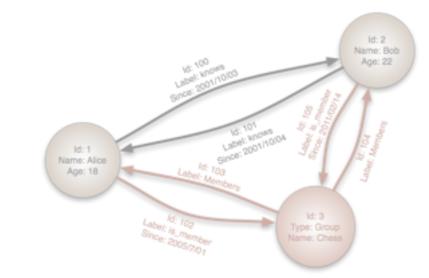
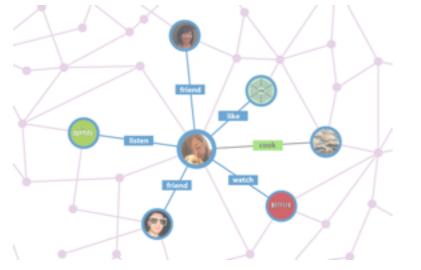




# Graph Databases Use Cases

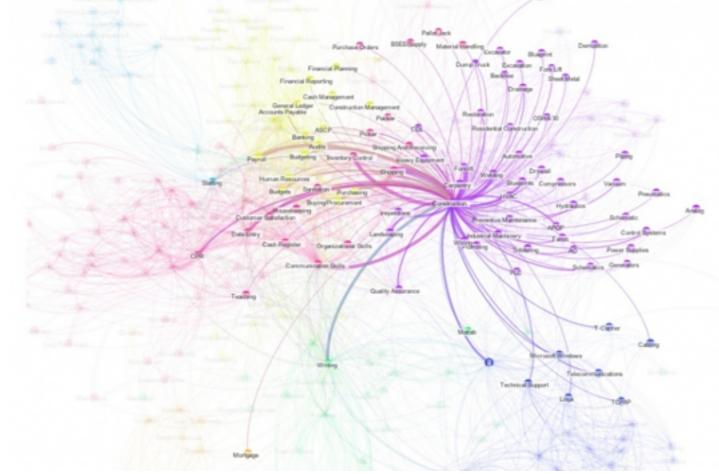


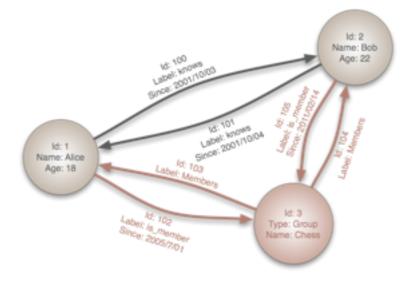






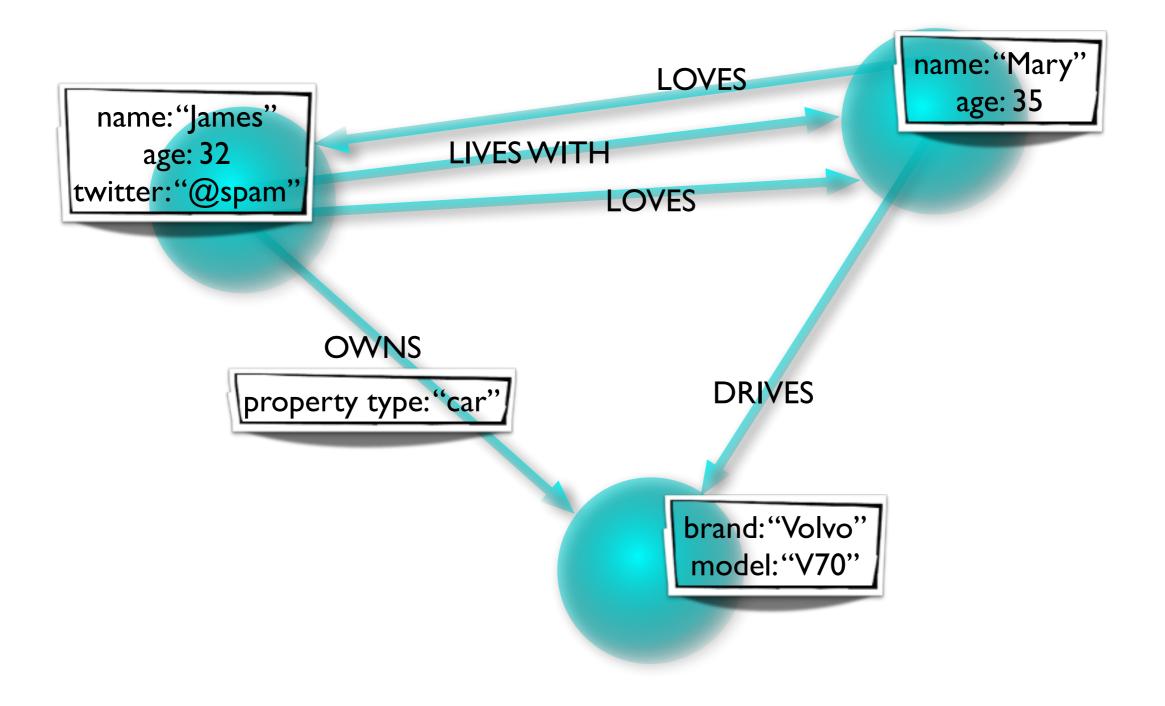
## What's a Graph?







