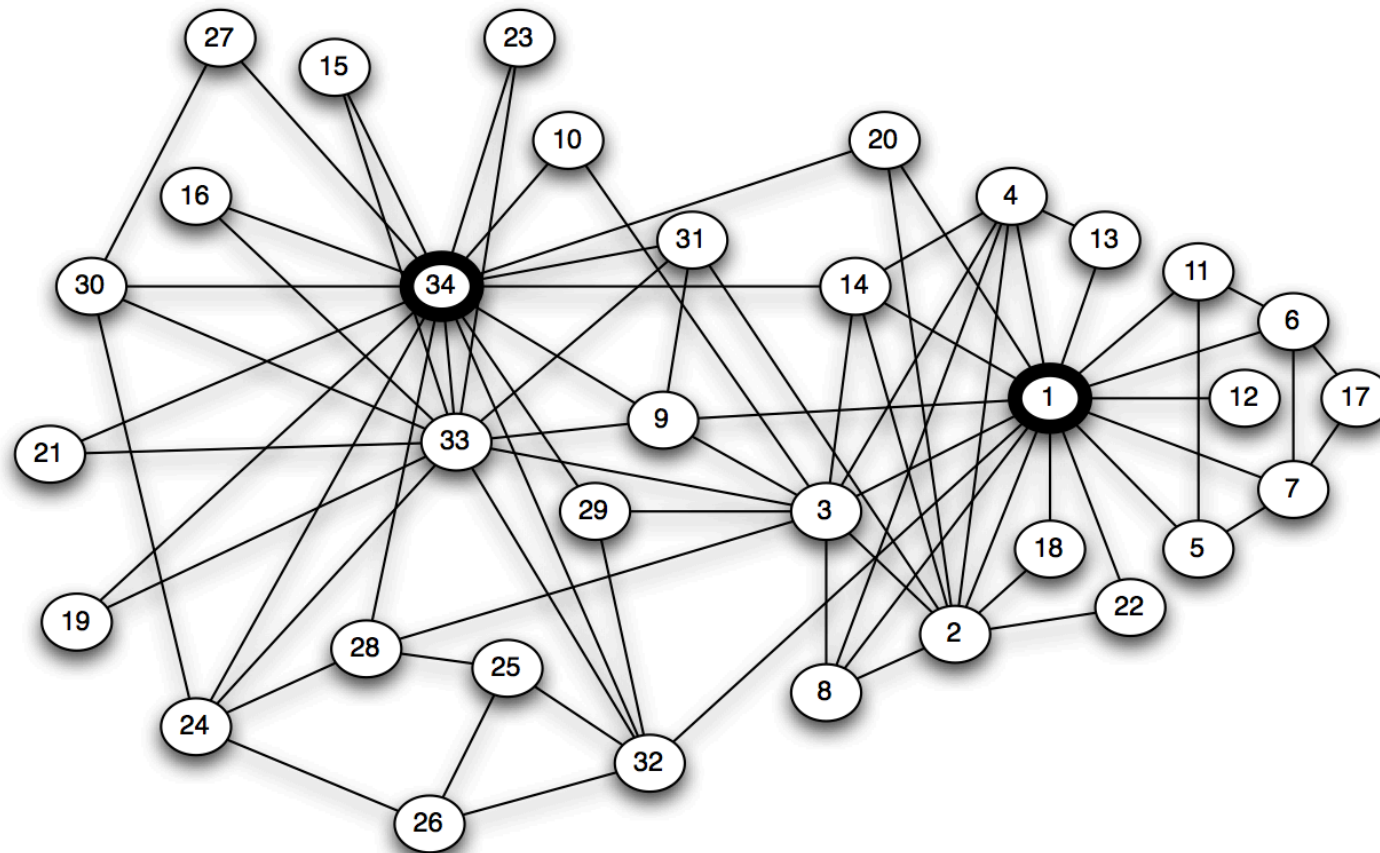
# LIKES is a local bridge



# LOVES cannot be a local bridge

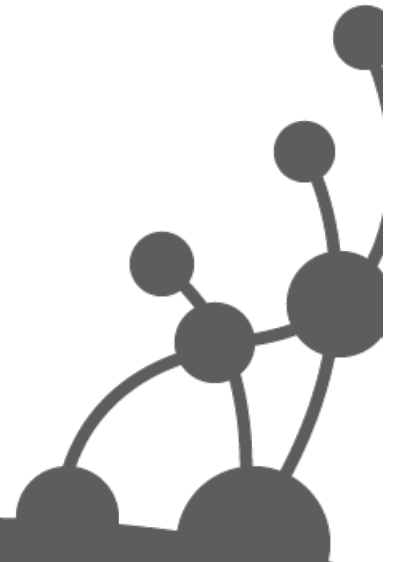


# University Karate Club



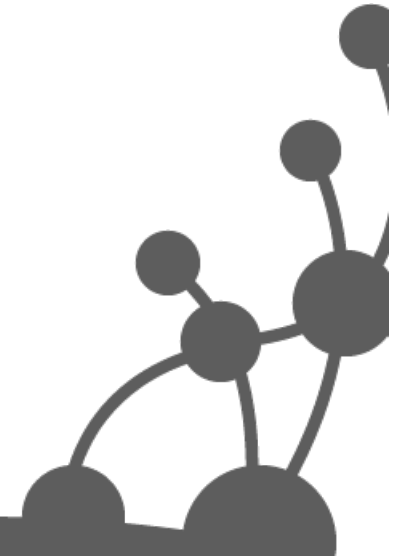
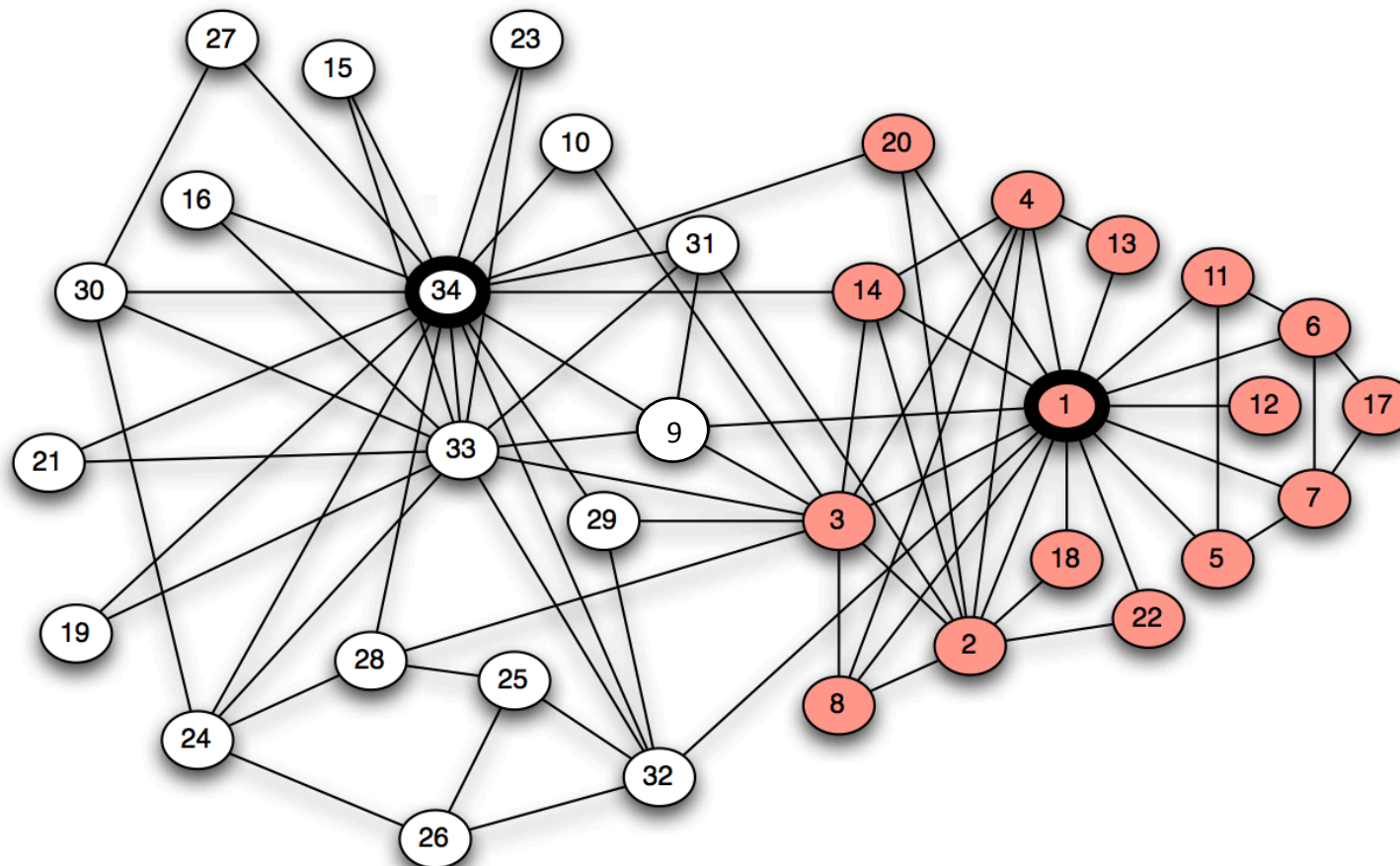
# Graph Partitioning

- (NP) Hard problem
  - Recursively remove bridges between dense regions
  - Choose your algorithm carefully – some are better than others for a given domain
- Probabilistic methods are cheaper
- Can use to (almost exactly) predict the break up of the karate club!



