

What <<b u s i n e s s >> can learn from Dating

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Neo Technology Overview



Company

- Neo Technology, Creator of Neo4j
- 90 employees with HQ in Silicon Valley, London, Munich, Paris and Malmö
- \$45M in funding from Fidelity, Sunstone, Conor, Creandum, Dawn Capital

Product

- Neo4j - World's leading graph database
- 1M+ downloads, adding 50k+ per month
- 170+ enterprise subscription customers including over 50 of the Global 2000



How Customers Use Neo4j



Social

Recom-
mendations

Search &
Discovery

Network & Data
Center

Master Data
Management

Identity &
Access

GEO

eHarmony

careerbuilder

Walmart



TOMTOM

me9tic

classmates

careerbuilder



Pitney Bowes



ebay now

Hinge



InfoJobs



ADP ADVENT



classmates

DOWN



LIQUID COMMON



Zenoss



KiwiRail

maaii



codex



TECHCRUNCH



DingLicom

ZEPHYR HEALTH INC



NOMURA

TRIBAL technologies



Perigee

mallow street



Lufthansa



ZEPHYR HEALTH INC

Didacti

DOWN



compete



3 Juice PLUS+

aikux.com

CHRONOTRACK

LIFECHURCH.TV

bwin.party



springcm



EQUILAR

YELAGO

doximity

LiveStation



Neo4j Leads the Graph Database Revolution



“Graph analysis is possibly the **single most effective competitive differentiator** for organizations pursuing data-driven operations and decisions after the design of data capture.”



“Forrester estimates that **over 25% of enterprises** will be using graph databases by 2017”



“Neo4j is the current market **leader in graph databases.**”

IT Market Clock for Database Management Systems, 2014

<https://www.gartner.com/doc/2852717/it-market-clock-database-management>

TechRadar™: Enterprise DBMS, Q1 2014

<http://www.forrester.com/TechRadar+Enterprise+DBMS+Q1+2014/fulltext/-/E-RES106801>

Graph Databases – and Their Potential to Transform How We Capture Interdependencies (Enterprise Management Associates)

<http://blogs.enterprisemanagement.com/dennisdrogseth/2013/11/06/graph-databasesand-potential-transform-capture-interdependencies/>

High Business Value in Data Relationships



Data is increasing in volume...

- New digital processes
- More online transactions
- New social networks
- More devices

... and is getting more connected

Customers, products, processes, devices interact and relate to each other

Using Data Relationships unlocks value

- Real-time recommendations
- Fraud detection
- Master data management
- Network and IT operations
- Identity and access management
- Graph-based search

We need to p
and use the s

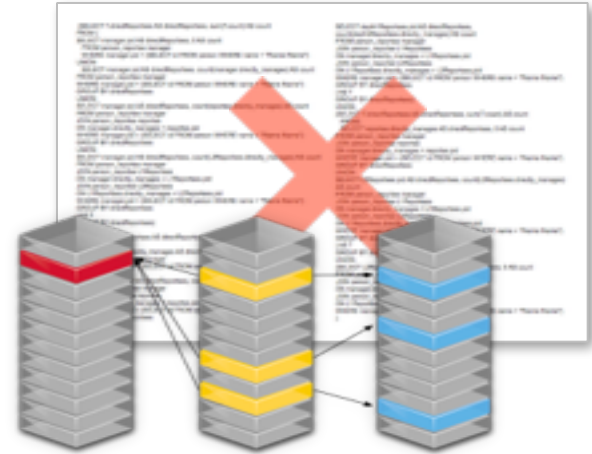


Relational DBs Can't Handle Relationships Well



- *Cannot model or store data and relationships* without complexity
- *Performance degrades* with number and levels of relationships, and database size
- *Query complexity grows* with need for JOINS
- *Adding new types of data and relationships* requires schema redesign, increasing time to market

... making traditional databases **inappropriate** when data relationships are valuable in **real-time**



Slow development
Poor performance
Low scalability
Hard to maintain

Aggregate Stores *Don't* Handle Relationships



- *No data structures* to model or store relationships
- *No query constructs* to support data relationships
- *Relating data requires "JOIN logic"* in the application
- *No ACID support* for transactions

... making NoSQL databases **inappropriate** when data relationships are valuable in **real-time**

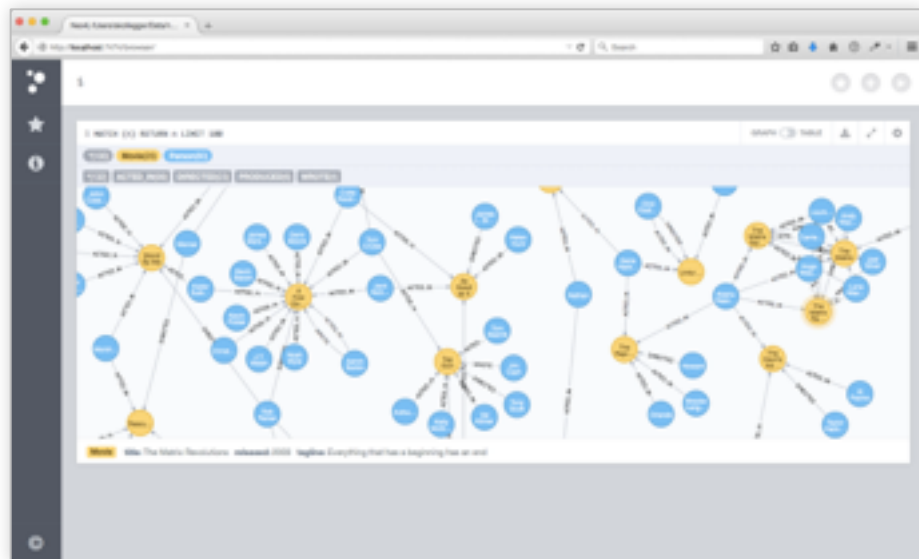


Neo4j – Re-Imagine Your Data as a Graph



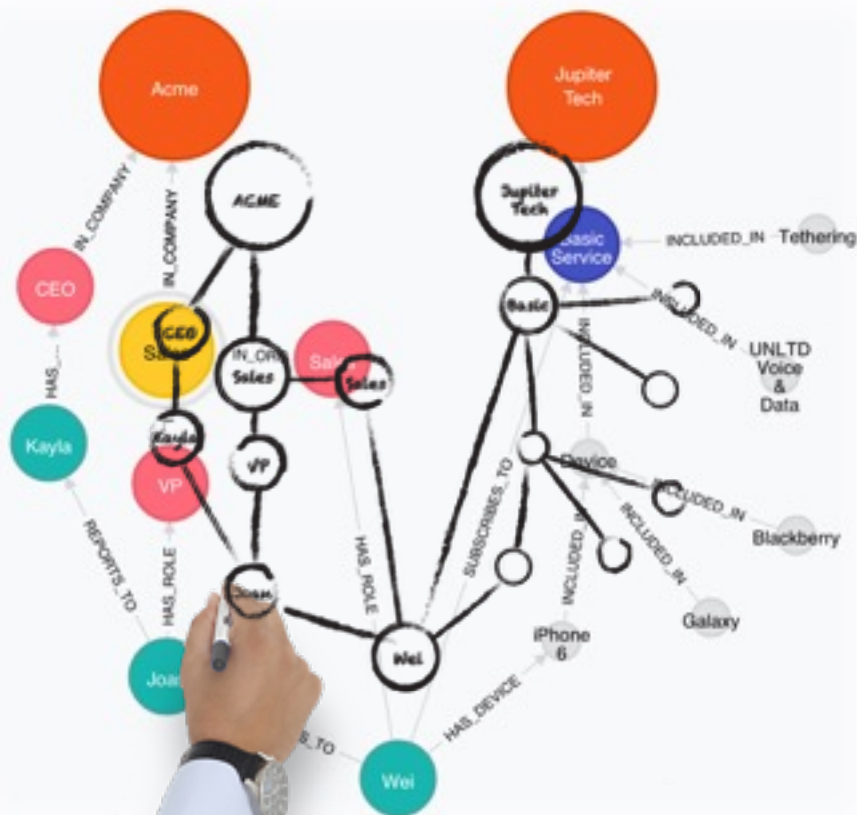
Neo4j is an *enterprise-grade graph database* that enables you to:

- **Model and store** your data as a **graph**
- **Query data relationships** with ease and in real-time
- **Seamlessly evolve applications** to support new requirements by adding new kinds of data and relationships



Agile development
High performance
Vertical and horizontal scale
Seamless evolution

The Whiteboard Model Is the Physical Model



Key Neo4j Product Features



Native Graph Storage

Ensures data consistency and performance

Native Graph Processing

Millions of hops per second, in real time

“Whiteboard Friendly” Data Modeling

Model data as it naturally occurs

High Data Integrity

Fully ACID transactions

Powerful, Expressive Query Language

Requires 10x to 100x less code than SQL

Scalability and High Availability

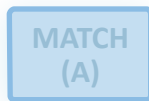
Vertical and horizontal scaling optimized for graphs

Built-in ETL

Seamless import from other databases

Integration

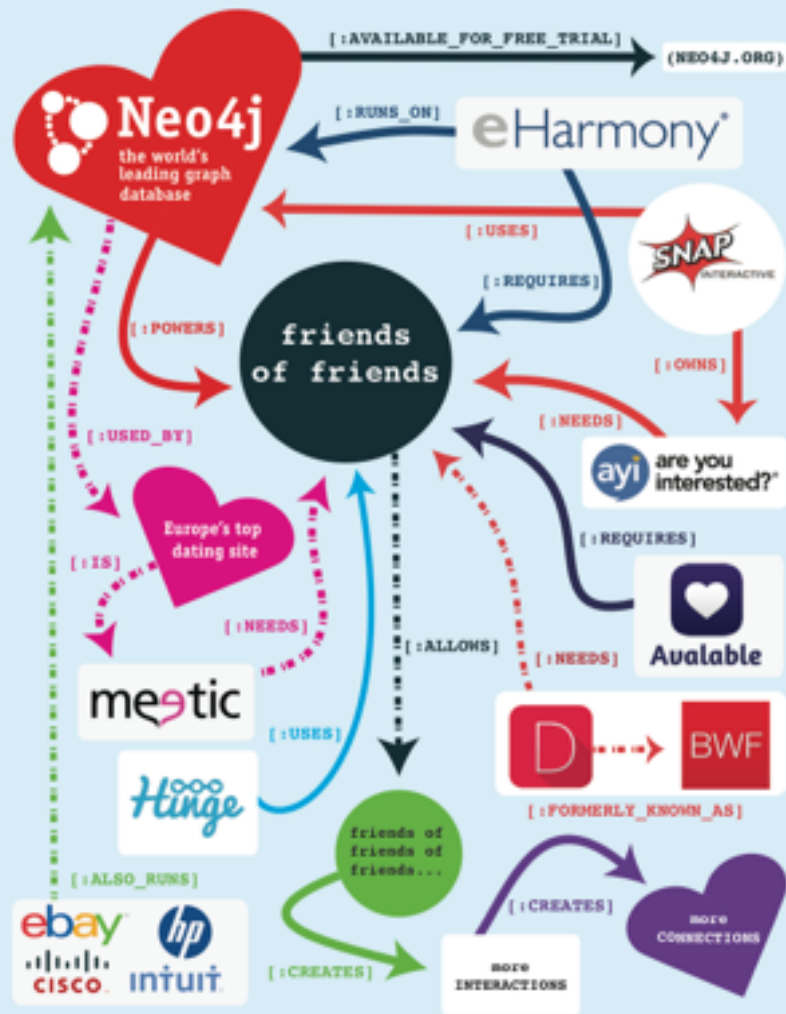
Drivers and APIs for popular languages



So...

What CAN Business learn from dating?