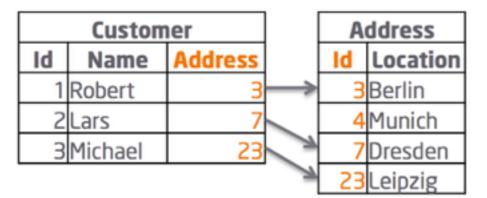
## Graph data model





## **Relational Tables**

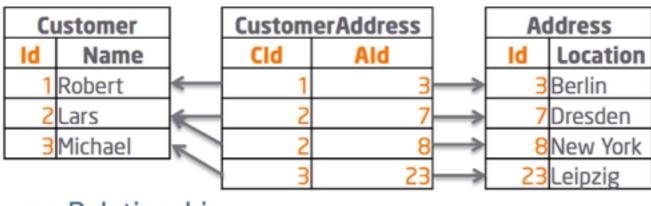
### SQL Join Hell (1)



	Customer		]	Address		
Id		Name		Id	Customer	Location
	1	Robert	<	3	1	Berlin
	2	Lars		7	2	Dresden
	3	Michael	7	8	2	New York
				23	3	Leipzig

#### 1:1 Relationship

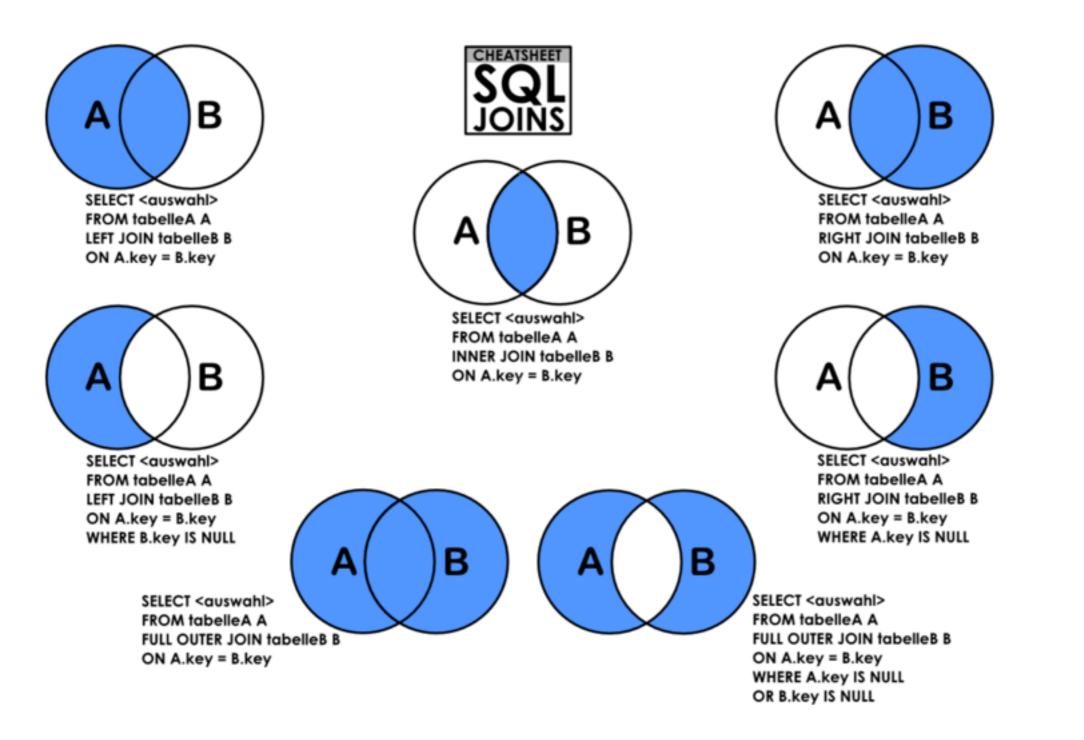




m:n Relationship



# Join this way...

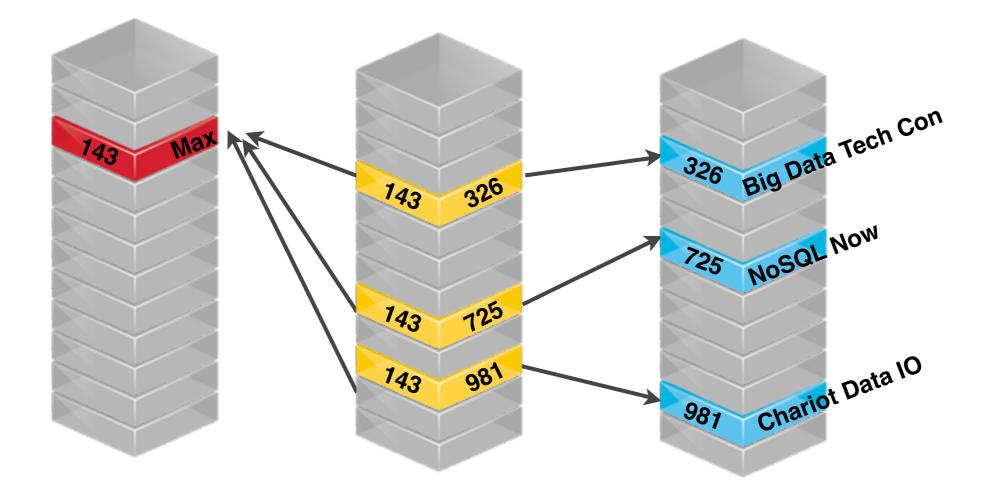


### The Problem



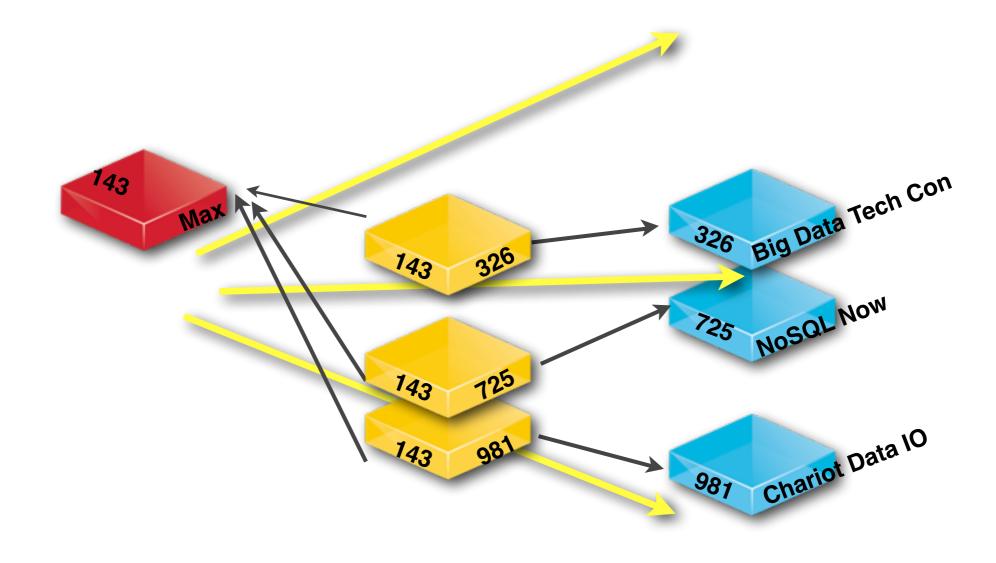
- all JOINs are executed every time you query (traverse) the relationship
- executing a JOIN means to search for a key in another table
- with Indices executing a JOIN means to lookup a key
- B-Tree Index: O(log(n))
- more entries => more lookups => slower JOINs