# University Karate Clubs

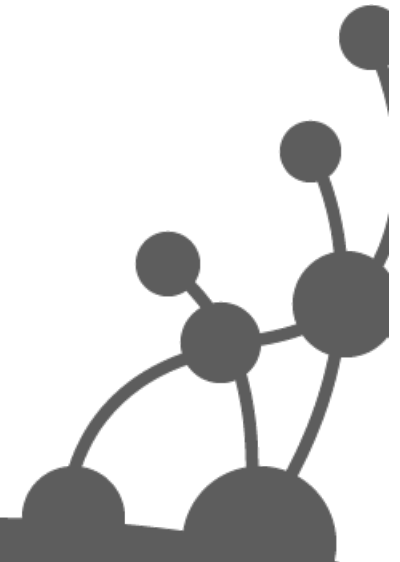
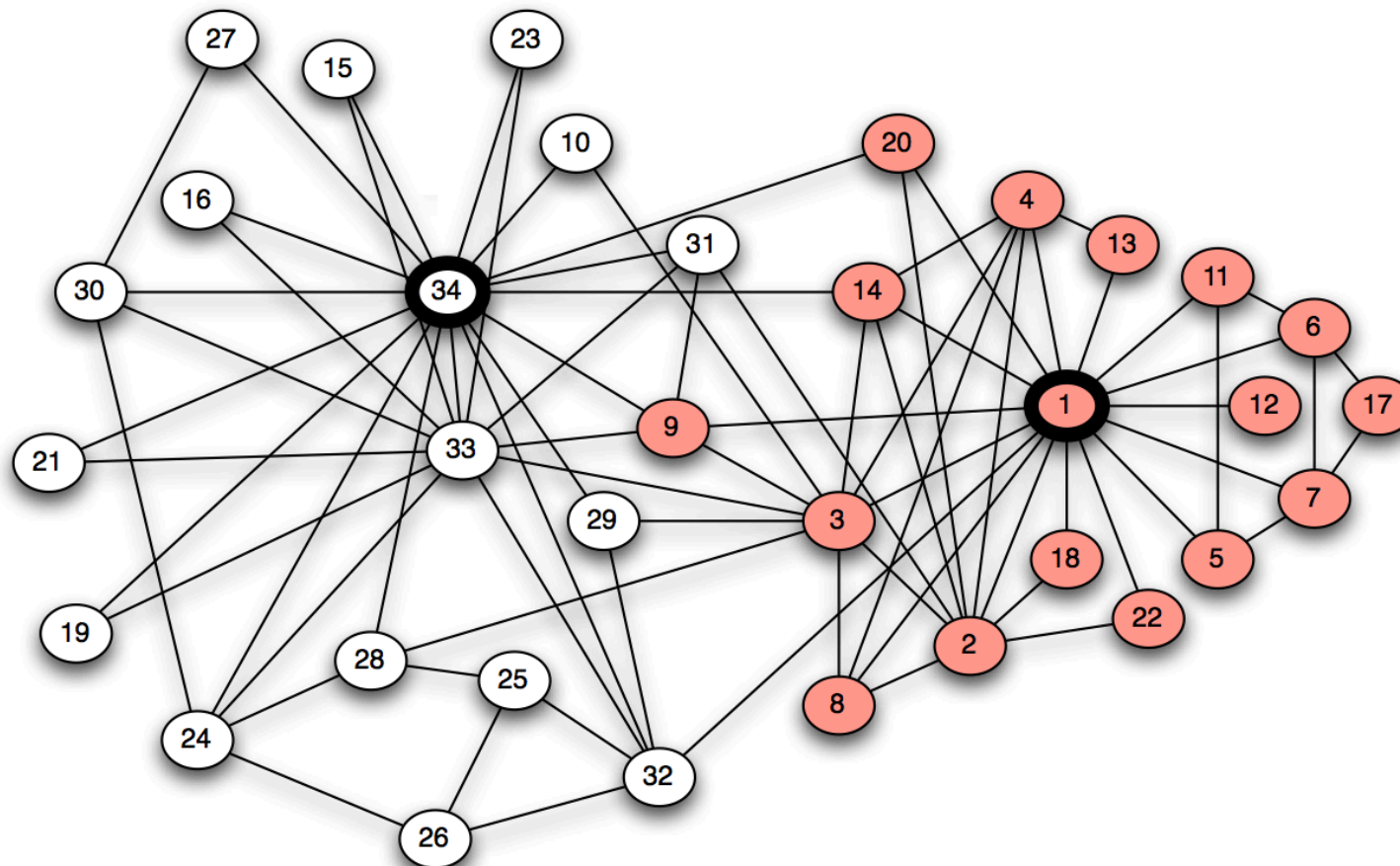
(predicted by Graph Theory)





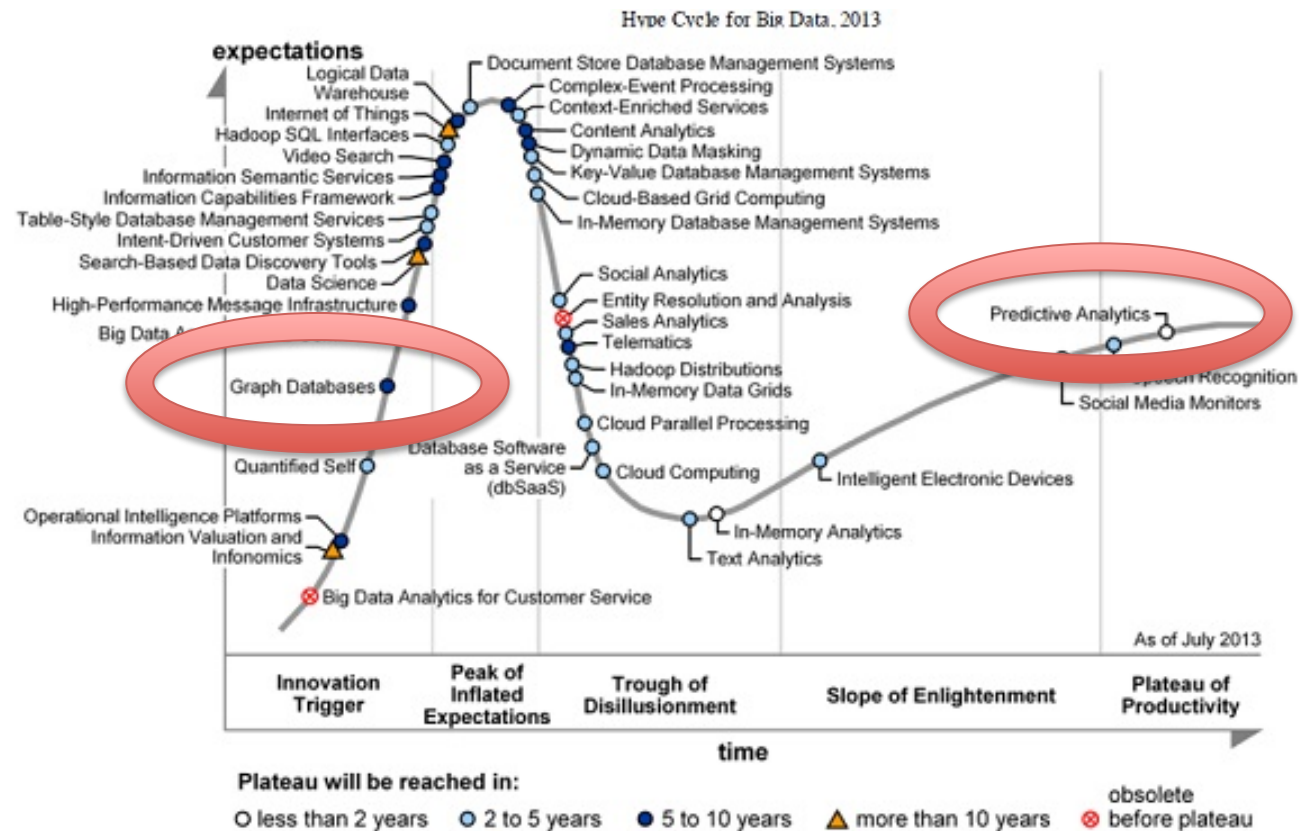
# University Karate Clubs

(what actually happened!)