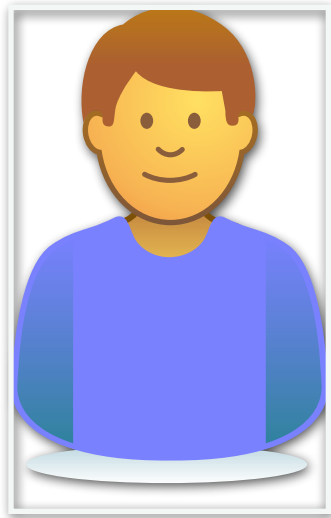
The 5 graphs of love



- The friends of friends graph
- The passion graph
- The location graph
- The safety graph
- The poser graph



Meet Jeremy



Jeremy



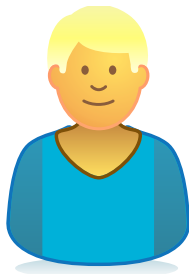
Jeremy has some friends



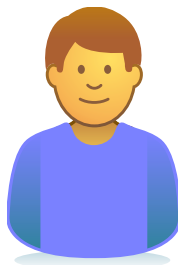
- Kerstin: his sister
- Peter: his friend
- Andreas: his colleague



Andreas



Peter



Jeremy



Kerstin

His friends introduced some more friends



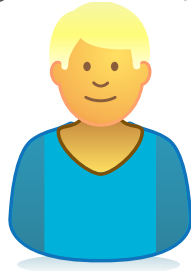
- Michael: master hacker, divorce, 2 children
- Johan: technology sage, likes fast cards
- Madelene: polyglot journalist, loves dogs
- Allison: marketing maven, likes long walks on the beach



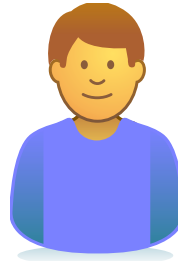
Michael



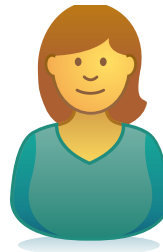
Andreas



Peter



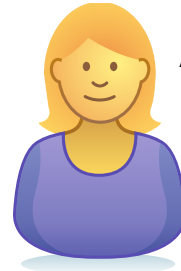
Jeremy



Madelene



Johan

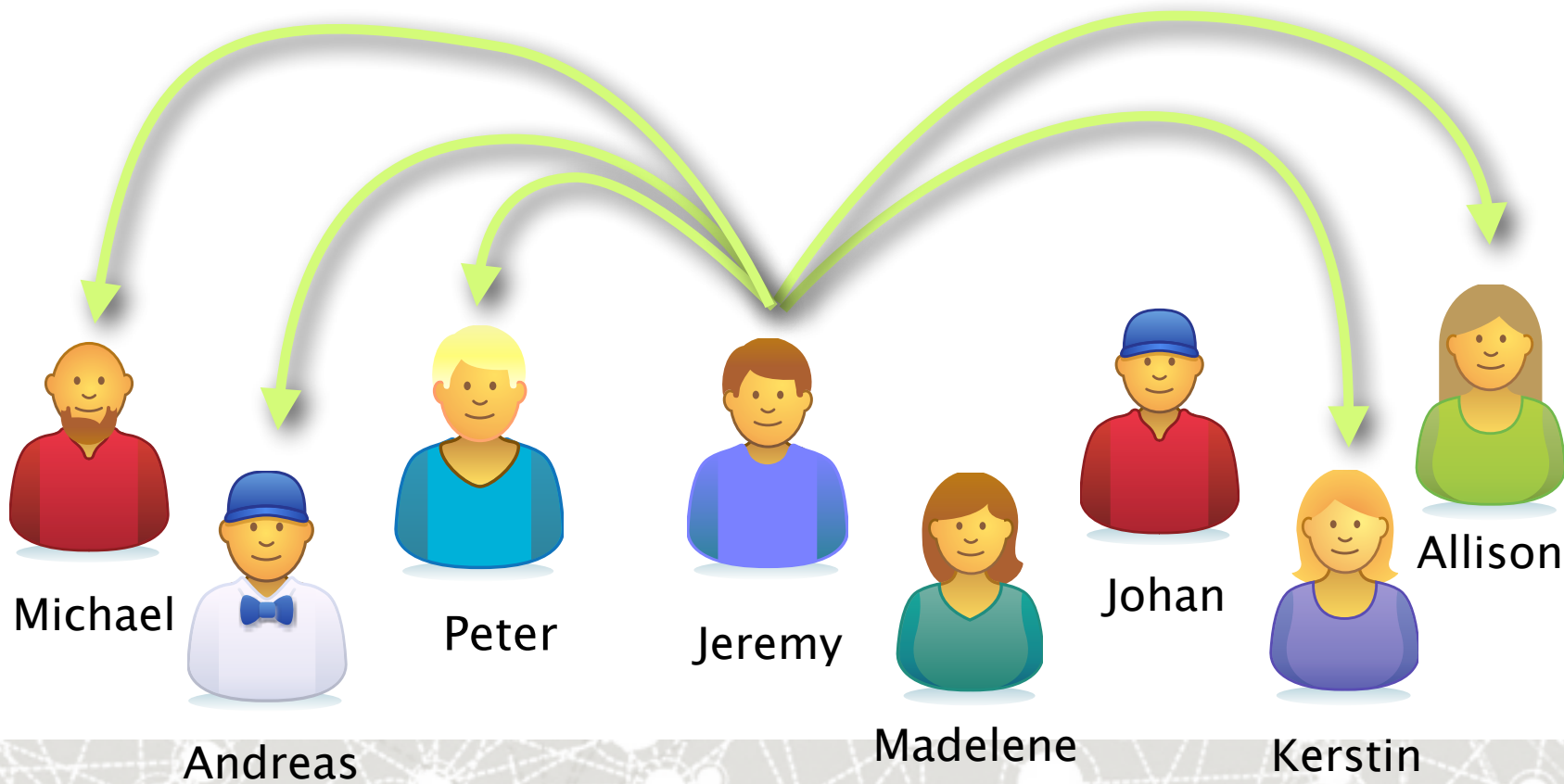


Kerstin



Allison

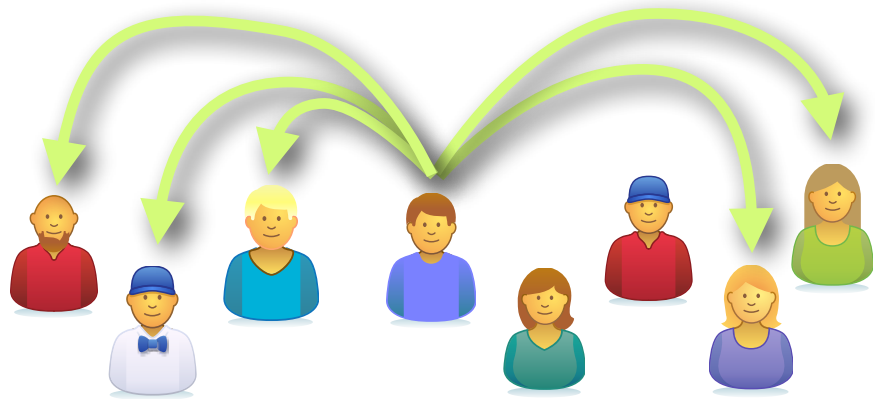
So we have a bunch of people



So we have a bunch of people



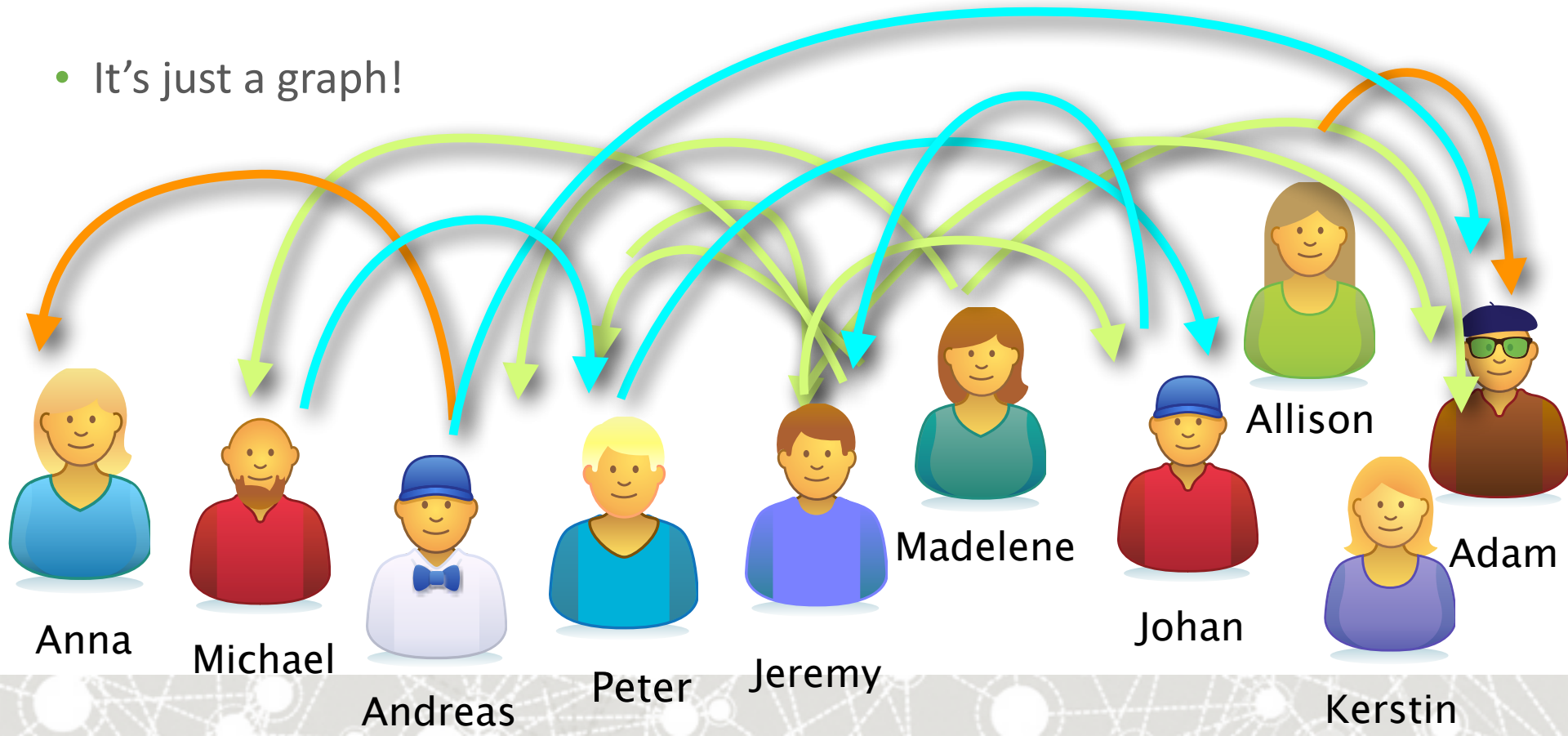
- How do we know they're friends?
- Either ask each pair: are you friends?
- Or we can add explicit connections
- Twitter, Facebook, LinkedIn, Snapchat etc



So we have a bunch of people



- It's just a graph!



Neo4j: The world's leading graph database



- optimised for the connections between records
- 'pre computed indexes' between records
- really, really fast at querying across records
- A relational database may tell you the average age of everyone here...