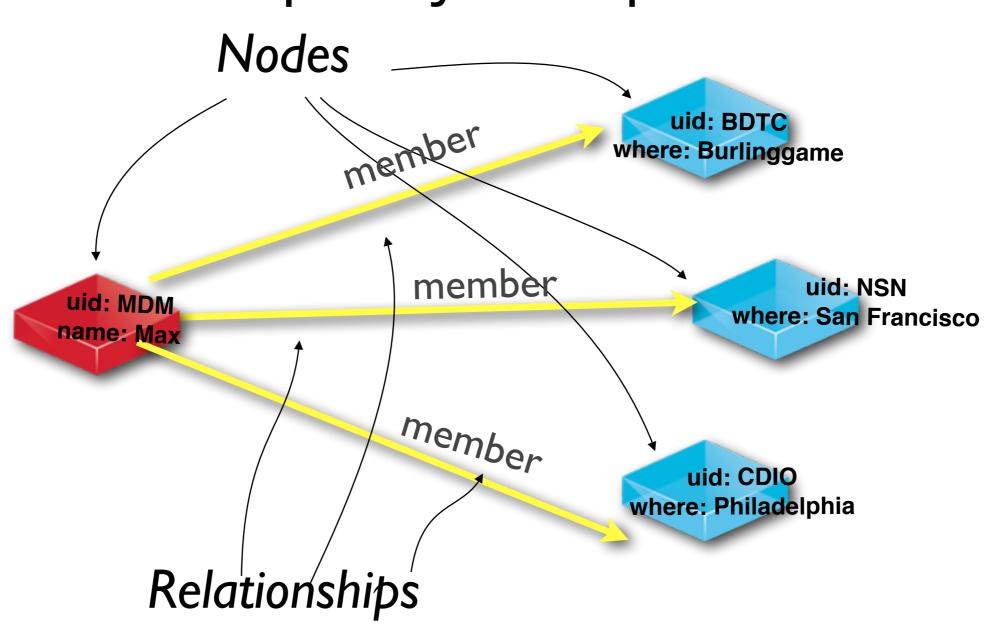
People Attend Conferences





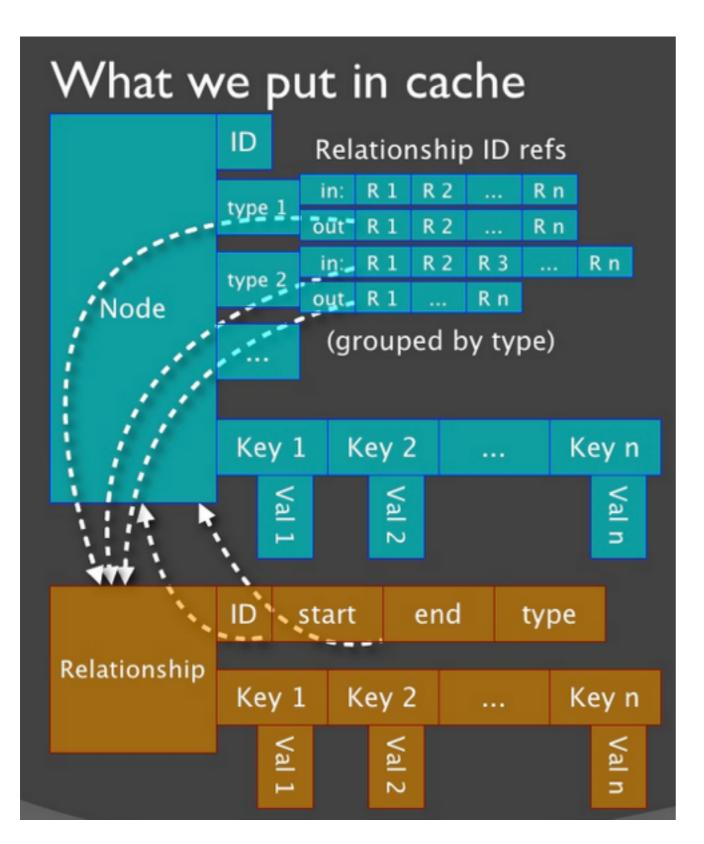


### A Property Graph





### The Neo4j Secret Sauce



- Pointers instead of look-ups
- Do all your "Joining" on creation
- Spin spin spin through this data structure

## Graph Buzz!



MacArthur 'Genius Grant' Winner Maria Chudnovsky on Graph Theory

Wednesday, October 03, 2012



Bright Launches Bright Packed With Jobs Data Seeking Tips





The First Word on Tech INFOWORLD TECH WATCH

AUGUST 29, 2012

#### Buzz grows around graph databases

Interest in graph databases will continue to grow, given its ability to analyze data delivered in a non-relational format, such as social networking data