# Productive Predictive Analytics in 2 Years

(says Gartner prediction)





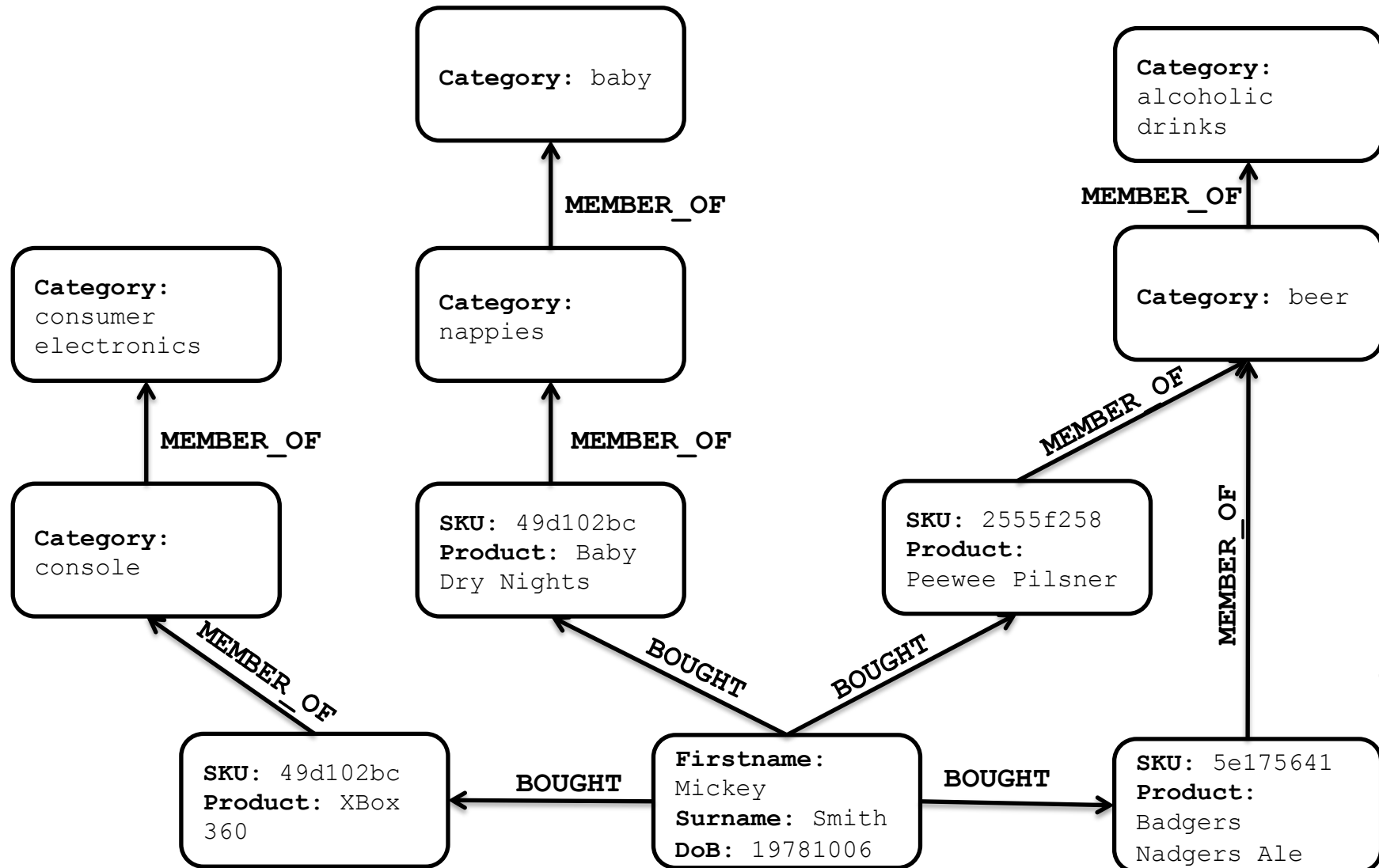
# Cypher

- Declarative graph pattern matching language
  - “SQL for graphs”
  - A humane tool pioneered by a tamed SQL DBA
- A pattern graph matching language
  - Find me stuff like...