...but a graph database will tell you who's most likely to buy you a beer later



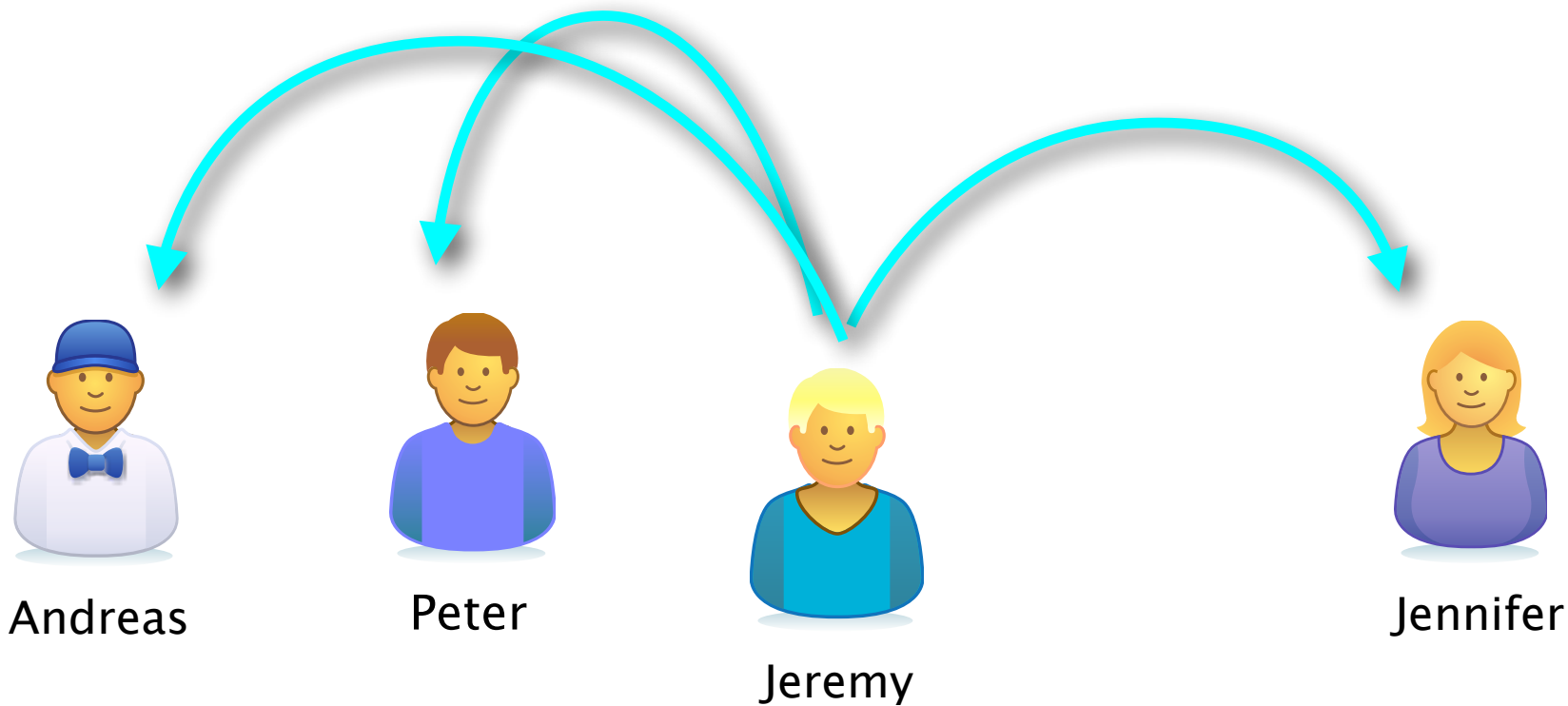
Friends of friends graph



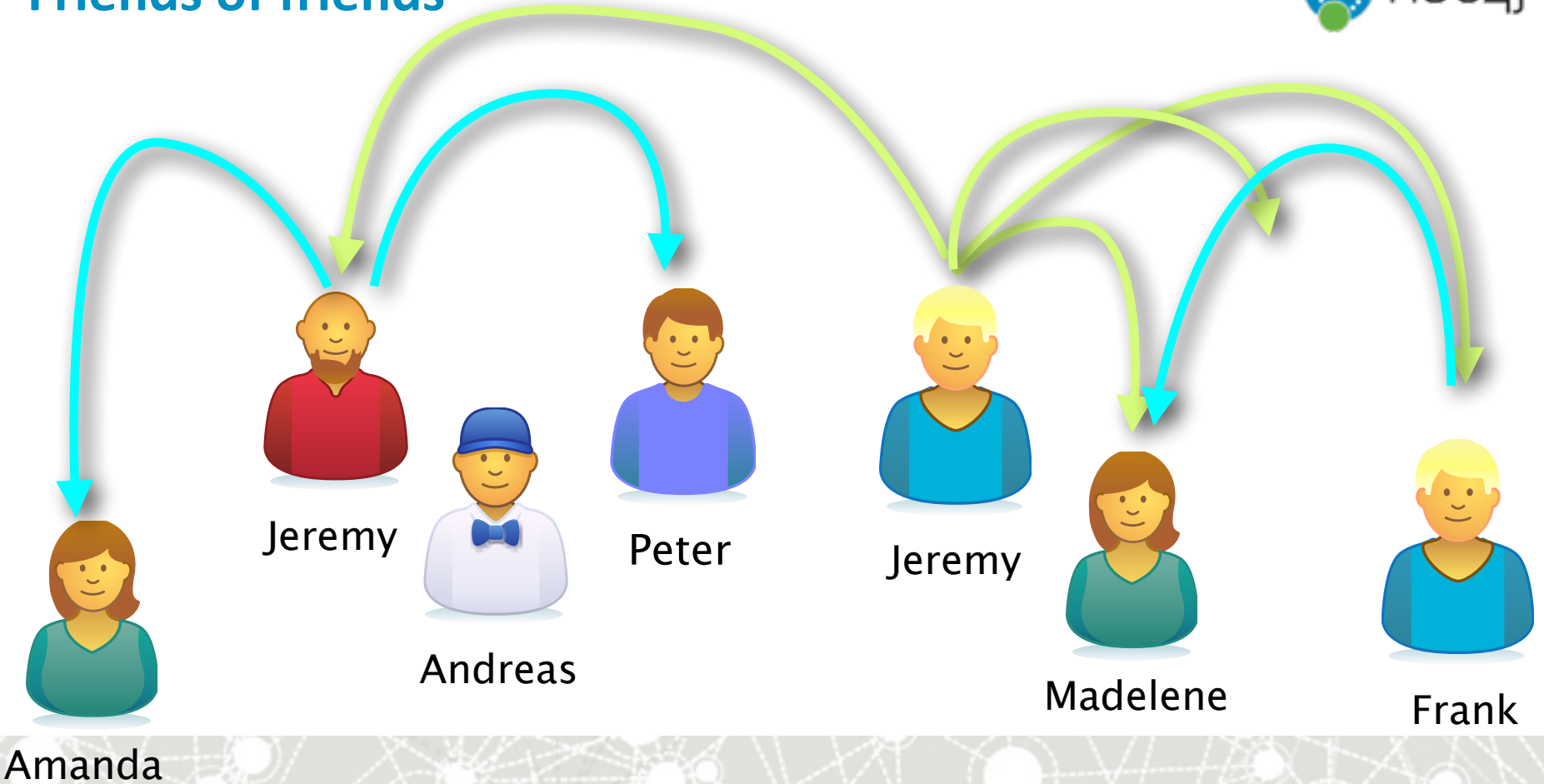
- According to SNAP Interactive if you are a female user, you have a:
 - 4% likelihood of interacting with a stranger
 - 10% likelihood of interacting with a friend of friend
 - 7% likelihood of interacting with a 3rd degree connection (friend of a friend)
 - Connections mean a much larger number of interactions = **#win**



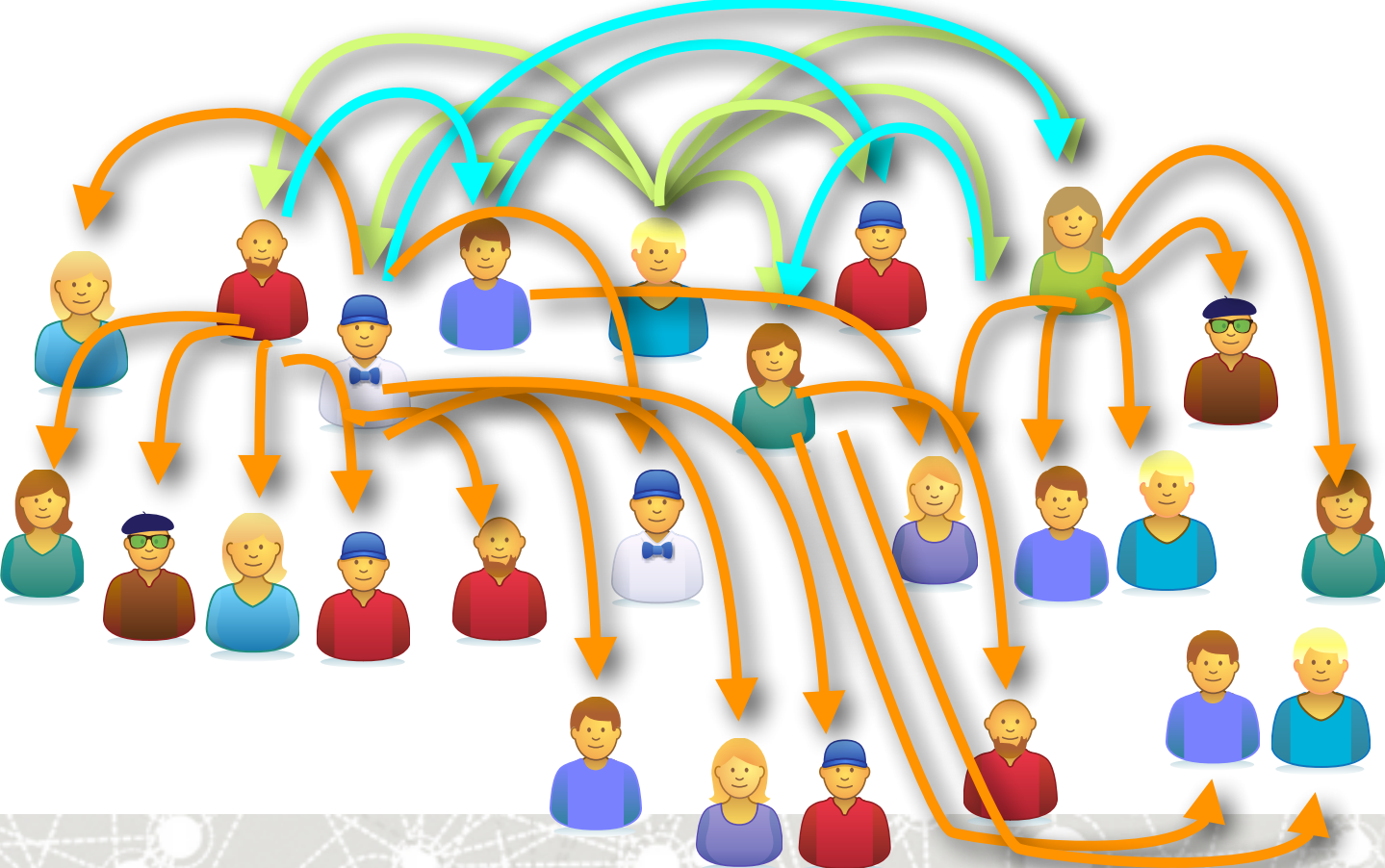
Friends



Friends of friends



Friends of friends of friends



Find Jeremy's FoFs



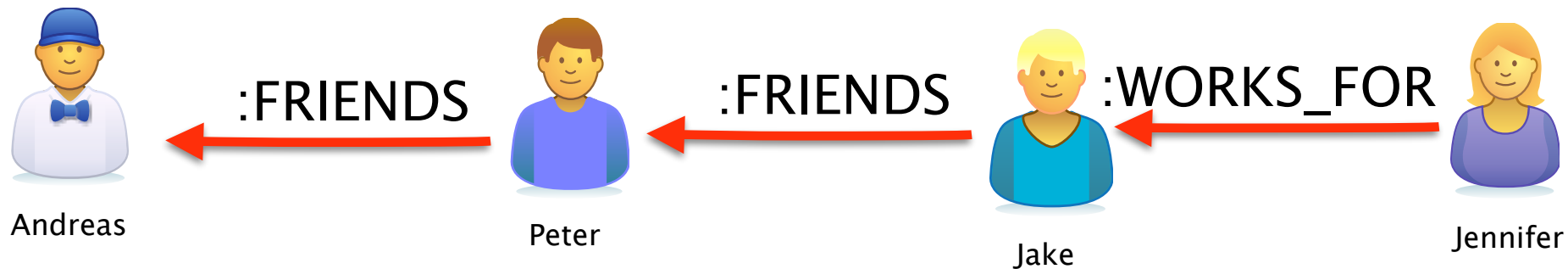
```
1 //Who are Jeremy's FoFs?  
2 MATCH (Jeremy:User{name:"Jeremy"})-[:KNOWS]->  
3 (friend)-[:KNOWS]->(fof)  
4 RETURN fof
```

Who does Jeremy shares the most friends with?

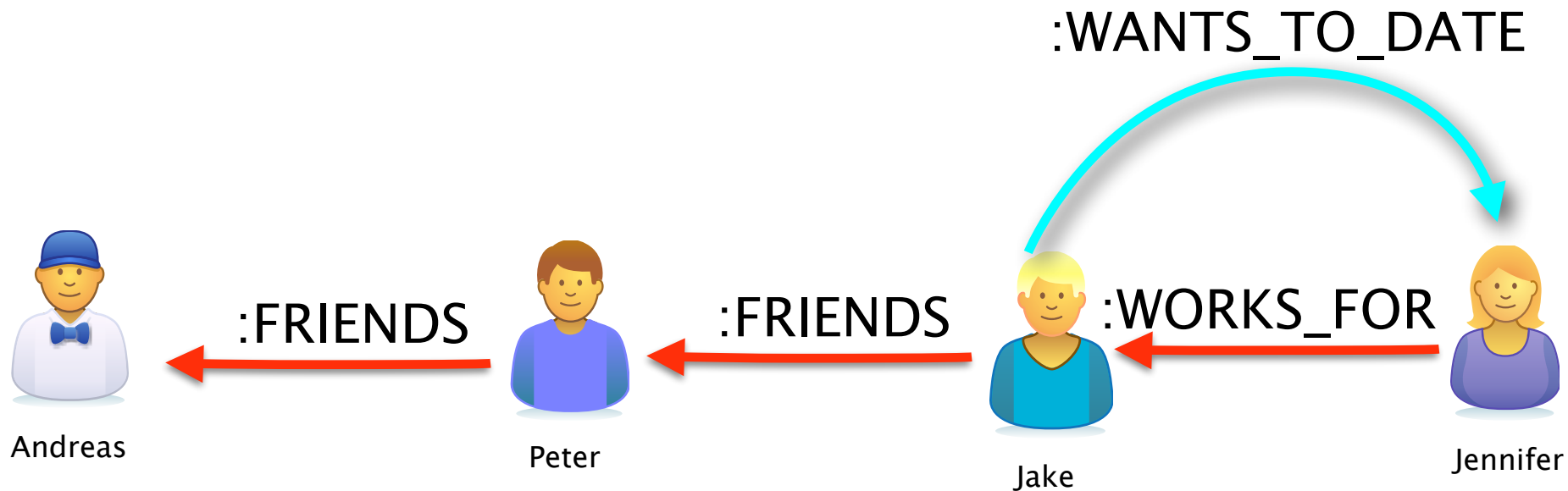


```
1 //Who does Jeremy share the most friends with?
2 MATCH (Jeremy:User{name:"Jeremy"})-[:KNOWS]->(friend)
3 -[:KNOWS]->(fof)
4 RETURN distinct fof,length(collect(friend)) as FriendsInCommon
5 ORDER BY FriendsInCommon desc
```