By Paul Krill | InfoWorld

September 26, 2012 by Emil Eifrem

💟 Follow @pjkrill

🖧 Like

969

Tweet

48

173

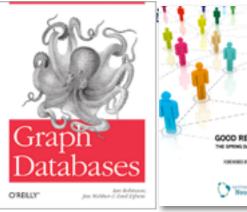
in Share

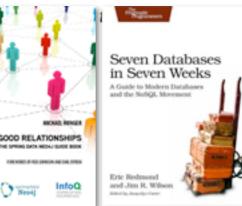
g +1

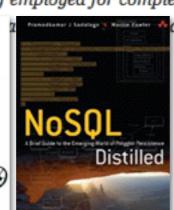
#### Graph Databases: The New Way to Access Super Fast Social Data 259

Facebook's Social Graph, Neo4j show rising use of graph databases

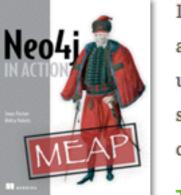
**Summary:** Facebook's Social Graph -- the database underlying its Graph Search engine unveiled yesterday-- is just one of many graph databases being employed for complex, connected data. Neo4j











I saw my own Interest Graph and it's scaryaccurate. We'd certainly pay for the ability to use the Gravity personalization technology I saw today at TechCrunch to help target content to users.

TC Michael Arrington, TechCrunch



## The Neo4j Graph Database

- Neo4j is the leading graph database in the world today
  - Most widely deployed: 500,000+ downloads
  - Largest ecosystem: active forums, code contributions, etc
  - Most mature product: in development since 2000, in 24/7 production since 2003

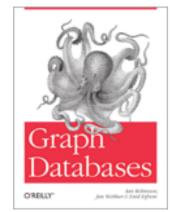
neo4j leaves me speechless. Good job at \* building the best graph database in the world!

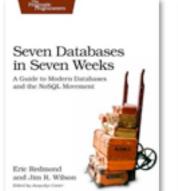


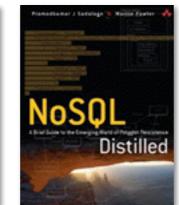
If #Cassandra rules its league (#distributed #decentralized #ColumnFamily) then #Neo4j does it in its own, which is #GraphDB. #noSQL

h Reply 13 Retweeted (Undo)

alisohani



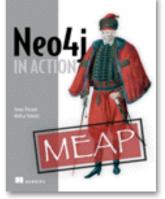








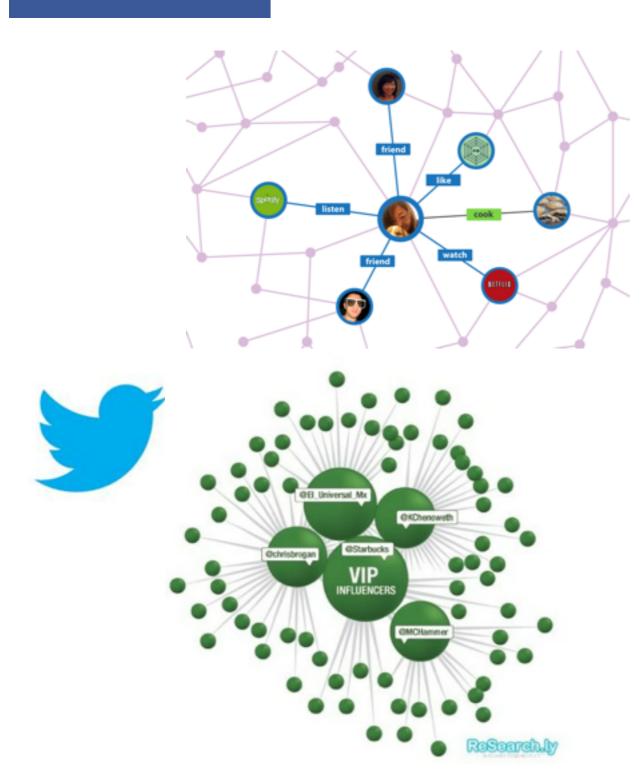




### Early Adopters of Graph Tech



facebook.





### The Knowledge Graph

Learn more about one of the key breakthroughs behind the future of search.

### Survival of the Fittest



#### Evolution of Web Search

### Pre-1999 WWW Indexing

Index

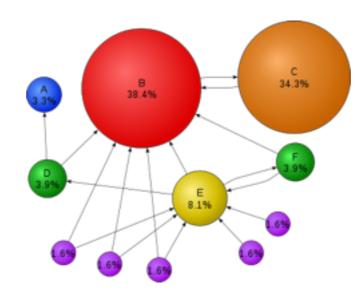
Page numbers in **bold face** refer to key term definitions Page numbers in *thalics* offse to images or diagrams Page numbers followed by a "t" indicate a table

absolute temperature scale, 350-331 sheekste zwo, 351 acceleration of gravity, A 23t accuracy, A.5 aretic acid (CHyCOOH) buffers, 575-576, 581-582 conjugate acid-base pairs, 540 ionization constant, 553, 554; manufacture of, 451 titrations, 590-592 as weak usid, 1440, 145, 551-552 arid bare pairs, conjugate, \$40-544 arid bare martinus, \$38 autoionization of water, 545-547 gus-freming exchange, 150–151 net ionic equations for, 140–150 neutralization, 146–150, 561–566 of sults, 146–151, 561–566 acid-bare titrations, 527-594 acidic solutions, 546, 650-632, A 328-A 338 acid ionization constant expressions, 550-551acid ionization constants  $(K_{a})$ . Sor ionization co aride (Kg) arideeis, 576 eride, 143. See also arid-base martinese, ionizatio constants, acids (Kg), specific entries, e.g. curbosolic acids, lartie arida Branated Lowey concept, 532-544 buffer solutions, 575-536 ronjagate acid-base pairs, 540-344 equilibrium constants, 405t ionization ronstants, 350-341, A 201 Lawis, 566-562 organic, 544 pH scale, 547-550 properties, 143-145 solubility of salas, 397-598 solutiona, 546, 650-652, A.326-A.338 strengthe, 143-146, 353-556 titrationa, 517-394 water's role, 540 actinides, 55, 230-251 activated complex, 433 activation energies (Eq), 434, 430–640, 443, 647, 449–450 active sites, 449 activities, 466, 547, 703-706 activity point, metal, 159-160 actual yields, 120 addition, 43, A.6, A.9 ublition, significant figures in, A.6