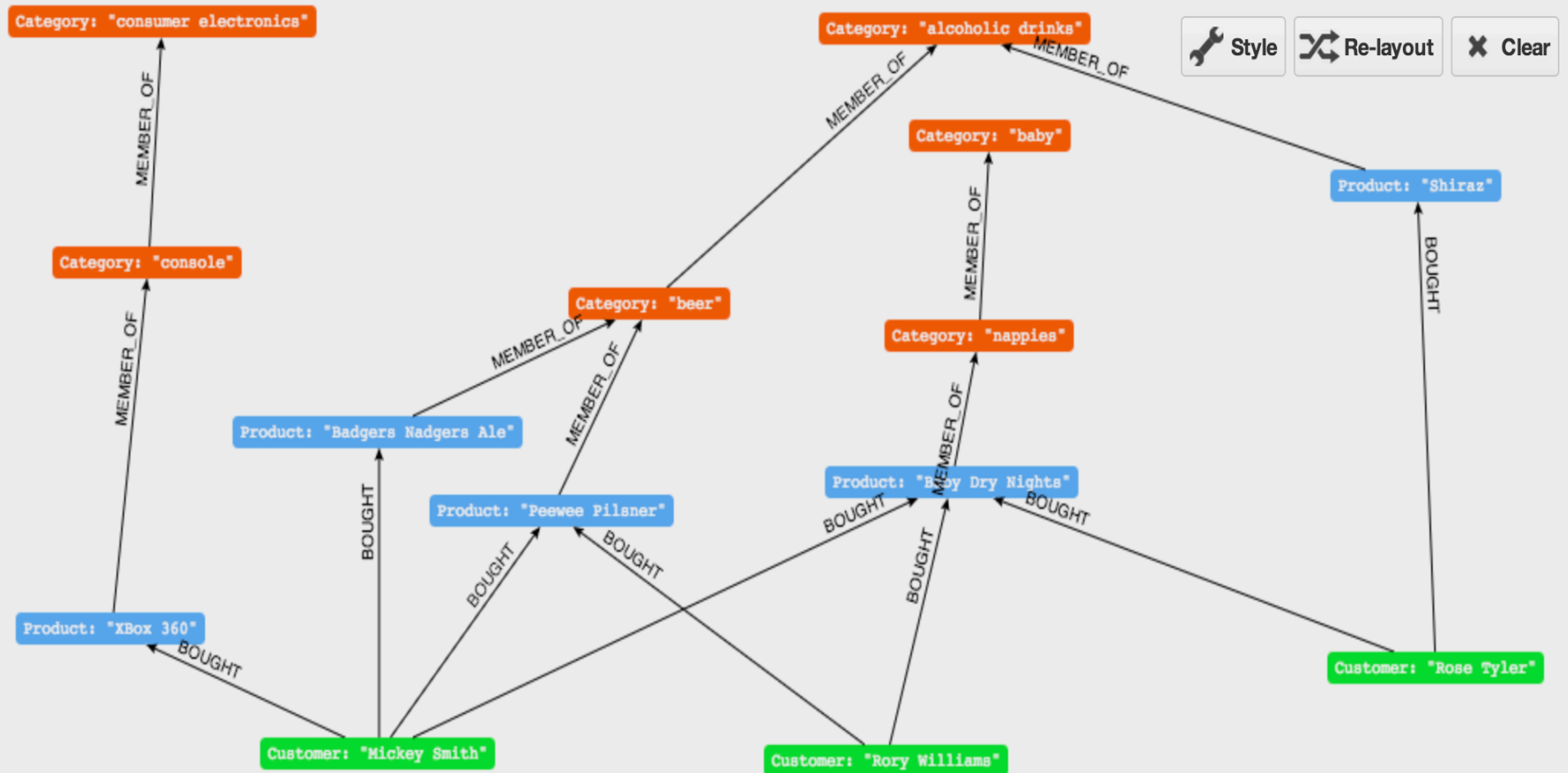


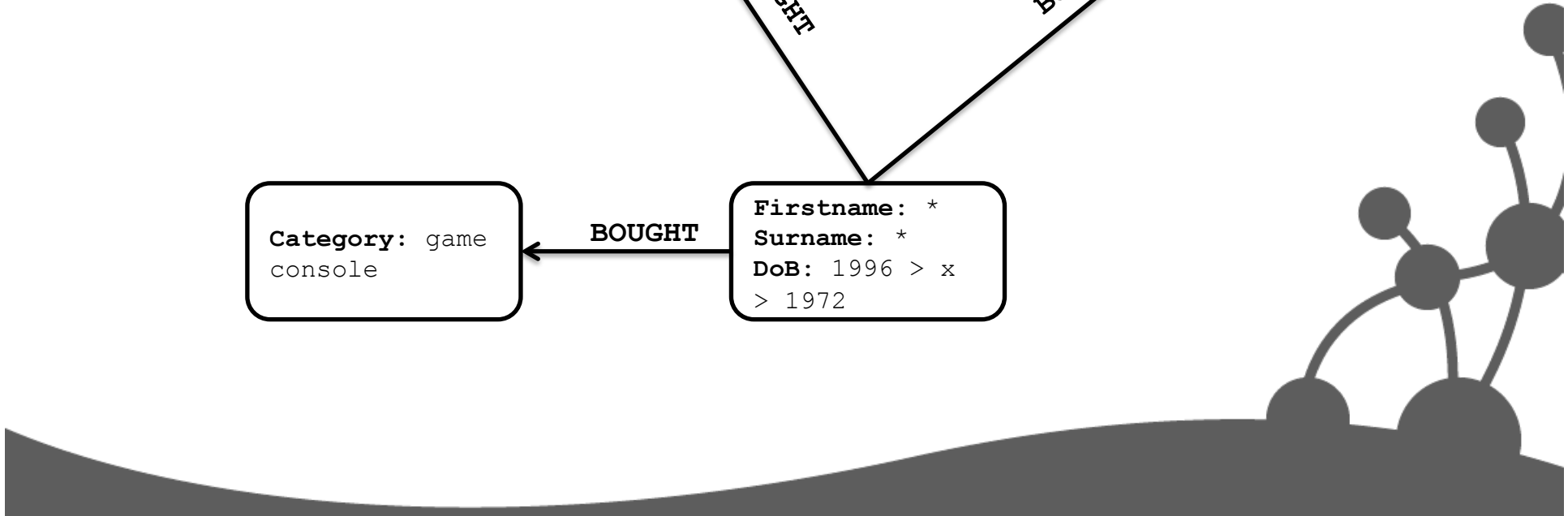
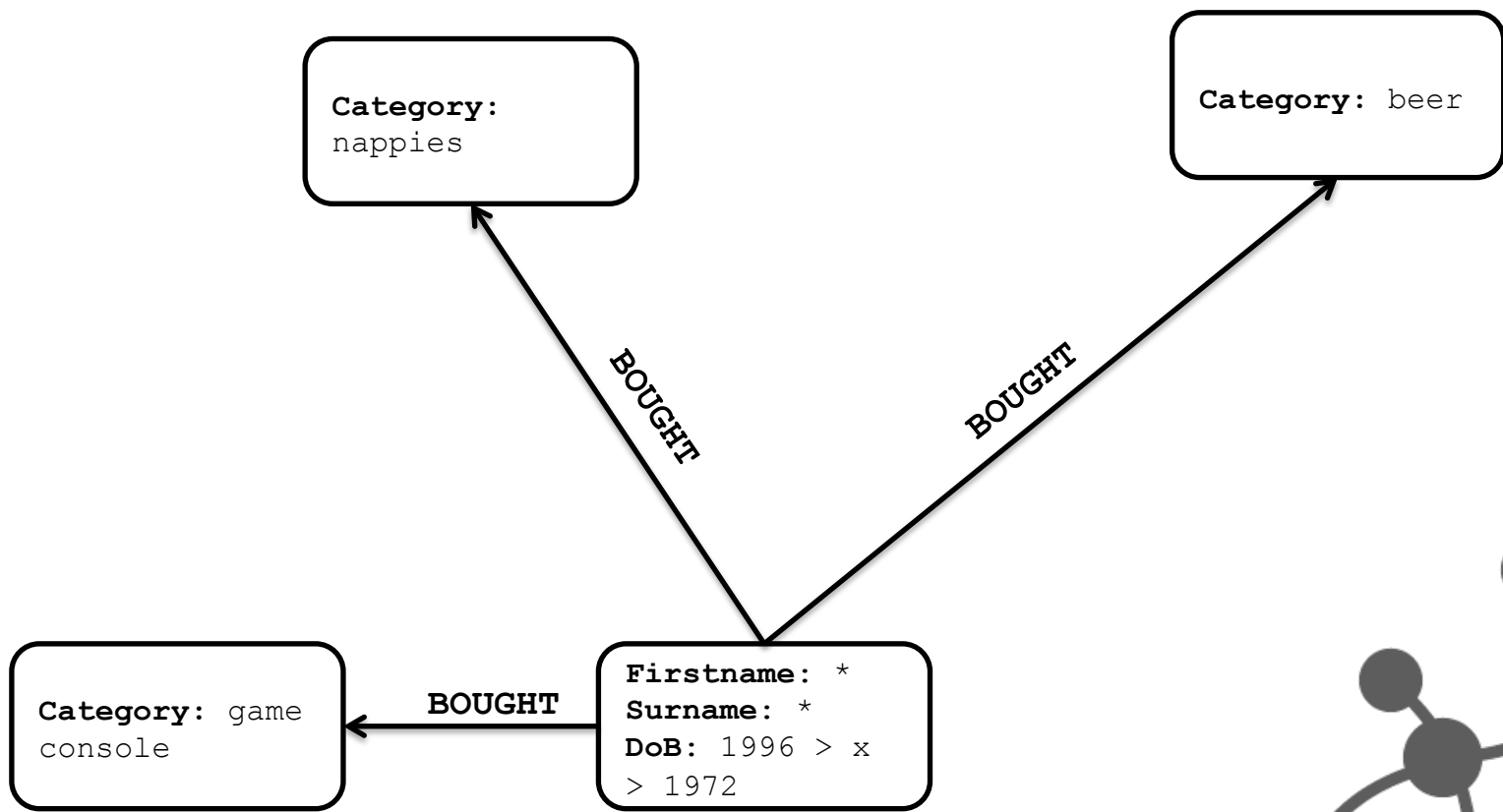