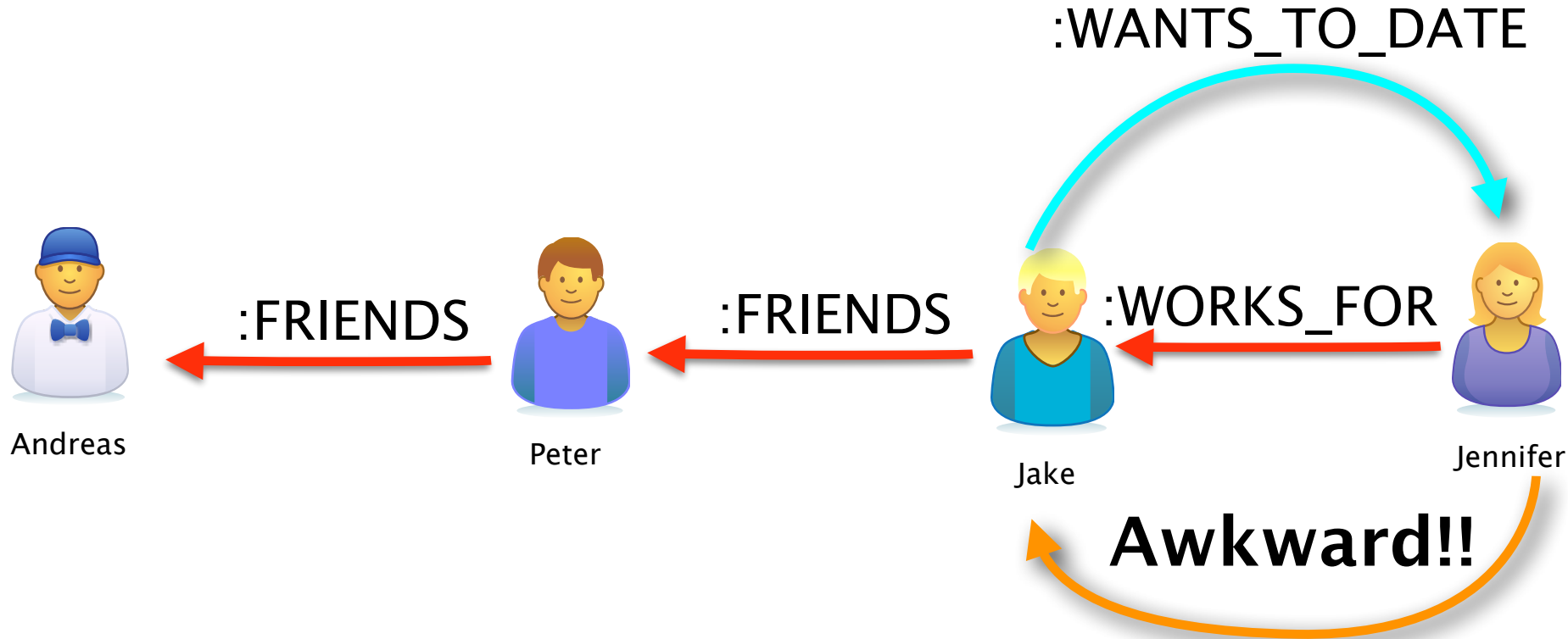
Complicated relationships



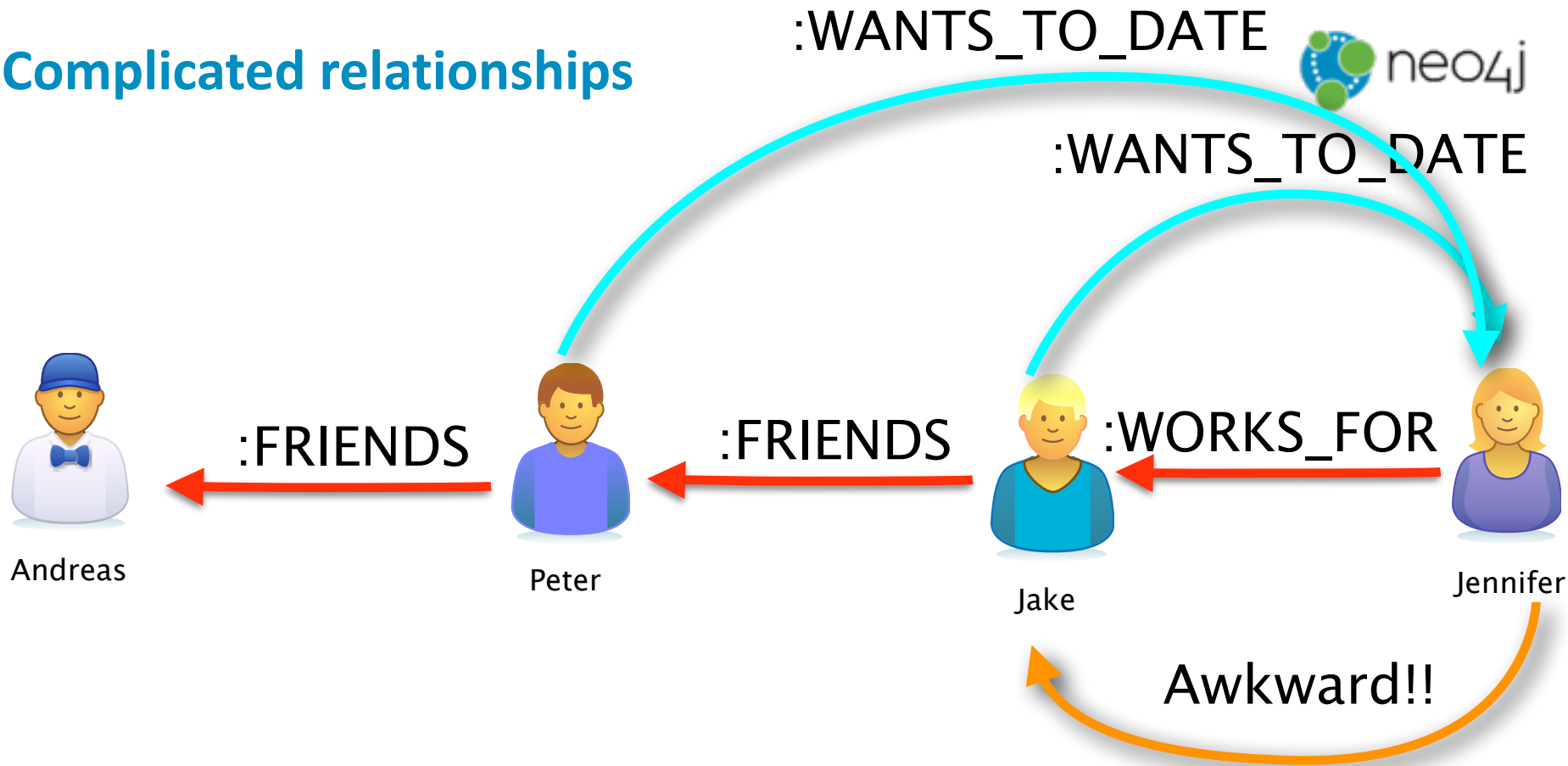
Complicated relationships



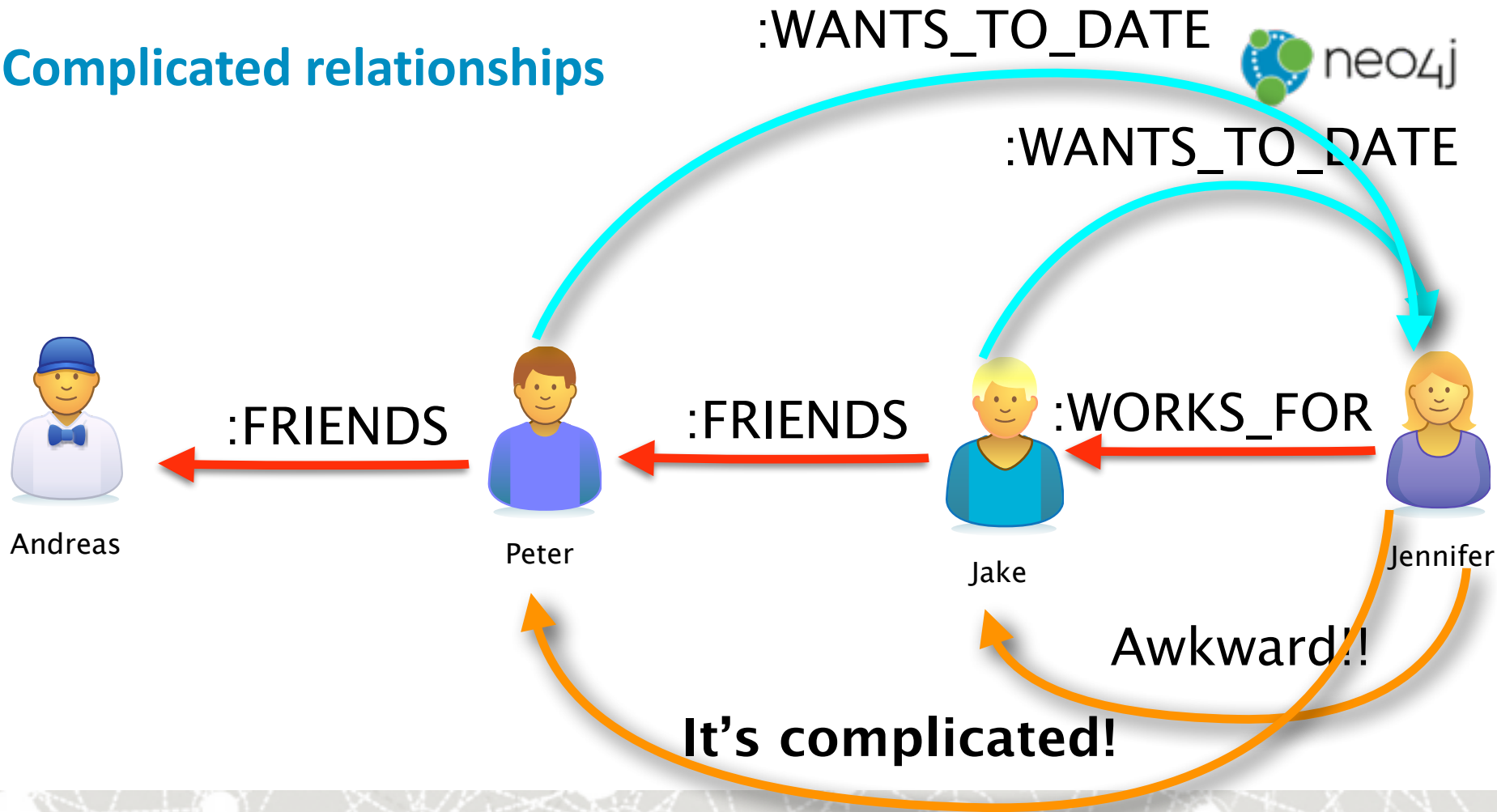
Complicated relationships



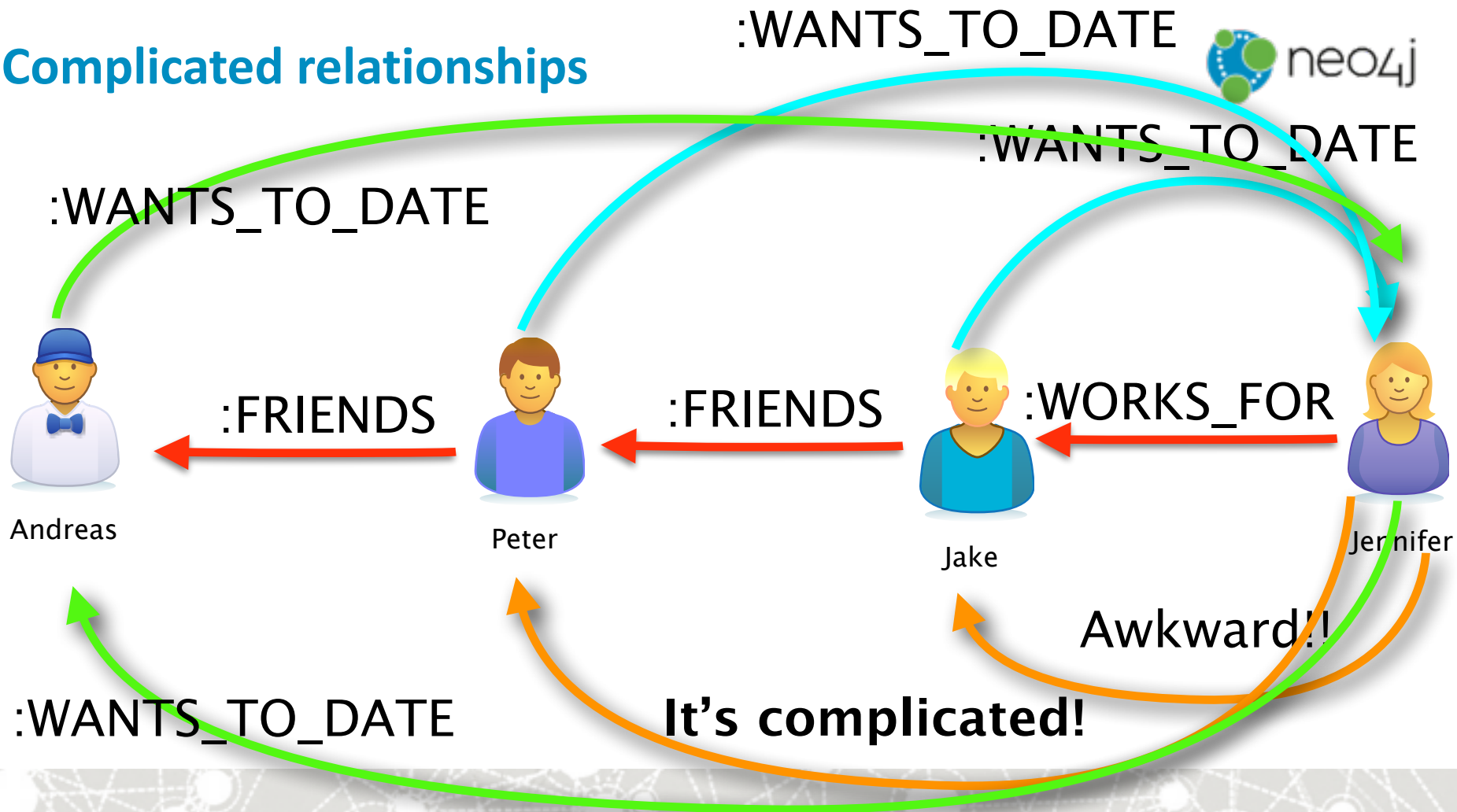
Complicated relationships



Complicated relationships



Complicated relationships



Friends of friends of friends



```
1 //FoFoF Long
2 MATCH (:User)-[:KNOWS]->(friend)-[:KNOWS]->(FoF)-
  [:KNOWS]->(FoFoF)
3 RETURN FoFoF
```

Friends of friends of friends



```
1 //FoFoF Long
2 MATCH (:User)-[:KNOWS]->(friend)-[:KNOWS]->(FoF)-
  [:KNOWS]->(FoFoF)
3 RETURN FoFoF
```

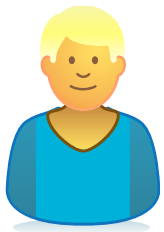
```
1 //FoFoF Short
2 MATCH (:User)-[:KNOWS*3]->(FoFoF)
3 RETURN FoFoF
```



Meet Jon...