ui: 342-340, 366-370, 380-301, 706 sklyl groups, 70-71 sirohols, 64, 503-507 aldebydes, 278-279 alkali metala, 55, 106 alkaline batteries, 670 ultuline eath metals, 55 ultuline fael cells, 674 alkalosis, 576 silveres, 68-71, 277-278, A.25-A.36 silveres, 280-283, A.26-A.27 sikyi gova, 70-71, A.25-A.26 sikyass, 281, A.27 siloteopes, 23-24, 201, 403-405 ájóha particles, 32–39, **693–**696, 697, 699–700 ájóha radiation, **693** alpha rays, 36-37 slaminum (AJ), 7, 21, 103, 634-635, 682 animes, 544-645 ammonia (NRA) anines, 545 Bestated-Lowey base, 738-539 complex iona, 567 ionization constant, 554, 561 standard molar esthalpy of formation, 2101 structure, 6-R structure, 6+8 synthesis, 107–108, 462, 494–495 V3EPR model, 312–313 ar weak bare, 145, 539 anmonium ions, 74, 77 anophous solids, 396 umpheres (A), 658 unphipentic species, 540 unphotesic metal hydrosides, 567, 602 unpätude, 223 unu (utomic mass unit), 46 analytical chemistry, 114 Anderson, Carl, 696 ungolar geometries, 314 unions, 72, 76, 78, 257-258 unodes, 37, 654. See also electroch unodic inhibition, 683-614 antihonding molecular orbitals, 298 untimetter/antielectrone, 696 aqueour equilitein. Ser also unid-bare titratione, buffers factors affecting solubility, 597-602 precipitation, 603-604 solubility product constant, 394-597 equeous solutions (eq), 100. See also buffer solutions, nobutional electrolysis, 675-678 electrolytes, 82-83, 136, 527-528, 654-655 iosir compounds, 136-139 molarity, 166-162 standard reduction potentials in, A.32t-A.34t stainhiometric relationships in, 167

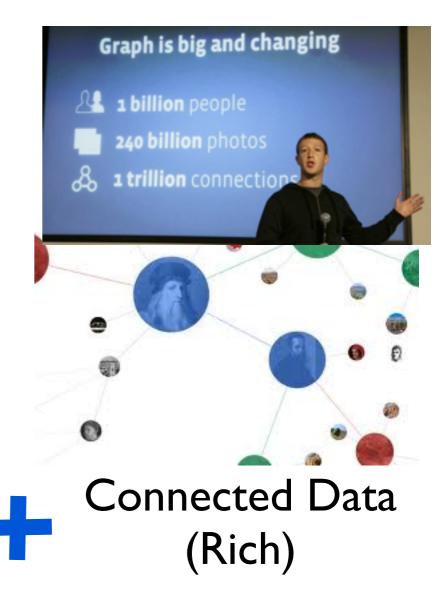
### <u> 1999 - 2012</u>

Google Invents PageRank



<u>2012-?</u>

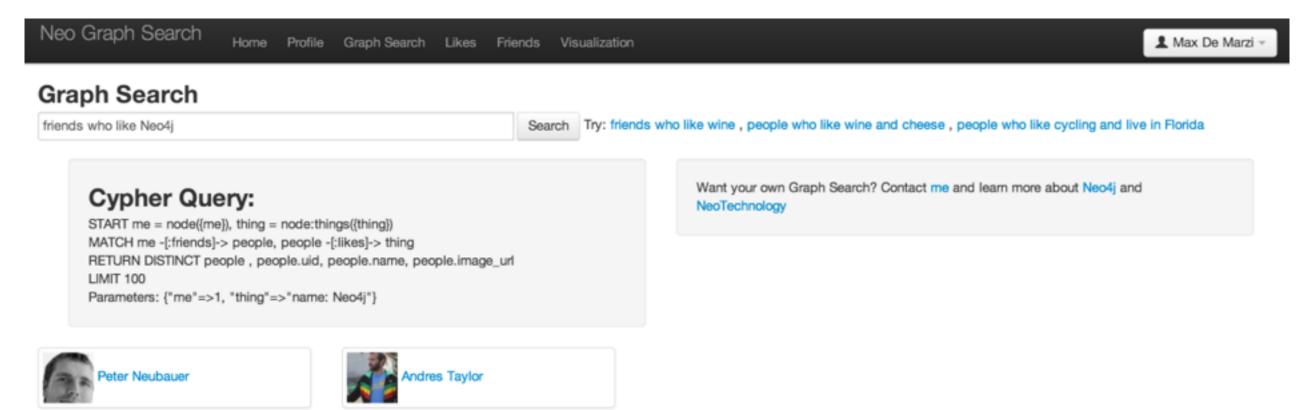
Google Knowledge Graph, Facebook Graph Search



Discrete Data 🗕

Connected Data (Simple)





### http://maxdemarzi.com/2013/01/28/facebook-graphsearch-with-cypher-and-neo4j/

### Survival of the Fittest Evolution of Online Recruiting





#### Home > Job Search

Search over **150,000** U.S. jobs. Perform your search below or get tips on searching.