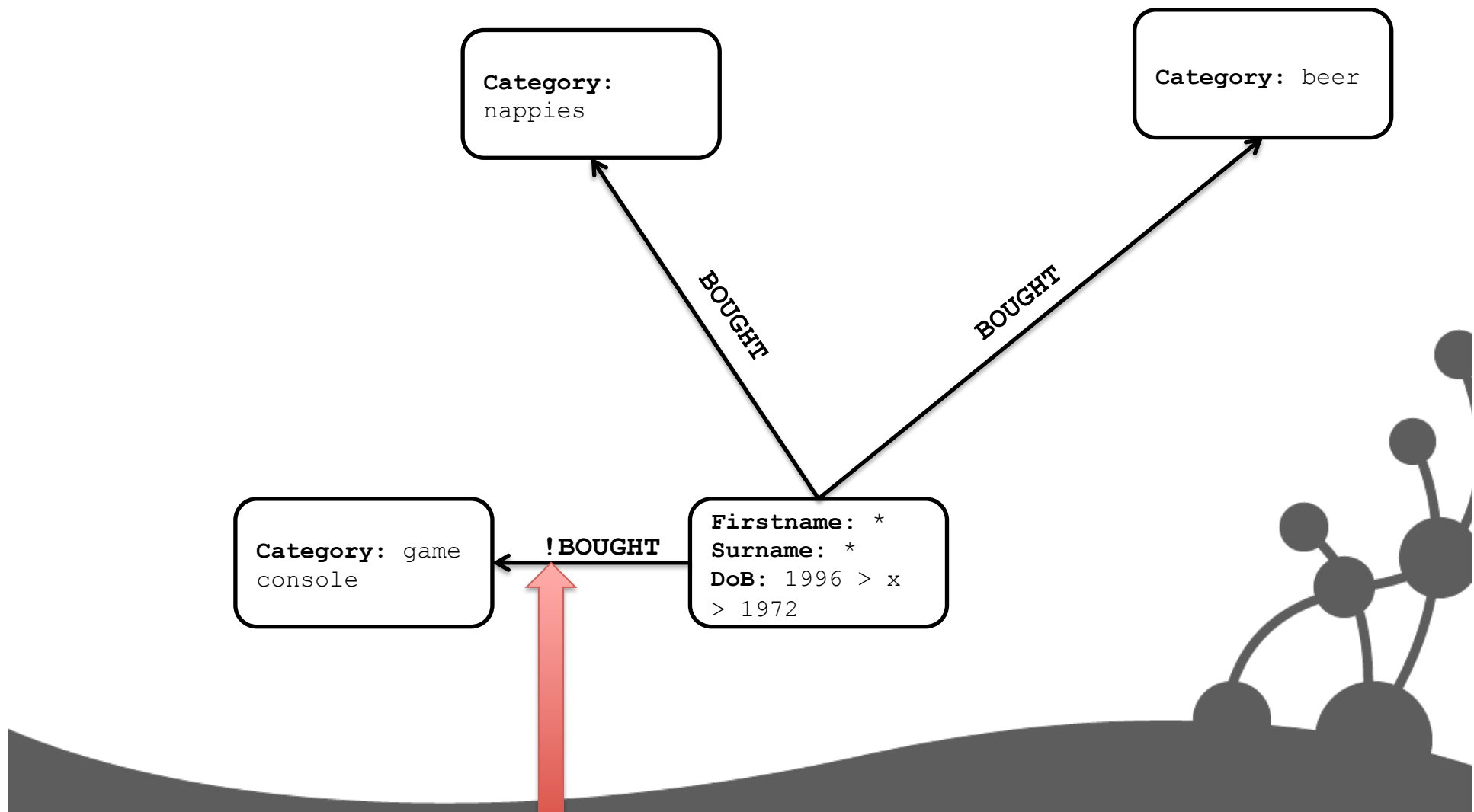


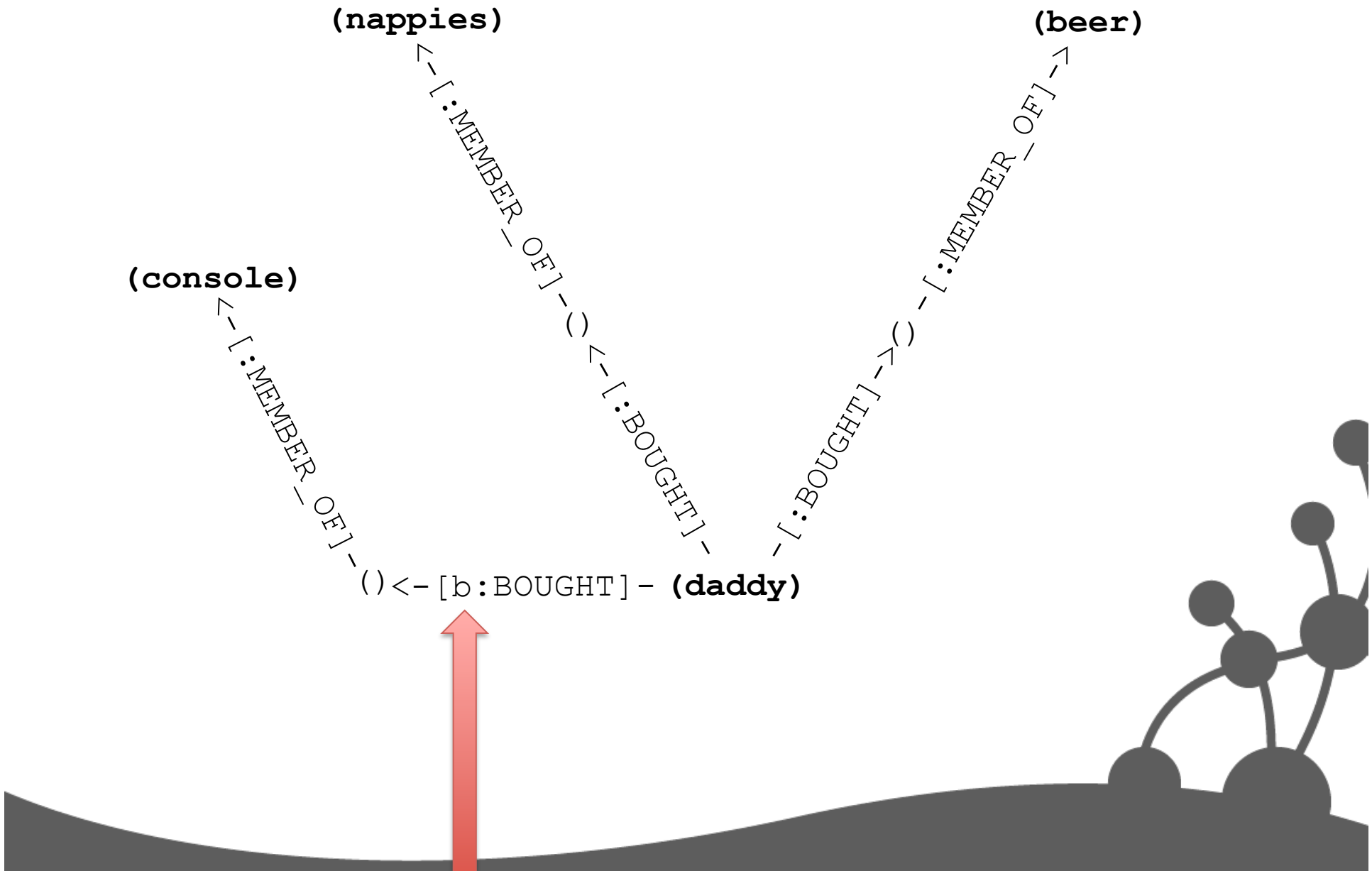

[+ Node](#)
[+ Relationship](#)


Returned 1 row. Query took 20ms









# Flatten the graph

```
(daddy) - [:BOUGHT] -> () - [:MEMBER_OF] -> (nappies)  
(daddy) - [:BOUGHT] -> () - [:MEMBER_OF] -> (beer)  
(daddy) - [b:BOUGHT] -> () - [:MEMBER_OF] -> (console)
```





# Wrap in a Cypher MATCH clause

```
MATCH (daddy) -[:BOUGHT]->()-[:MEMBER_OF]->(nappies),  
(daddy) -[:BOUGHT]->()-[:MEMBER_OF]->(beer),  
(daddy) -[b:BOUGHT]->()-[:MEMBER_OF]->(console)
```



# Cypher WHERE clause

```
MATCH (daddy)-[:BOUGHT]->()-[:MEMBER_OF]->(nappies),  
      (daddy)-[:BOUGHT]->()-[:MEMBER_OF]->(beer),  
      (daddy)-[b:BOUGHT]->()-[:MEMBER_OF]->(console)  
WHERE b is null
```



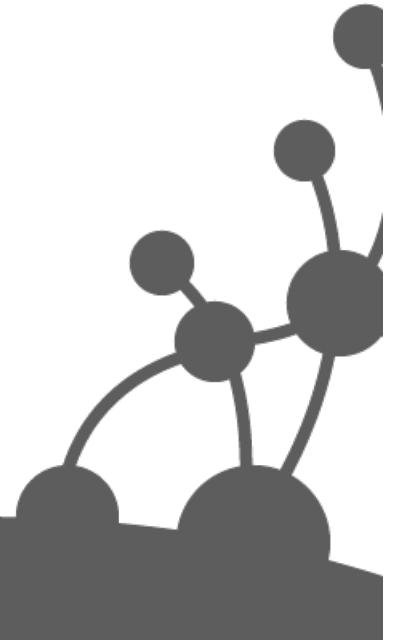
# Cypher 1.x recommendations

```
START beer=node:categories(category='beer'),  
      nappies=node:categories(category='nappies'),  
      xbox=node:products(product='xbox 360')
```

```
MATCH (daddy)-[:BOUGHT]->()-[:MEMBER_OF]->(beer),  
      (daddy)-[:BOUGHT]->()-[:MEMBER_OF]->(nappies),  
      (daddy)-[b?:BOUGHT]->(xbox)
```

```
WHERE b is null
```

```
RETURN distinct daddy
```