- from: UK
- seeking: females
- appearance: hot, hot, hot!
- personality: fun loving, easy going
- interests: cooking, chemistry



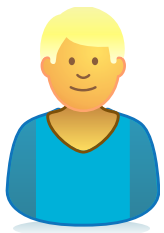
Jon

```
1 MATCH (Jon:User{name: 'Jonathan'})  
2 RETURN Jon
```

The passion graph



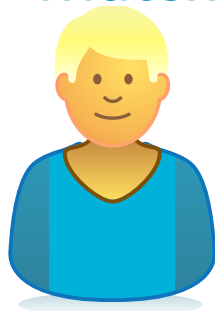
- Jon wants to find someone he can share his passions with



Jon

```
1 MATCH (Jon:User{name: 'Jonathan'})  
2 RETURN Jon
```

Match specific interests



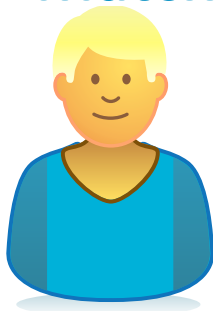
Jon



:REPORTED_INTEREST



Match specific interests



Jon

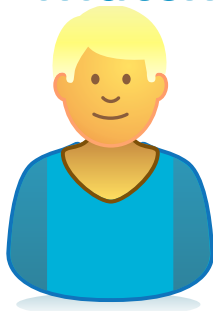


:REPORTED_INTEREST



```
1 MATCH (date)-[:REPORTED_INTEREST]->(interest)
2 WHERE interest.name = "Foodie" AND date.gender = 'F'
3 RETURN date
```

Match specific interests



Jon

:HAS_INTEREST



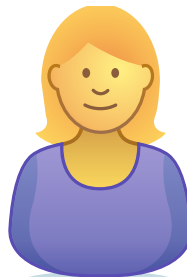
```
MATCH (Jeremy:User {name: "Jonathan"})-[:HAS_INTEREST]->(interest)
MATCH (interest)<-[:HAS_INTEREST]-(anotherPerson)
RETURN interest, anotherPerson
```



Anne



Julia

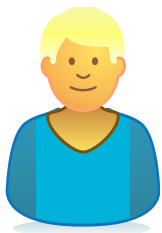


Jennifer

The location graph



- Jon wants to find a date but refuses to have a long distance relationship



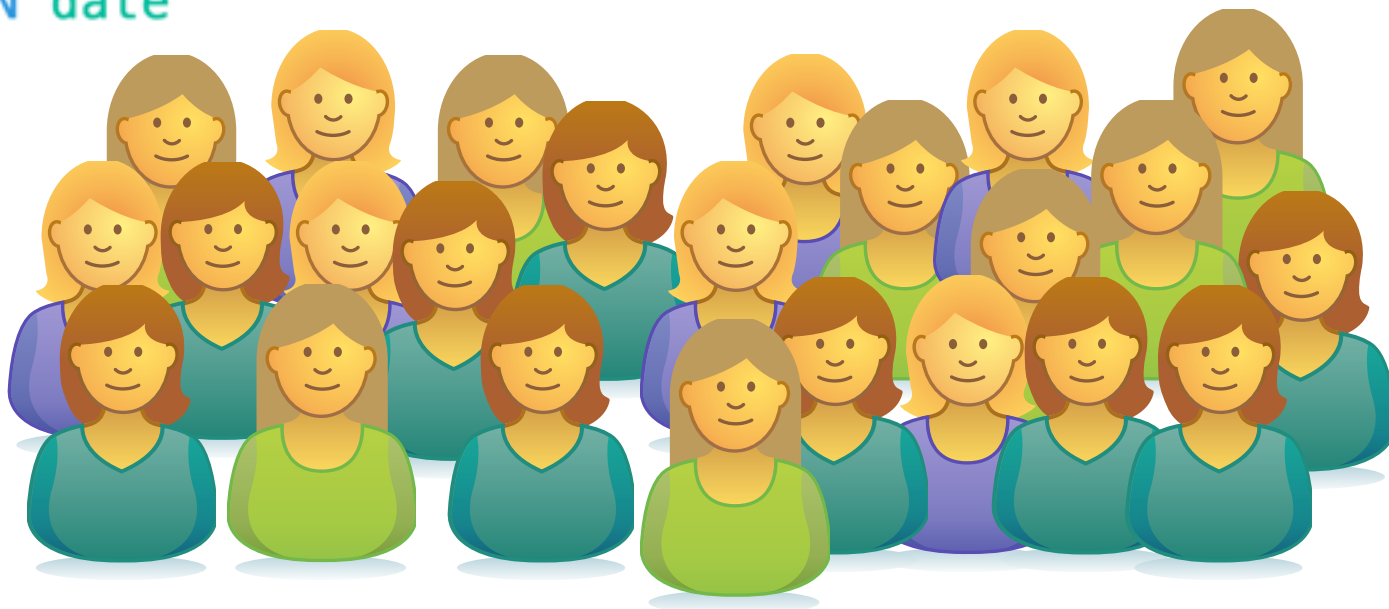
Jon

```
1 MATCH (Jon:User{name: 'Jonathan'})  
2 RETURN Jon
```

The location graph

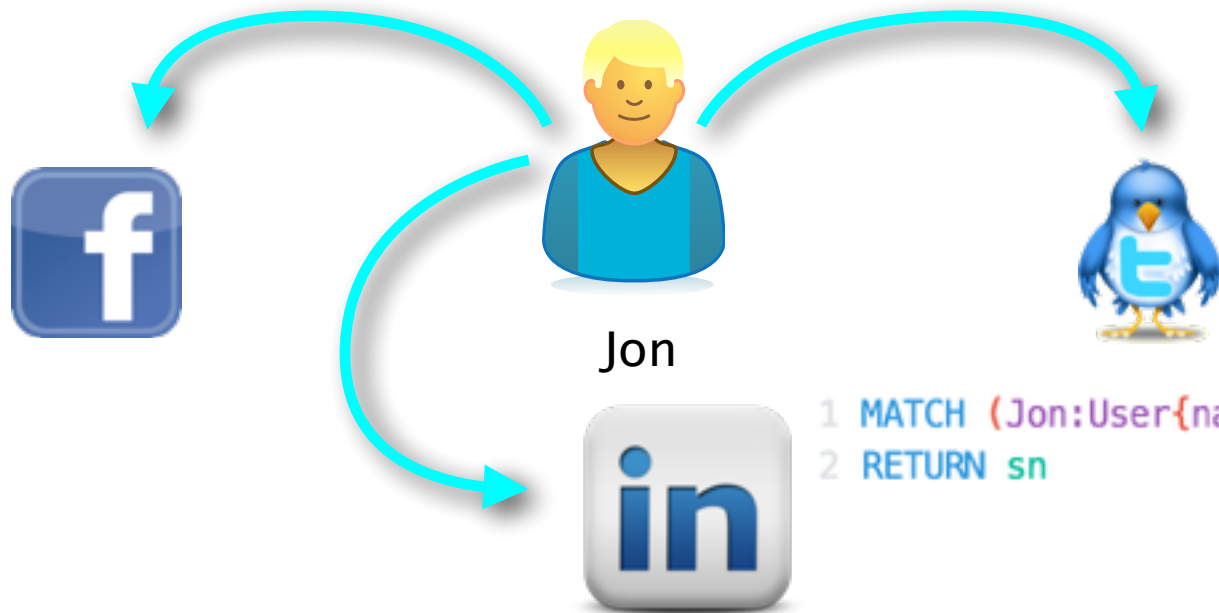


```
1 START date = node('withinDistance:[51.99,-0.19, 100.0]')
2 WHERE date.gender = 'F'
3 RETURN date
```



The safety graph

- Jon uses social networks



```
1 MATCH (Jon:User{name:'Jonathan'})-[:HAS_SN]-(sn)
2 RETURN sn
```

Let's dig into his Twitter



Jonathan Dreamyman

@JonathanDreamym



He follows some strange people...



BooBooLoon @YodasMethDealer
Doritos logo enthusiast.



Following



Fernando @ParisDrugLord
19 • Rap Shuffle • First Professionally Made Rap
Shuffle Magazine (80+Pages) Coming Soon • Ugly
But Wealthy By 30 • DON'T STEAL SHIT FROM MY
WEBSITE



Following



DrugLord Records @DrugLordRecords
NYC-based boutique record label with the
philosophy: CRIME IS LORD Artists: @TheLaBiancas
@JJBrine



Following



Drug Lord @the_high_life
This is not about weed. Butt connoisseur, Quote RT
me and you'll get blocked



Following



Drug Lord @BakedPrincess__
Beyoncé is a virgo & I am too, so basically I AM
Beyoncé. Flawless ✨



Following



Cocaine So White... @CocaineSoWhite
Original Cocaine So White Account



Following



Meth Tweets @DontMethWithMe
Meth 24/7 Email: methtweets@yahoo.com



Following



...and tweets about strange things!



Tweets



Jonathan Dreamyman @JonathanDreamym · now

Chemistry sucks in general but illegal drugs are fabulous.

Expand

↩ Reply 🗑 Delete ★ Favorite ... More



Jonathan Dreamyman @JonathanDreamym · 14m

Walter could add my secret ingredient, vanilla extract! I get lots of sales in the hood.

Expand

↩ Reply 🗑 Delete ★ Favorite ... More



Jonathan Dreamyman @JonathanDreamym · 17m

Walter White is my hero!

Expand

↩ Reply 🗑 Delete ★ Favorite ... More



Jonathan Dreamyman @JonathanDreamym · 17m

I am a huge fan of making drugs!

Expand

↩ Reply 🗑 Delete ★ Favorite ... More



Jonathan Dreamyman @JonathanDreamym · 18m

I met a drug dealer last weekend. He was awesome! His gf was quite a nice lady as well.

Expand

↩ Reply 🗑 Delete ★ Favorite ... More



Some basic word analysis



Tweets



Jonathan Dreamyman @JonathanDreamym · now

Chemistry sucks in general but illegal drugs are fabulous.

Expand

↩ Reply 🗑 Delete ★ Favorite ⋮ More



Jonathan Dreamyman @JonathanDreamym · 14m

Walter could add my secret ingredient, vanilla extract! I get lots of sales in the hood.

Expand

↩ Reply 🗑 Delete ★ Favorite ⋮ More



Jonathan Dreamyman @JonathanDreamym · 17m

Walter White is my hero!

Expand

↩ Reply 🗑 Delete ★ Favorite ⋮ More



Jonathan Dreamyman @JonathanDreamym · 17m

I am a huge fan of making drugs!

Expand

↩ Reply 🗑 Delete ★ Favorite ⋮ More



Jonathan Dreamyman @JonathanDreamym · 18m

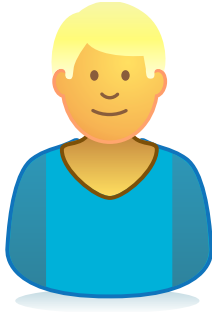
I met a drug dealer last weekend. He was awesome! His gf was quite a nice lady as well.

Expand

↩ Reply 🗑 Delete ★ Favorite ⋮ More



Let's update based on behaviour



Jon

:DEMONSTRATED_INTEREST



```
1 CREATE (Jonathan:Person{name:"Jonathan Dreamyman"})
2 -[:DEMONSTRATED_INTEREST]->
3 (drugs:Interest{name:"Drugs"})
```

Any ladies ok with this?



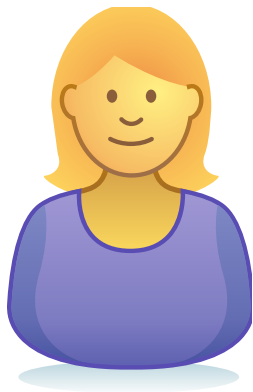
```
1 //Jon's FoFs who like Drugs
2 MATCH (date:User)-[:KNOWS*2]-(Jon:User{name : 'Jonathan'}),
3 (date)-[*..2]-(Drugs:Interest{name: 'Drugs'})
4 WHERE date.gender = 'F'
5 RETURN distinct date
```



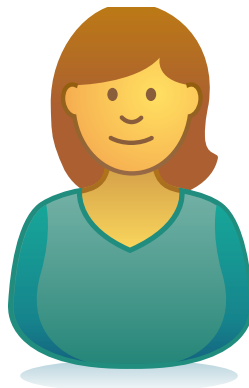
Any ladies ok with this?