#### Location Search:

------ Select a location -----Alaska-Anchorage Alaska-Fairbanks Alaska-Juneau Alaska-Valdez Alabama-Anniston

#### Category Search:

----- Select a category -----Accounting/Finance/Banking Administrative/Clerical Creative Arts/Media Education/Training Engineering/Architecture/Design

#### Keyword Search:



#### Discrete Data





Invite more friends - ask them to share their connections

#### Jobs with Connections

Sr. Statistical Analyst, Product Innovation Netflix – Los Gatos, CA From: Job.com – 1 days ago Creative Director frog design — San Francisco, CA From: Experteer – 8 days ago

Java Server Software Engineer Electronic Arts – Redwood City, CA From: Experteer – 3 days ago Want better jobs? Tell us your current job title

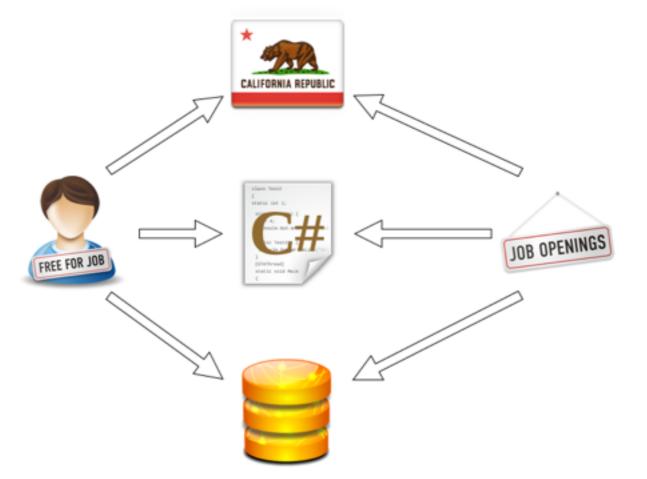
EMERGENCY MEDICAL TECH U. S. NAVY - Fremont, CA

U.S.NAVY - Fremont, CA From: Monster - 14 days ago



#### Connected Data





http://maxdemarzi.com/2012/10/18/matches-are-thenew-hotness/



```
START me=node:users_index(name={user})
MATCH skills<-[:has]-me-[:lives_in]->city<-[:in_location]-job-[:requires]->requirements
WHERE me-[:has]->()<-[:requires]-job
WITH DISTINCT city.name AS city_name,
        job.name AS job_name,
        LENGTH(me-[:has]->()<-[:requires]-job) AS matching_skills,
        LENGTH(job-[:requires]->()) AS job_requires,
        COLLECT(DISTINCT requirements.name) AS req_names,
        COLLECT(DISTINCT skills.name) AS skill_names
RETURN city_name, job_name,
FILTER(name IN req_names WHERE NOT name IN skill_names) AS missing
ORDER BY matching_skills / job_requires DESC, job_requires
LIMIT 10
```

#### http://maxdemarzi.com/2012/10/18/matches-are-thenew-hotness/



#### **Matching Jobs for Daniel**

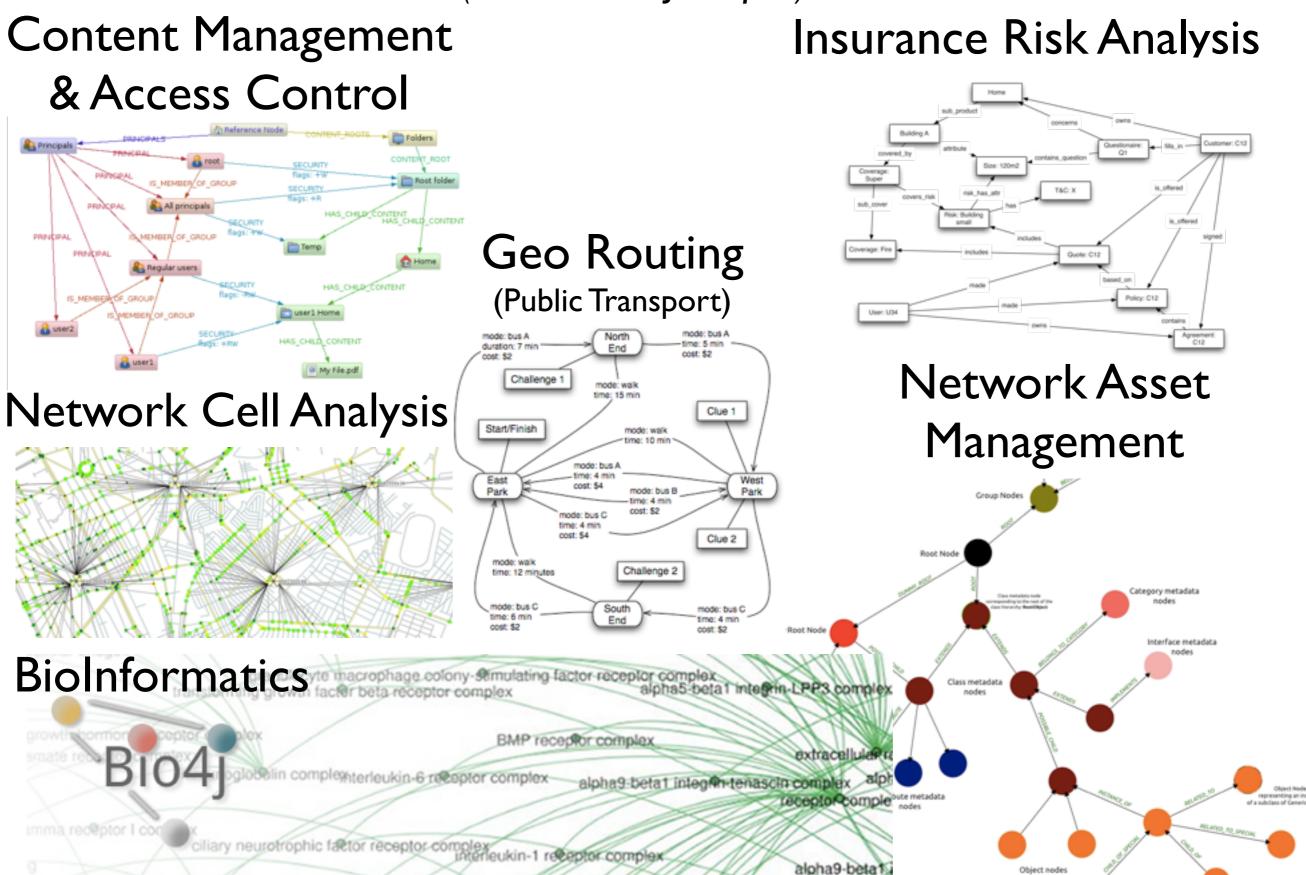
Reload the page or click Match Another User