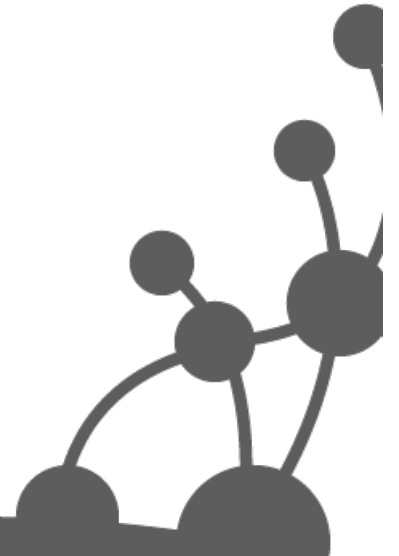


# Cypher 2.0 recommendations

```
MATCH (d:Person) -[:BOUGHT]->() -[:MEMBER_OF]->(n:Category),  
      (d:Person) -[:BOUGHT]->() -[:MEMBER_OF]->(b:Category),  
      (c:Category)
```

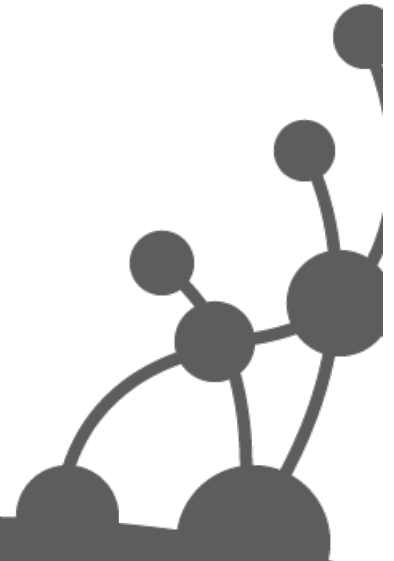
```
WHERE n.category = "nappies" AND  
      b.category = "beer" AND  
      c.category = "console" AND  
      NOT((d) -[:BOUGHT]->() -[:MEMBER_OF]->(c))
```

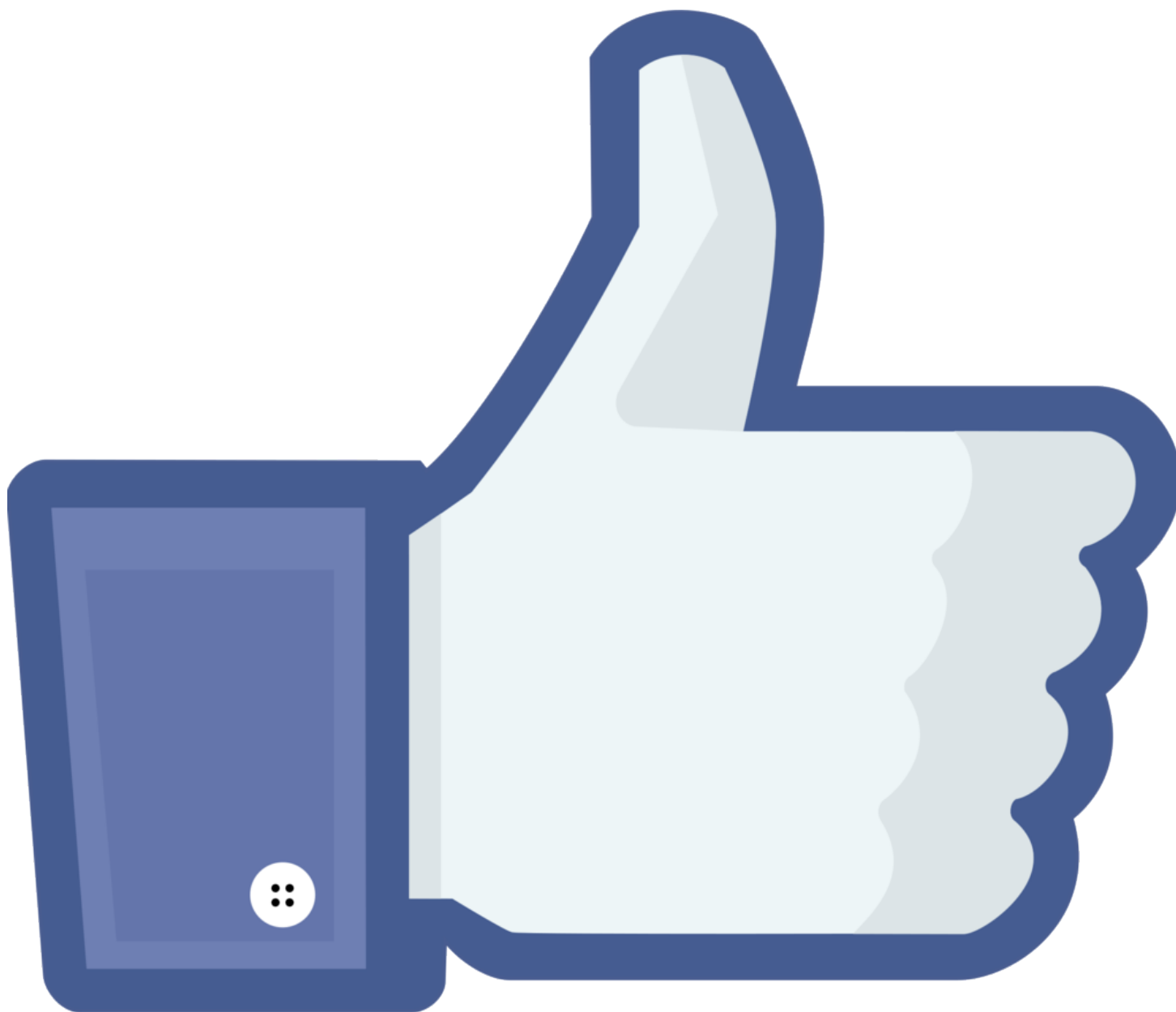
```
RETURN DISTINCT d AS daddy
```



# Results

```
==> +-----+
==> | daddy                                     |
==> +-----+
==> | Node[15]{name:"Rory Williams",dob:19880121} |
==> +-----+
==> 1 row
==> 0 ms
==>
neo4j-sh (0)$
```