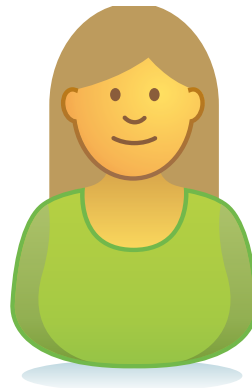
```
1 //Jon's FoFs who like Drugs
2 MATCH (date:User)-[:KNOWS*2]-(Jon:User{name : 'Jonathan'}),
3 (date)-[*..2]-(Drugs:Interest{name:'Drugs'})
4 WHERE date.gender = 'F'
5 RETURN distinct date
```



Jennifer



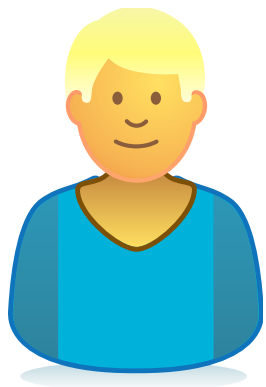
Jane



Maria

The passion graph

- Jon loves Ajax

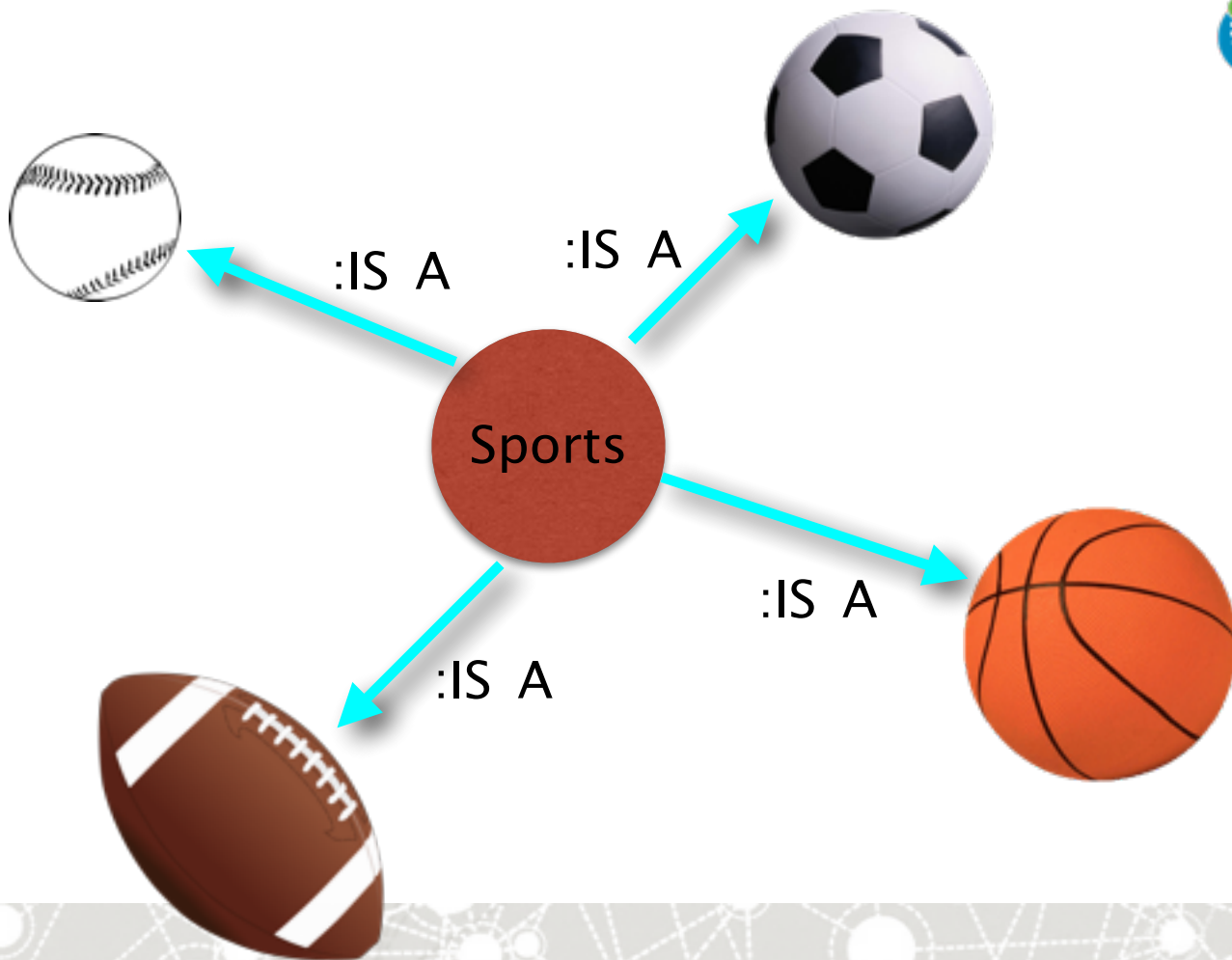


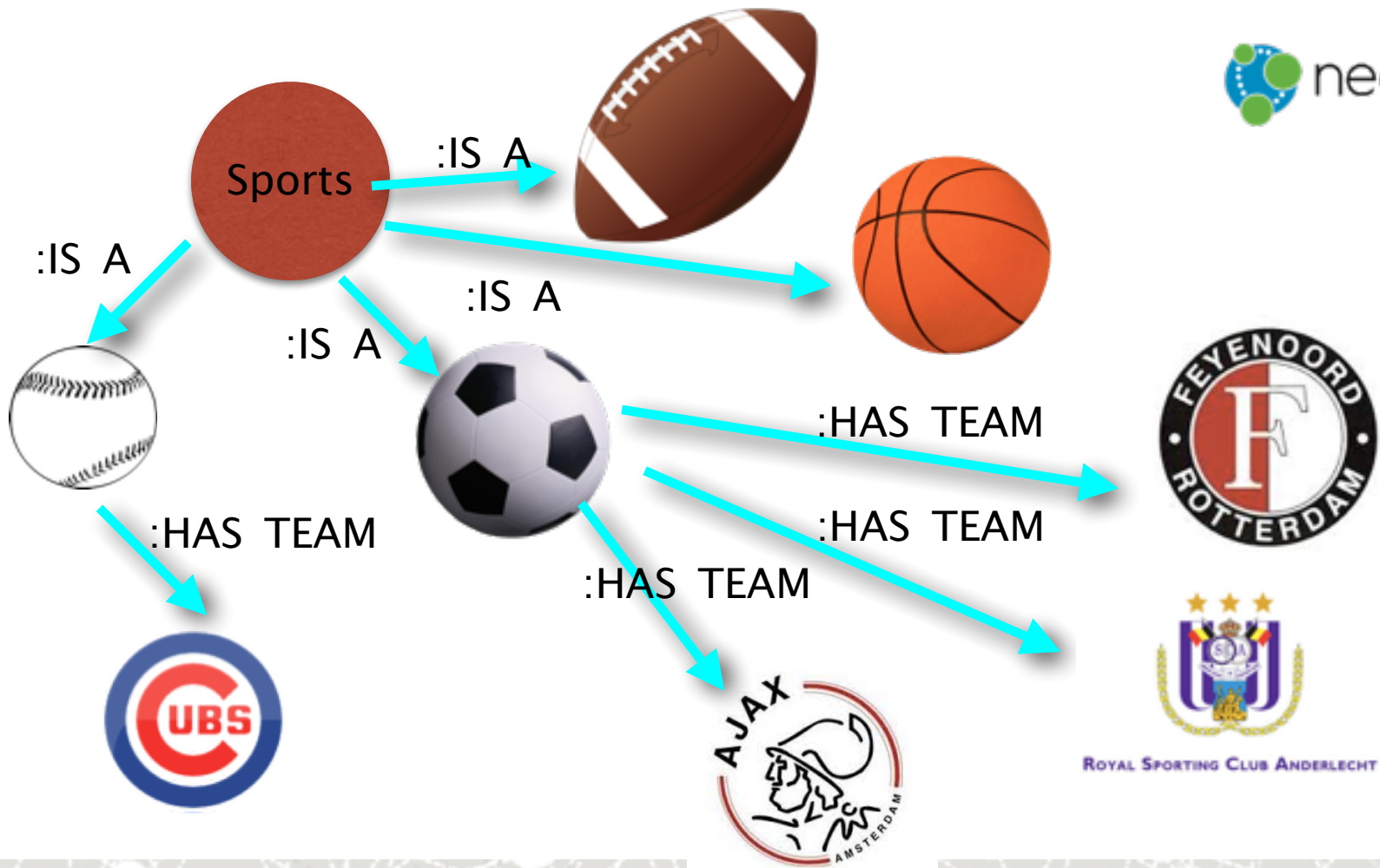
Jon

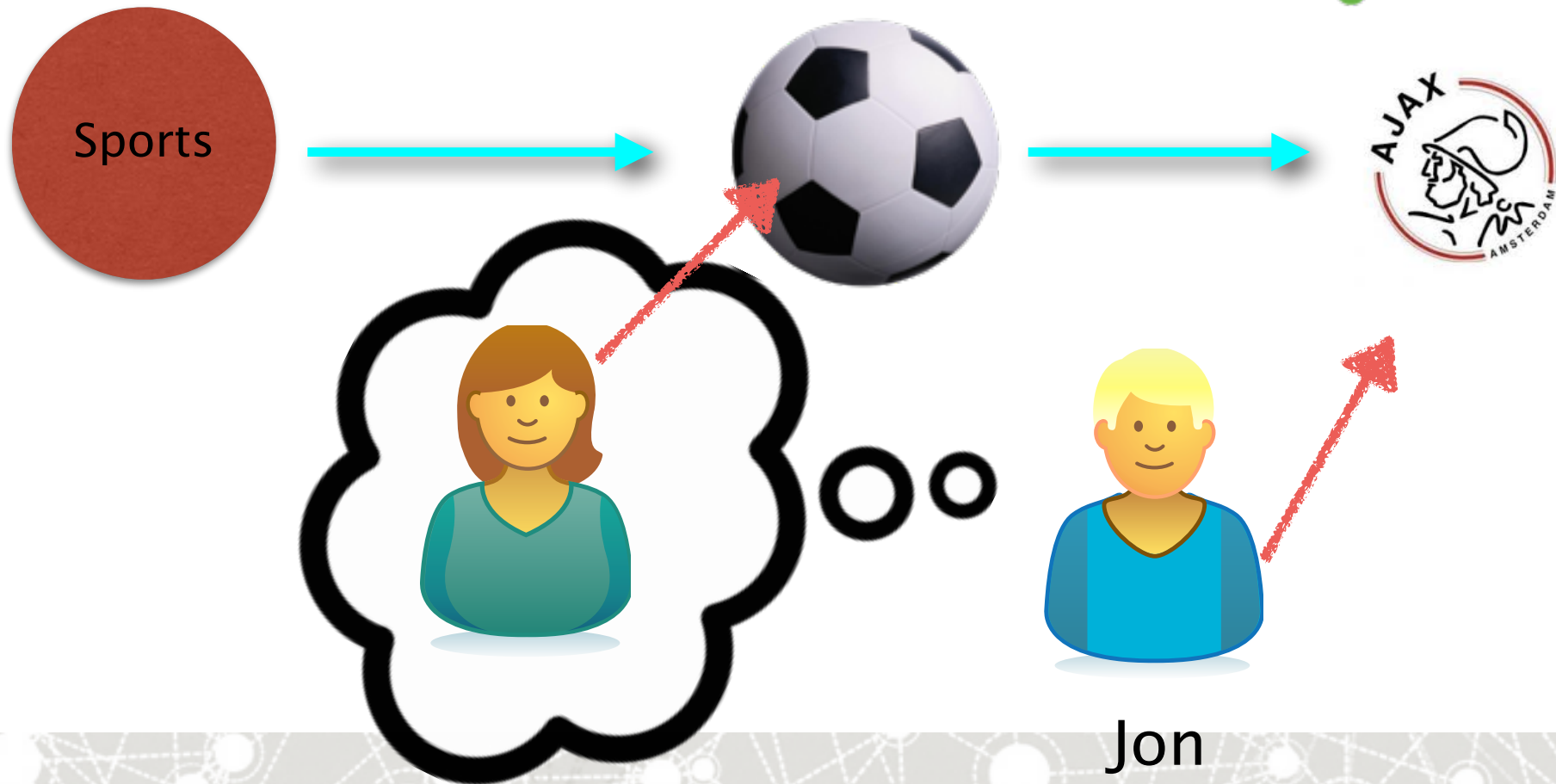


:HAS_INTEREST









- Jon has no luck with online dating. All of his interactions are with spam profiles



The poser graph



- Let's find real people with at least 1 social network and a minimum of 2 posts



Find ladies who aren't spam bots



```
1 //Find real people
2 MATCH (date:User)-[:KNOWS*2]-(Jon:User{name : 'Jonathan'}),
3 (date)-[r:HAS_SN]->(SN)
4 WHERE date.gender = 'F' AND r.posts > 2
5 RETURN distinct date
```

Find ladies who like football

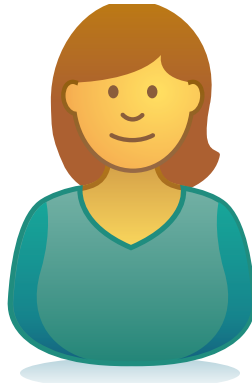


```
1 // Jon's FoF's who like football
2 MATCH (date:User)-[:KNOWS*2]-(Jon:User{name: 'Jonathan'}),
3       (date)-[*..2]-(a:Interest{name: "Football"})
4 WHERE date.gender = 'F'
5 RETURN distinct date
```

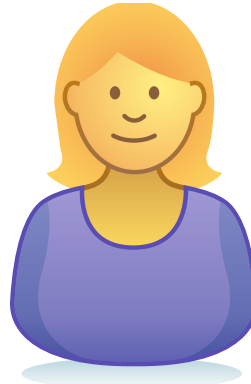
Find ladies who like football



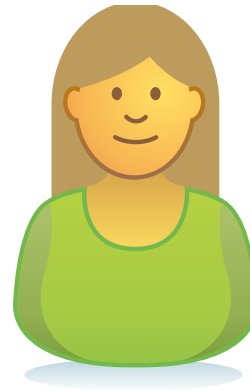
```
1 // Jon's FoF's who like football
2 MATCH (date:User)-[:KNOWS*2]-(Jon:User{name: 'Jonathan'}),
3       (date)-[*..2]-(a:Interest{name: "Football"})
4 WHERE date.gender = 'F'
5 RETURN distinct date
```