Job	Match	Missing
Spring-C-Django-SQL in Dallas	100% Match	
Neo4j-C-CSS in Dallas	Partial Match	+ CSS
C-Spring-Java in Dallas	Partial Match	+ Java
Java-Redis-C-Neo4j in Dallas	Partial Match	+ Java + Redis

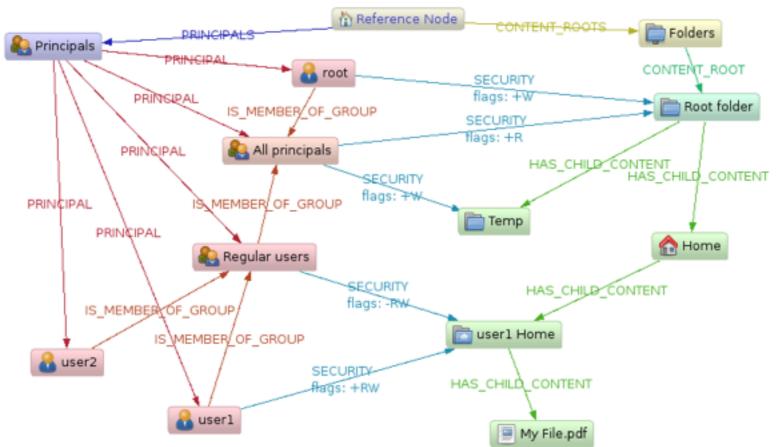
#### http://maxdemarzi.com/2012/10/18/matches-are-thenew-hotness/

### Emergent Graph in Other Industries

(Actual Neo4j Graphs)





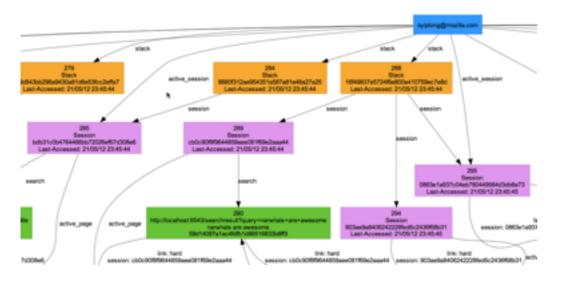


### http://maxdemarzi.com/2013/03/18/permissionresolution-with-neo4j-part-1/

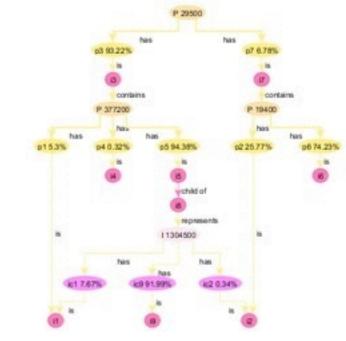
### Emergent Graph in Other Industries

(Actual Neo4j Graphs)