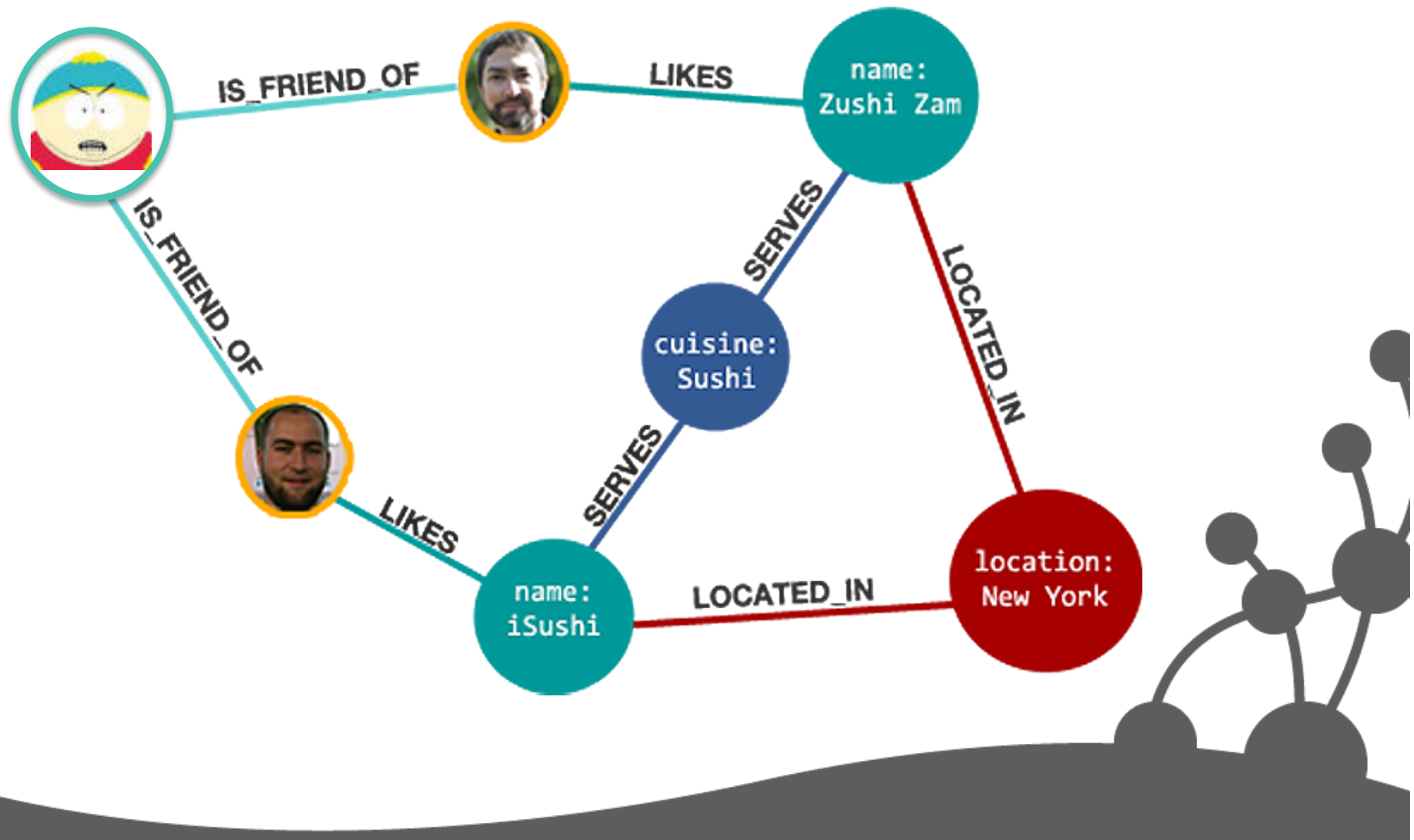
# Facebook Graph Search

Which sushi restaurants in  
NYC do my friends like?



See <http://maxdemarzi.com/>

# Graph Structure

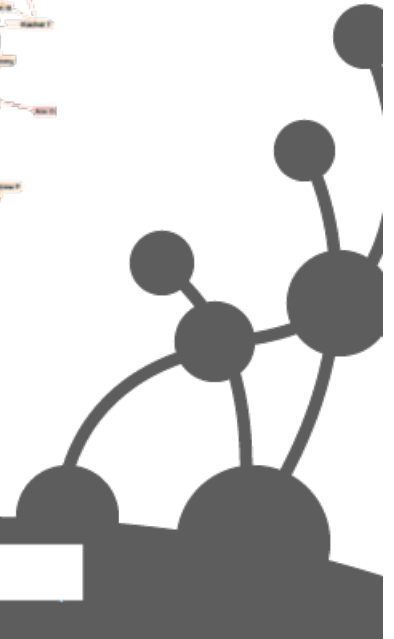
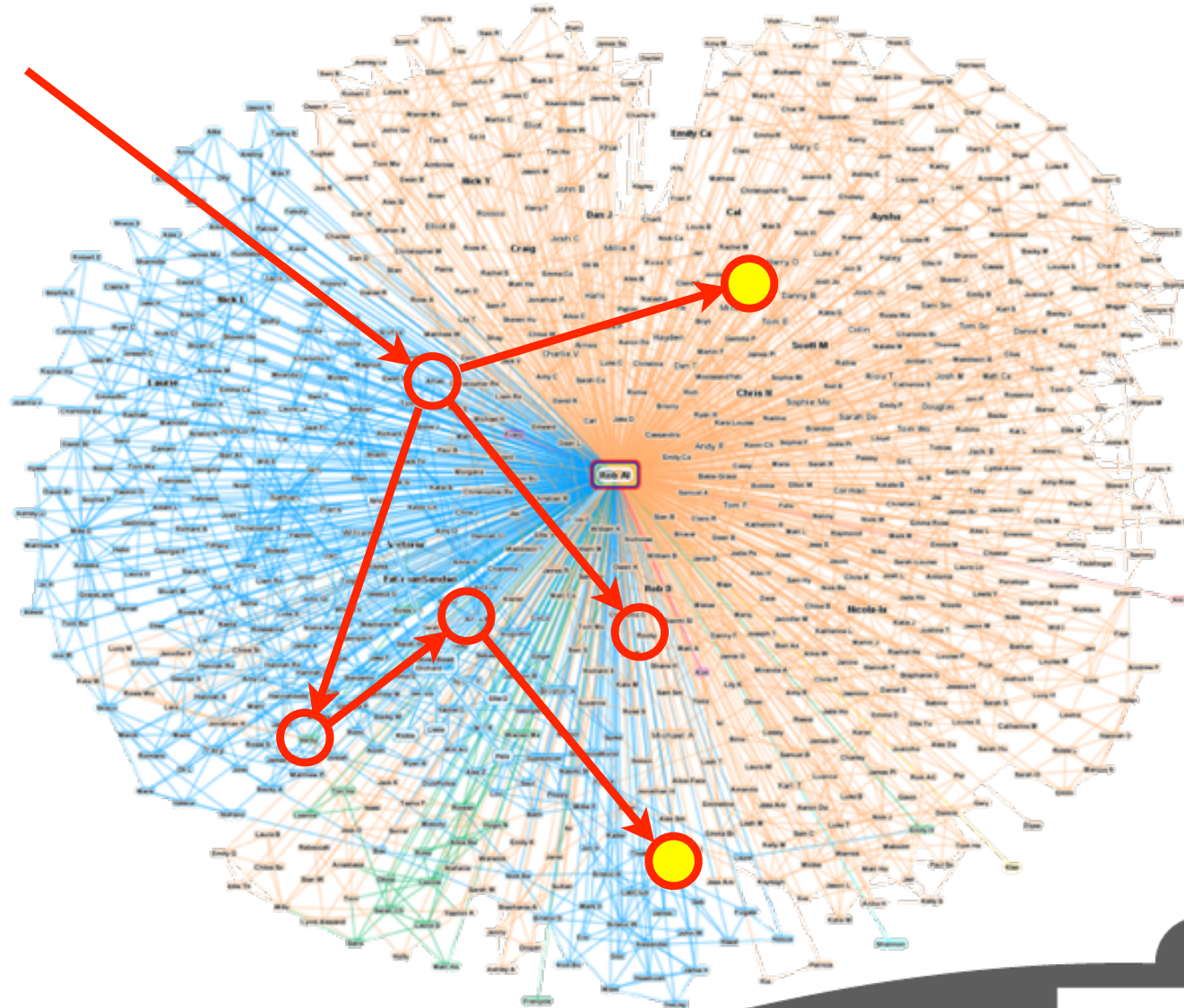