Katie



Jennifer



Greta

Find Jon's perfect date

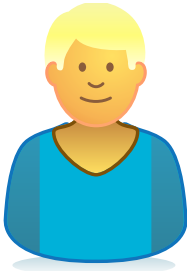


```
1 MATCH (date:User)-[:KNOWS*2]-(Jon:User{name: 'Jonathan'}),
2     (date)-[:HAS_INTEREST]->(food:Interest{name: 'Foodie'}),
3     (date)-[*..2]-(a:Interest{name: 'Football'}),
4     (date)-[*..2]-(Drugs:Interest{name: 'Drugs'})
5 WHERE date.gender = 'F'
6 RETURN distinct date
```

Find Jon's perfect date



```
1 MATCH (date:User)-[:KNOWS*2]-(Jon:User{name: 'Jonathan'}),  
2     (date)-[:HAS_INTEREST]->(food:Interest{name: 'Foodie'}),  
3     (date)-[*..2]-(a:Interest{name: 'Football'}),  
4     (date)-[*..2]-(Drugs:Interest{name: 'Drugs'})  
5 WHERE date.gender = 'F'  
6 RETURN distinct date
```



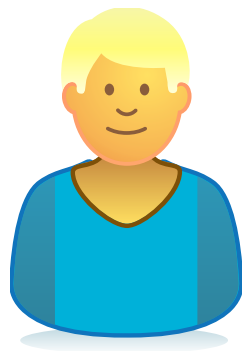
Jon



:PERFECT_FOR



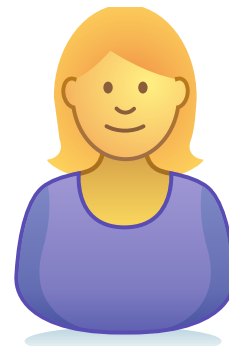
Jennifer



Jon



:HAS_DATE_WITH



Jennifer



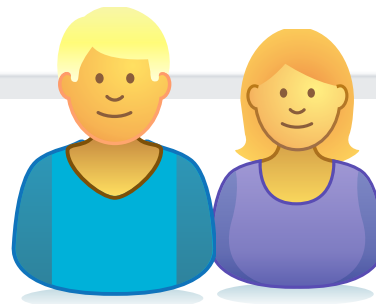
Jon and Jennifer delete their profiles and sail off into the sun set



```
1 MATCH (Jon:User{name:'Jonathan'})-[a]-(),(Jen:User{name:'Jennifer'})-[b]-()
2 DELETE a,b,Jon,Jen
```

CYPHER MATCH (Jon:User{name:'Jonathan'})-[a]-(),(Jen:User{name:'Jennifer'})-[b]-() DELETE a,b,Jon,Jen

✓ Deleted 2 nodes, deleted 16 relationships, returned 0 rows in 165 ms



Jon Jennifer

What's that got to do with Business?

Take a look at “Fraud”



Why Fraud?

Types of Fraud

- First-Party Fraud
- Synthetic Identities and Fraud Rings
- Insurance Fraud
- Ecommerce Fraud

Types of Analysis

- Traditional Analysis
- Graph-Based Analysis

Fraud Detection and Prevention



First-Party Fraud



- Fraudster's aim: apply for lines of credit, act normally, extend credit, then...run off with it
- Fabricate a network of synthetic IDs, aggregate smaller lines of credit into substantial value
- Often a hidden problem since only banks are hit
 - Whereas third-party fraud involves customers whose identities are stolen



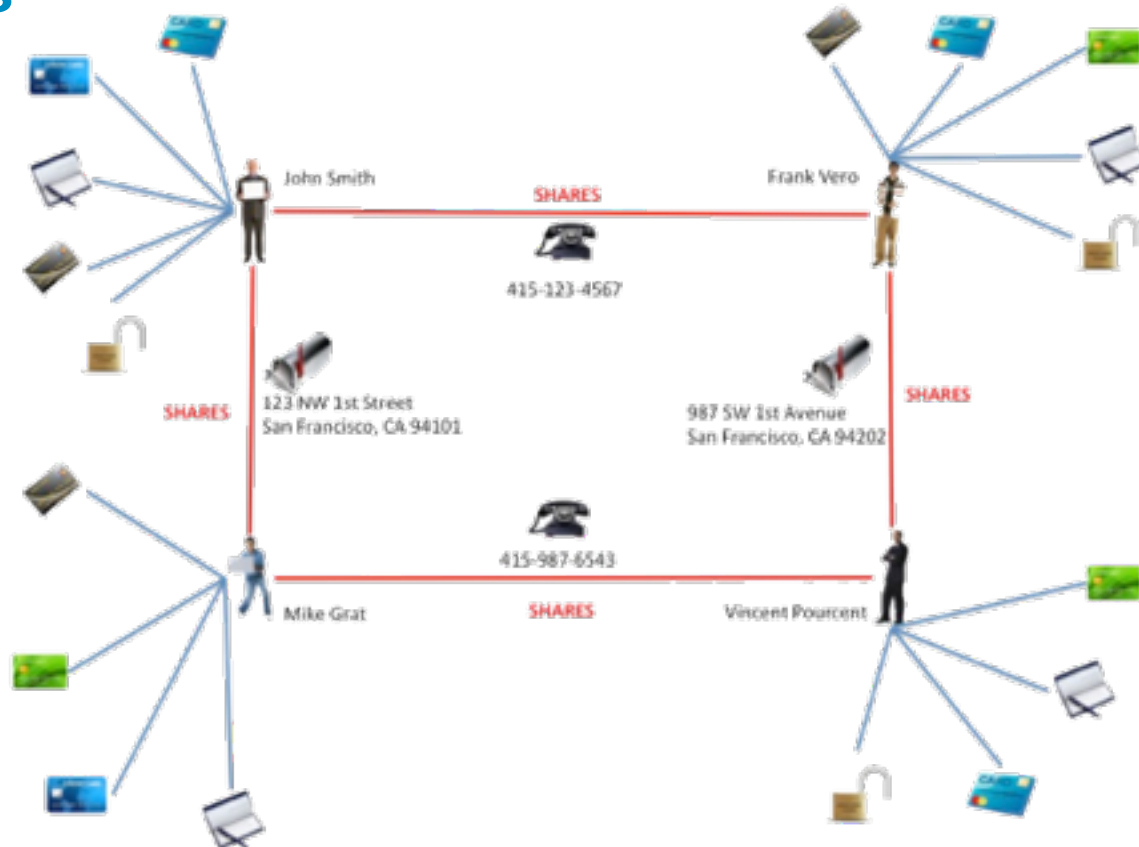
So what?



- \$10's **billions** lost by US banks **every year**
- **25%** of the total consumer credit write-offs in the USA
- Around **20%** of unsecured bad debt in EU and USA is misclassified
 - In reality it is first-party fraud

These are **enormous**
numbers

Fraud Ring



Then the fraud happens...



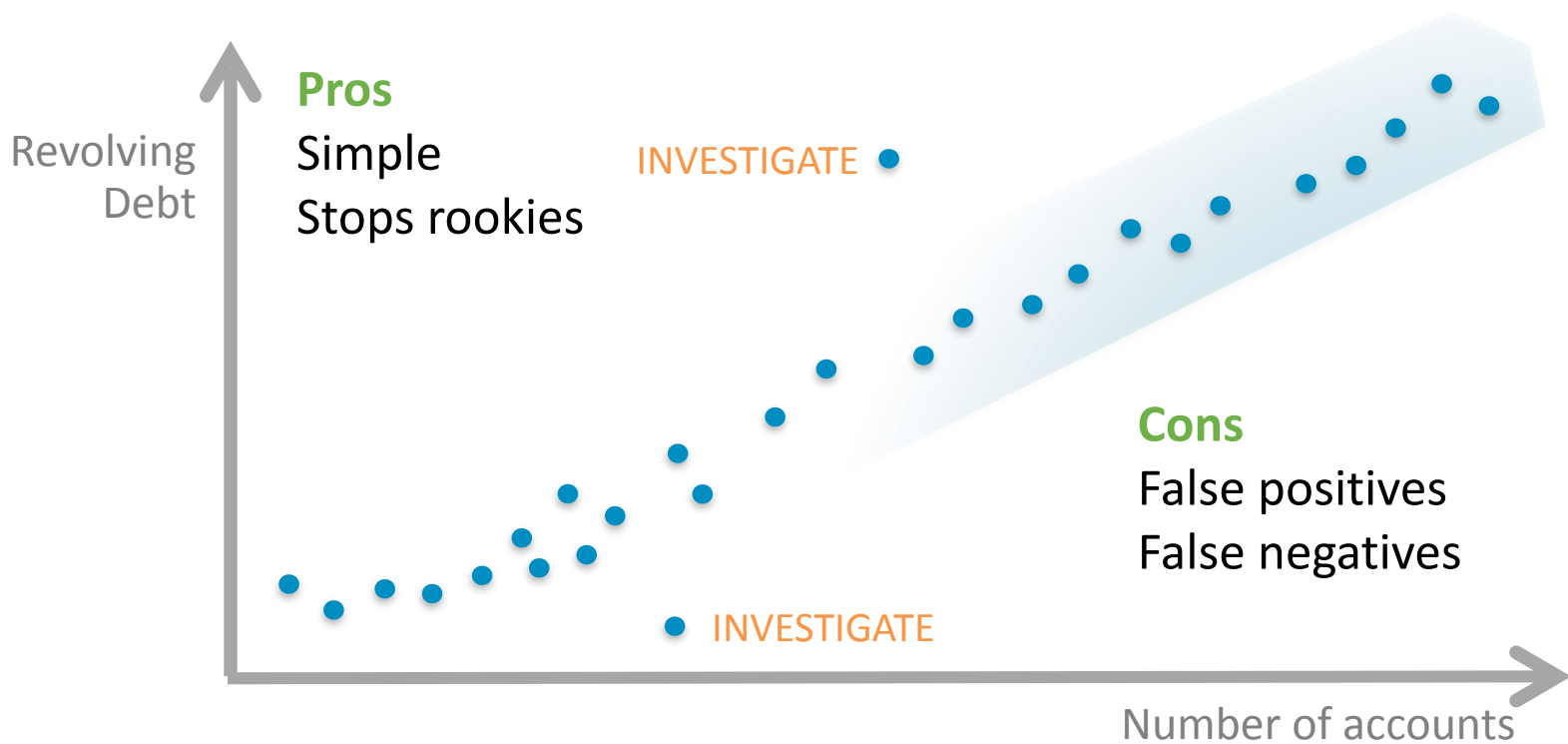
- Revolving doors strategy
 - Money moves from account to account to provide legitimate transaction history
- Banks duly increase credit lines
 - Observed responsible credit behaviour
- Fraudsters max out all lines of credit and then **bust out**

... and the bank loses



- Collections process ensues
 - Real addresses are visited
 - Fraudsters deny all knowledge of synthetic IDs
 - Bank writes off debt
- Two fraudsters can easily rack up \$80k
- Well organised crime rings can rack up many times that

Discrete Data Analysis fails to predict...



...and Makes it Hard to React



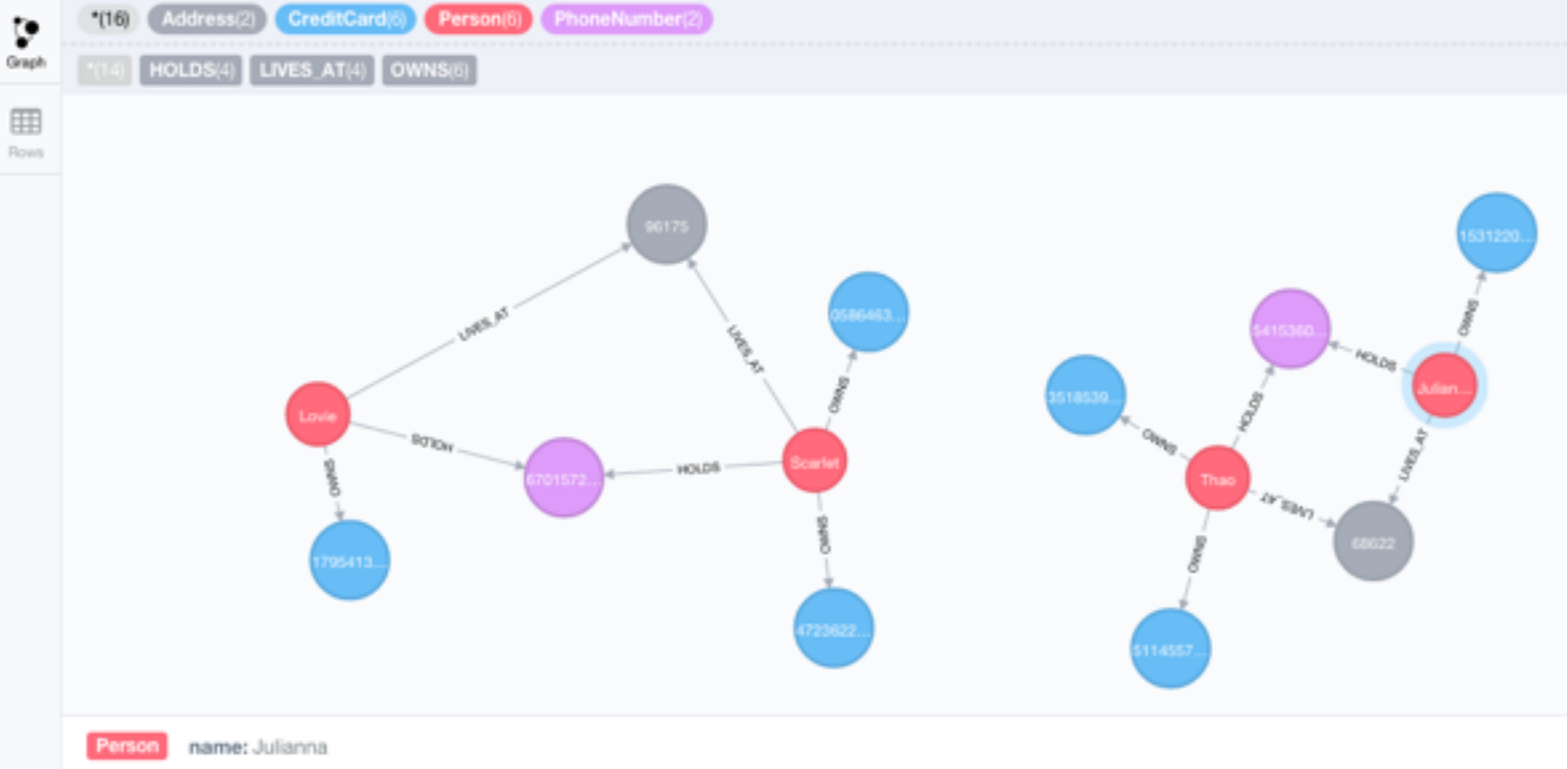
- When the bust out starts to happen, how do you know what to cancel?
- And how do you do it faster than the fraudster to limit your losses?
- A graph, that's how!



Probably Non-Fraudulent Cohabitors



```
$ MATCH (p1:Person)-[:HOLDS|LIVES_AT*]->()-[:HOLDS|LIVES_AT*]-(p2:Person) WHERE p1 <> p2 RETURN p1 LIMIT 10
```

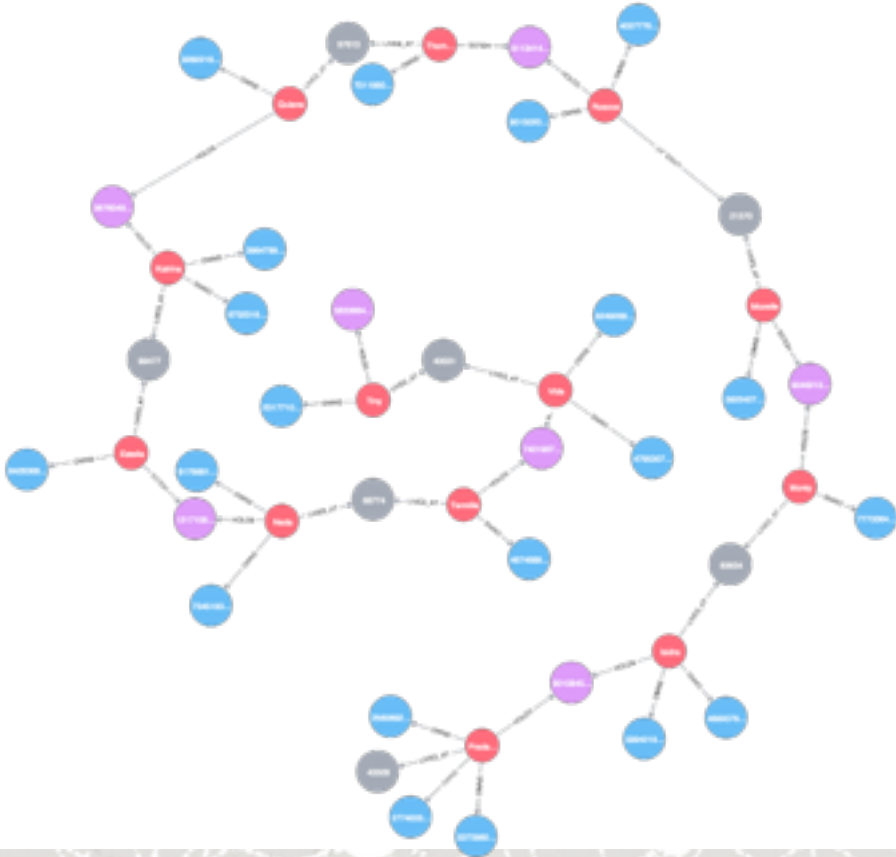


Probable Cohabitors Query



```
MATCH (p1:Person) -[:HOLDS|LIVES_AT*] -> ()
      <-[:HOLDS|LIVES_AT*] - (p2:Person)
WHERE p1 <> p2
RETURN DISTINCT p1
```

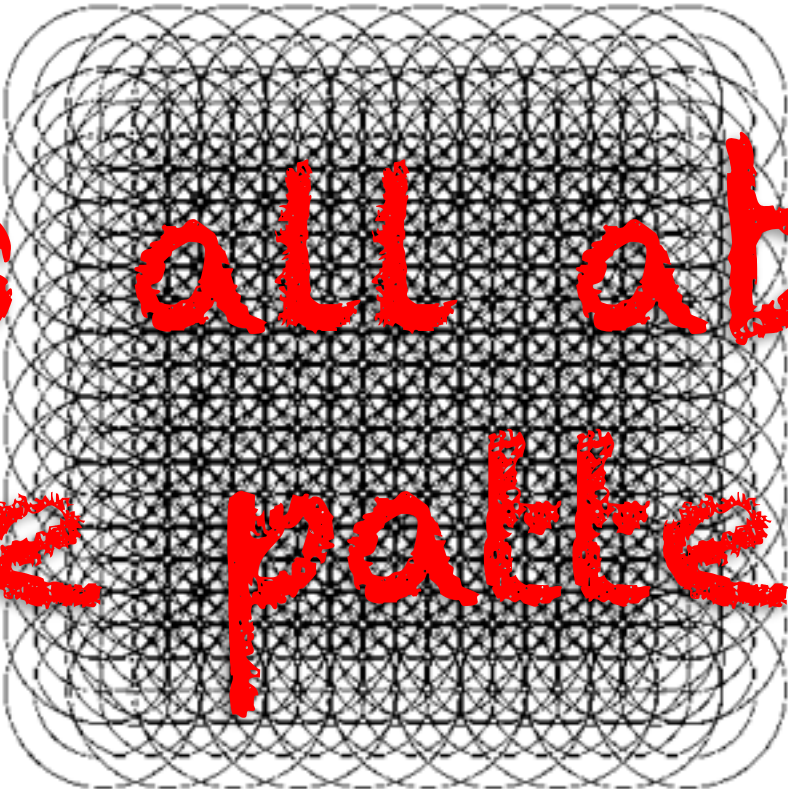
Dodgy-Looking Chain



Risky Card Numbers



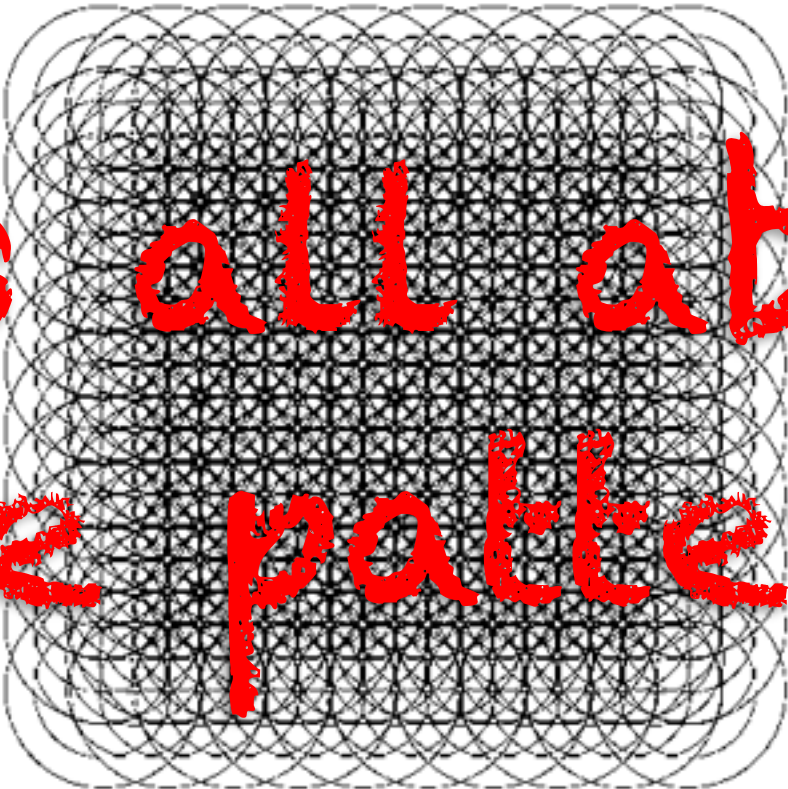
```
MATCH (p1:Person)-[:HOLDS|LIVES_AT*]->()  
      <-[:HOLDS|LIVES_AT*]-(p2:Person)  
      -[:HOLDS|LIVES_AT*]->()  
      <-[:HOLDS|LIVES_AT*]-(p3:Person)  
WHERE p1 <> p2 AND p2 <> p3 AND p3 <> p1  
WITH p1, p2, p3  
MATCH (p1)-[:OWNS]->(c1:CreditCard),  
      (p2)-[:OWNS]->(c2:CreditCard),  
      (p3)-[:OWNS]->(c3:CreditCard)  
UNWIND [c1, c2, c3] AS creditCardNumbers  
RETURN creditCardNumbers
```



It's all about
the patterns

Let me just repeat that



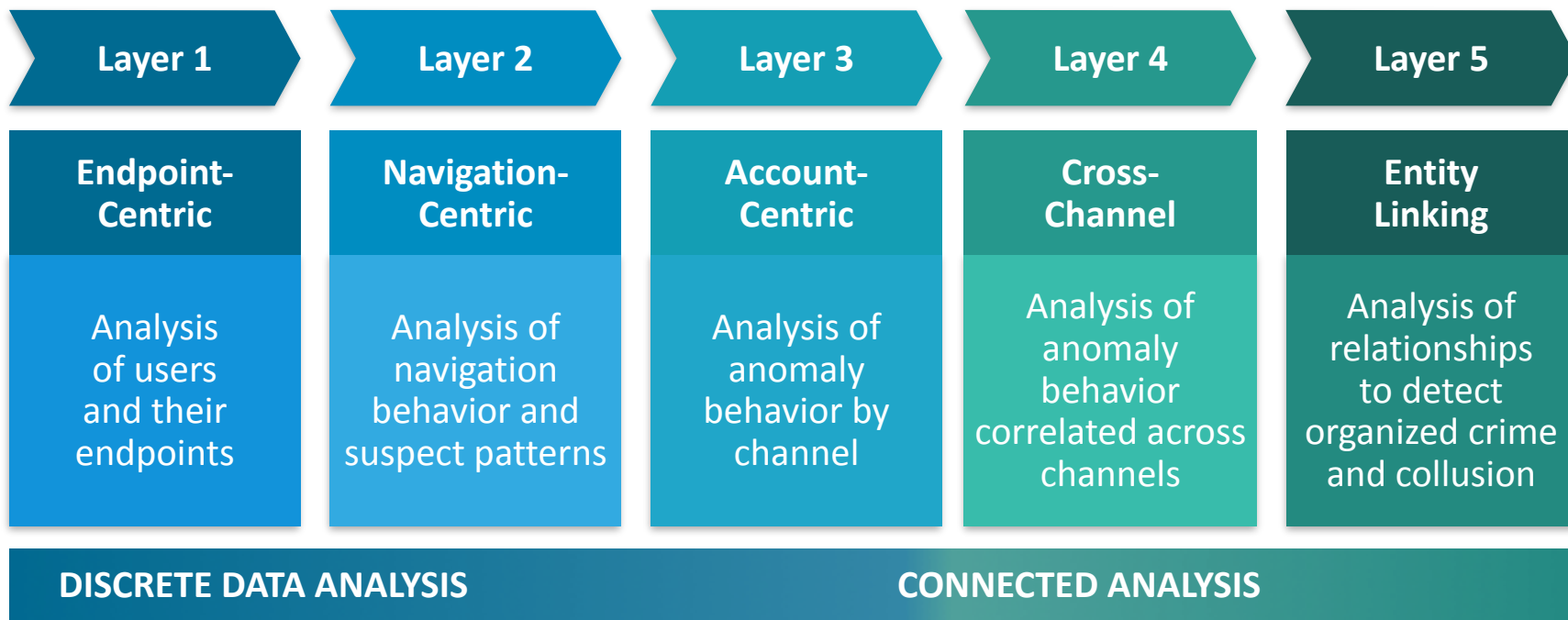


It's all about
the patterns

How does this fit with traditional fraud prevention?



Gartner's Layered Fraud Prevention Approach ⁽⁴⁾



(4) <http://www.gartner.com/newsroom/id/1695014>





Ask for help if you get stuck



- Online training - <http://neo4j.com/graphacademy/>
- Videos - <http://vimeo.com/neo4j/videos>
- Use cases - <http://www.neotechnology.com/industries-and-use-cases/>
- Meetups every week in Southwark
- Book @ <http://www.graphdatabases.com>



Or ping me?



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Thanks for listening