# Cypher query

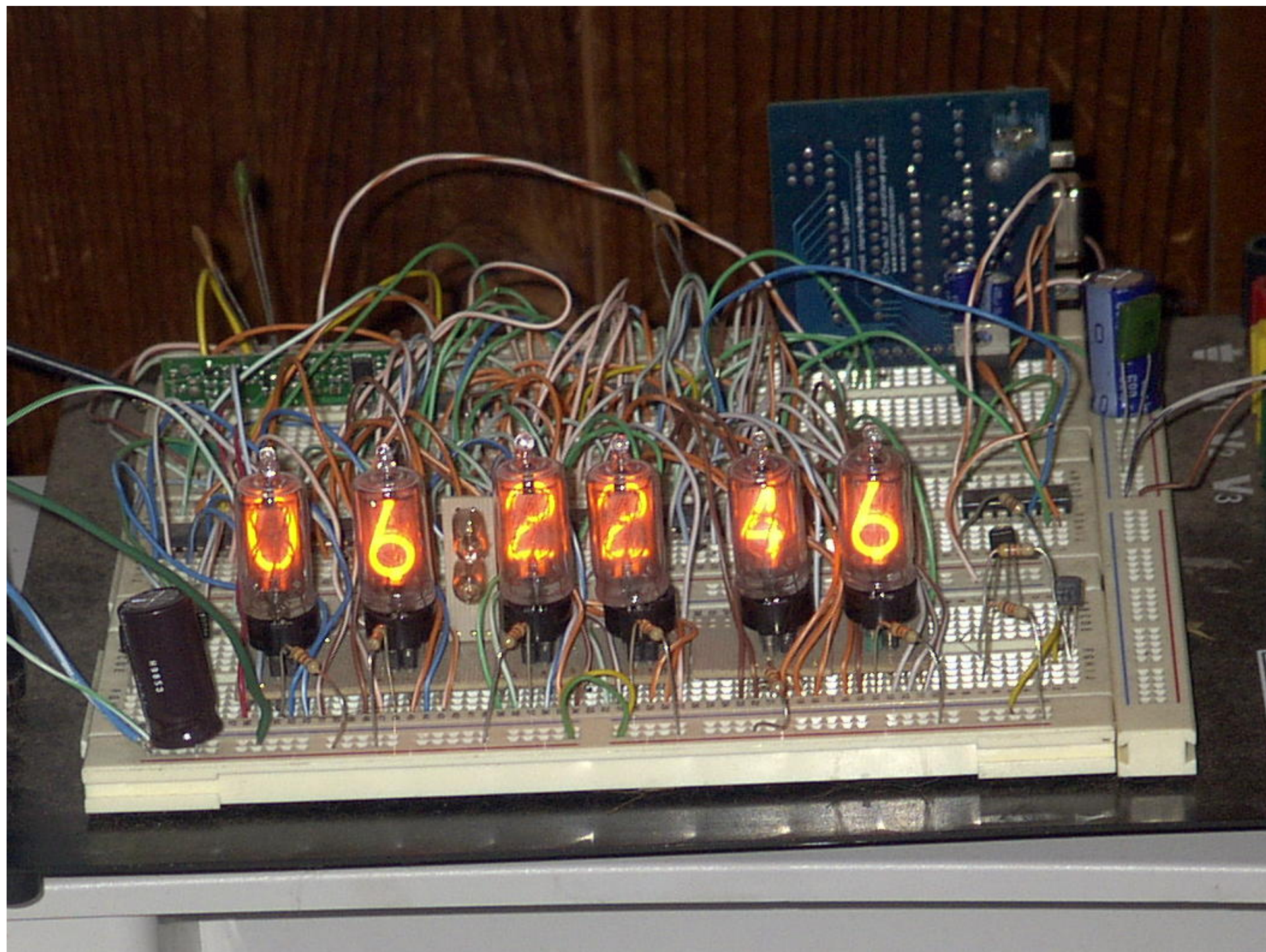
```
START me=node:person(name = 'Jim'),  
        location=node:location(location='New York'),  
        cuisine=node:cuisine(cuisine='Sushi')  
  
MATCH (me)-[:IS_FRIEND_OF]->(friend)-[:LIKES]->(restaurant)  
        -[:LOCATED_IN]->(location), (restaurant)-[:SERVES]->(cuisine)  
  
RETURN restaurant
```



# Search structure

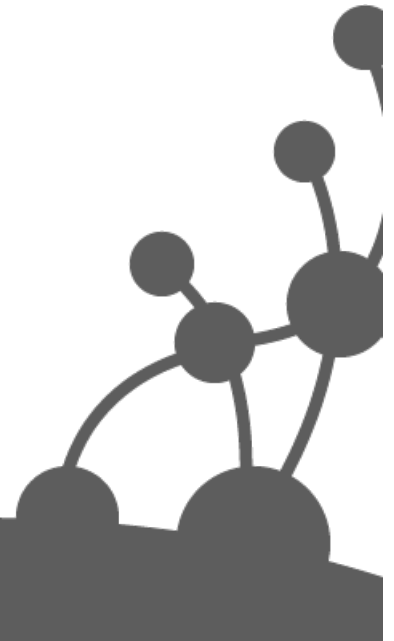






# What's Neo4j good for?

- Data centre management
- Supply chain/provenance
- Recommendations
- Business intelligence
- Social computing
- MDM
- Web of things
- Time series/event data
- Product/engineering catalogue
- Web analytics, user journeys
- Scientific computing
- Spatial
- Geo/Seismic/Meteorological
- Bio/Pharma
- And many, many more...





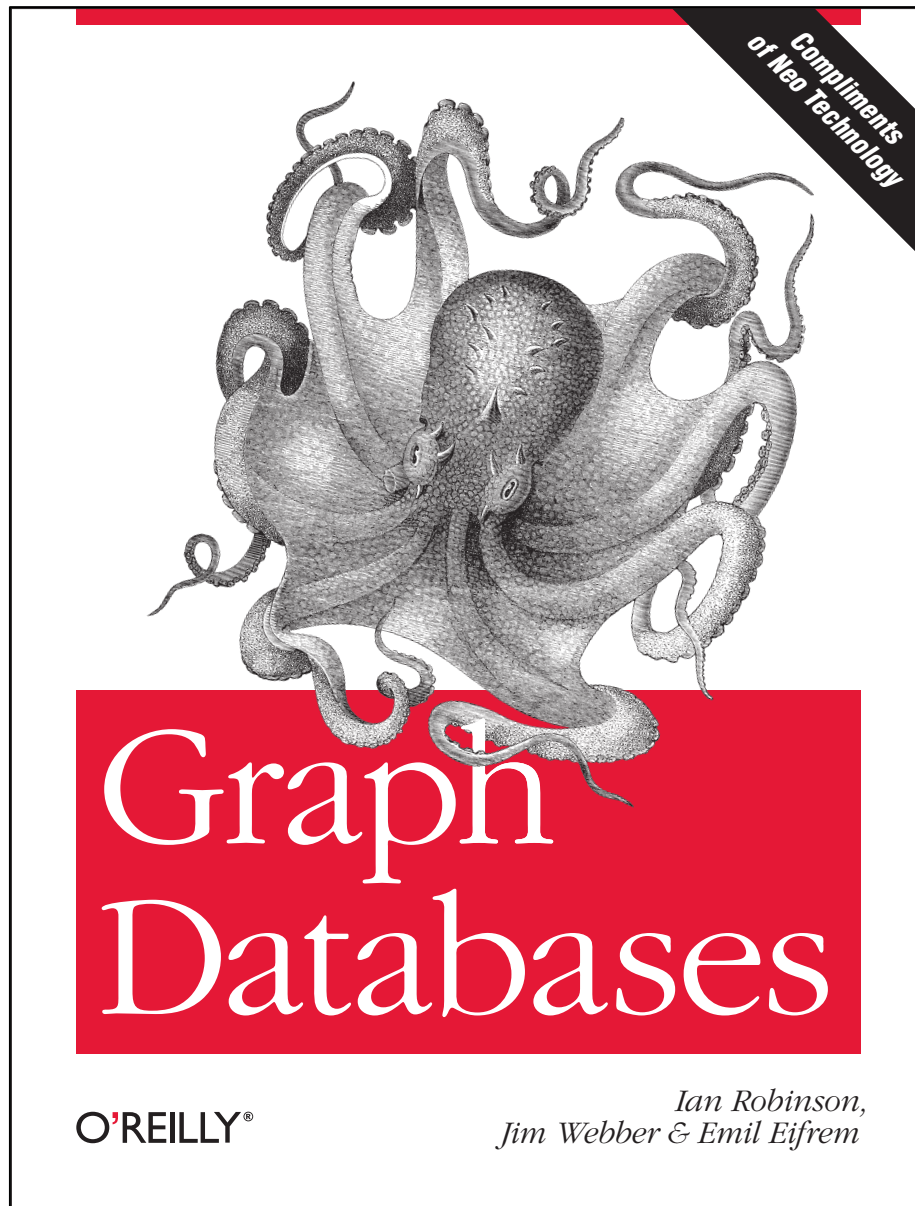
Can Neo4j compute

42?



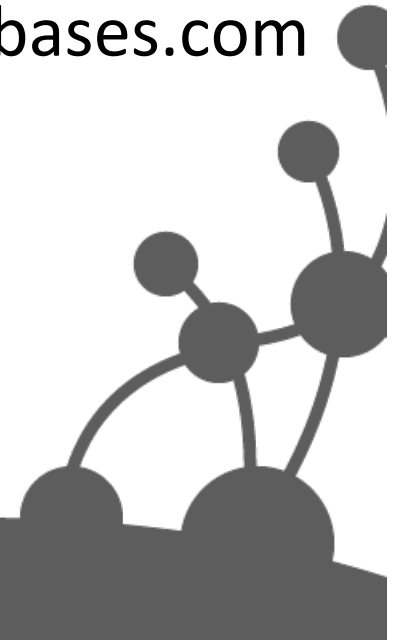
```
MATCH (you) - [:HAPPINESS] -> (answer)  
RETURN answer
```





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<http://graphdatabases.com>





# Thanks for listening

@jimwebber

<http://neo4j.org>