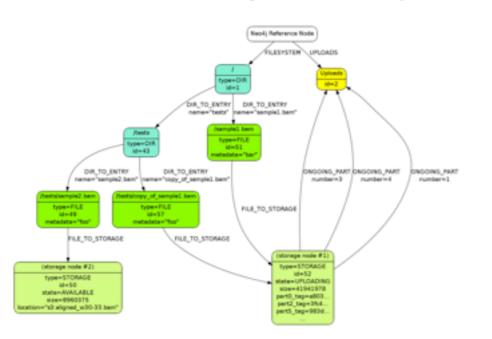
#### Web Browsing



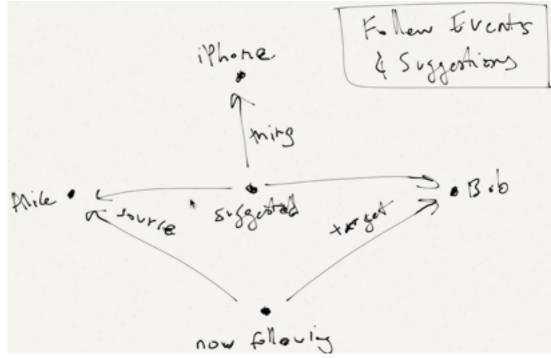
#### **Portfolio Analytics**



#### Gene Sequencing



### Mobile Social Application





Neo Love Matching Match Another User

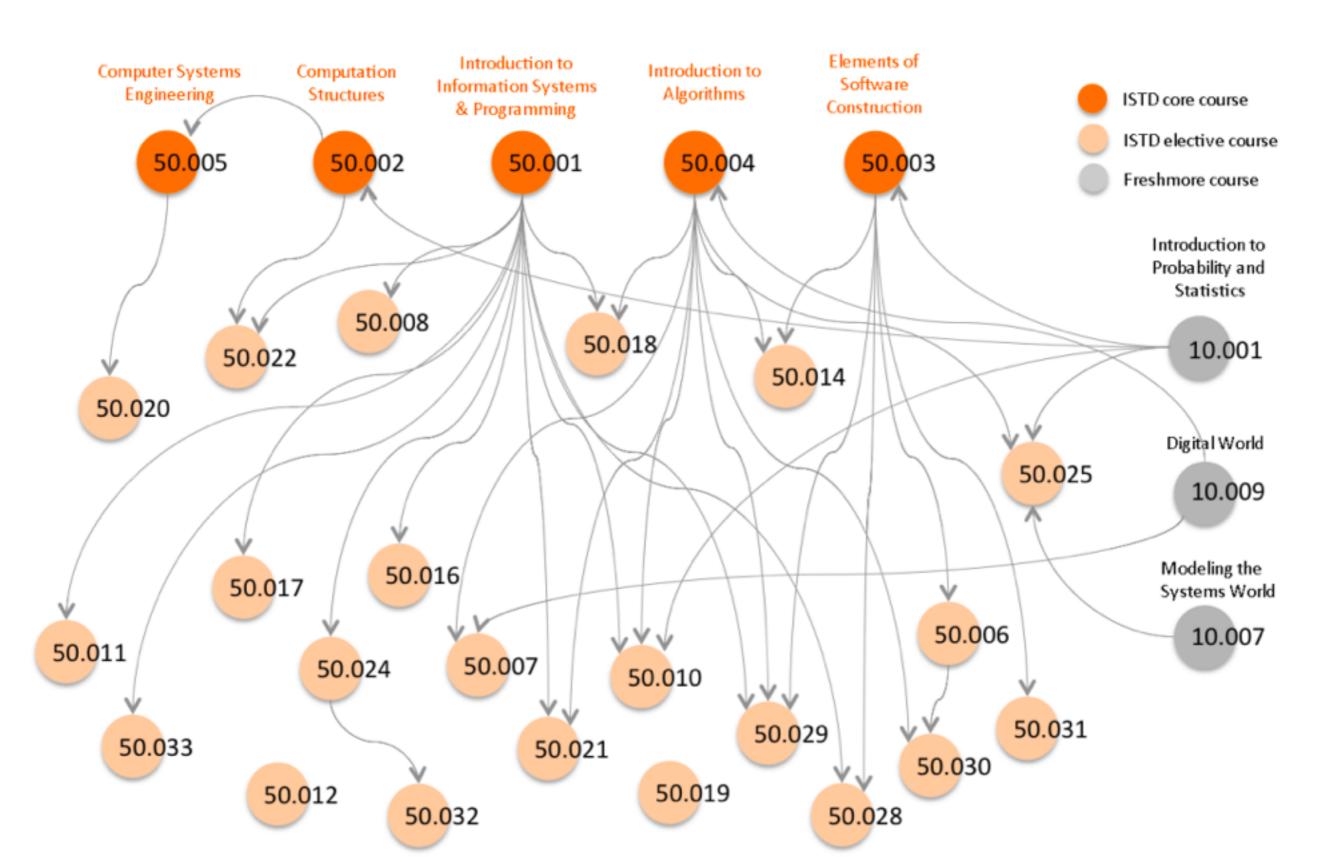
#### Matching Loves for Luvenia

Reload the page or click Match Another User Their Matches Love Compatibility Your Matches Chet in Philadelphia 2.0 Likeable Humorous Thoughtful Mckinley in Philadelphia 1.0 Serious Creative Gene in Philadelphia 1.0 Likeable Serious Keith in Philadelphia 1.0 🗸 Quiet Inventive

#### http://maxdemarzi.com/2013/04/19/match-making-withneo4j/

### Curriculum Graph





# Early Adopter Segments (What we expected to happen - view from several years ago)

Core Industries & Use Cases:	Web / ISV	Finance & Insurance	Datacom / Telecom
Network & Data Center Management			
MDM			
Social			
Geo			

### Neo4j Adoption Snapshot



Select Commercial Customers (Community Users Not Included)

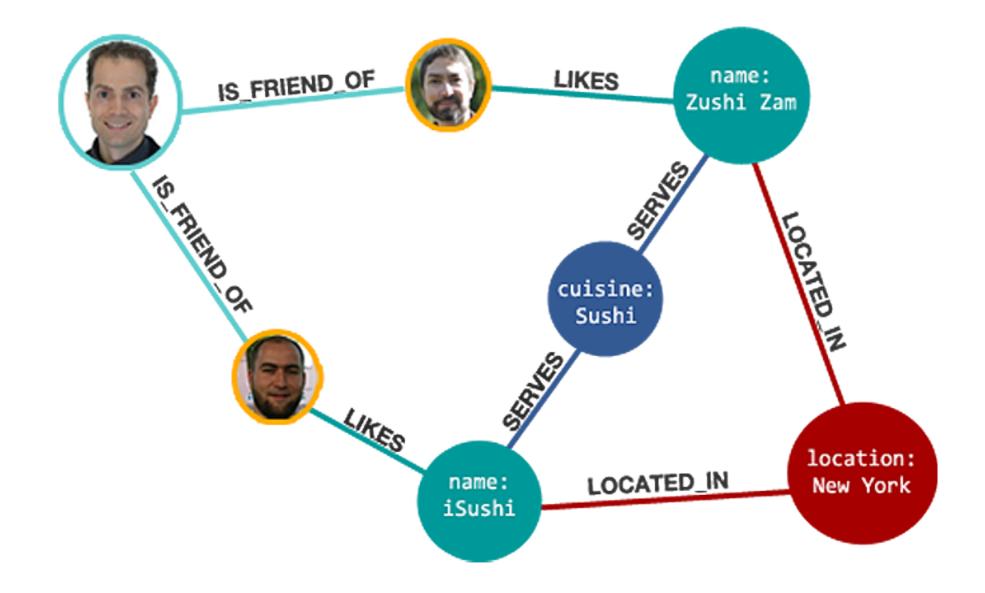




# What Can You Do With Graphs?











```
MATCH (me:Person) - [:IS_FRIEND_OF] -> (friend),
 (friend) - [:LIKES] -> (restaurant),
  (restaurant) - [:LOCATED_IN] -> (city:Location),
  (restaurant) - [:SERVES] -> (cuisine:Cuisine)
```

WHERE me.name = 'Philip' AND city.location='New York' AND cuisine.cuisine='Sushi'

**RETURN** restaurant.name

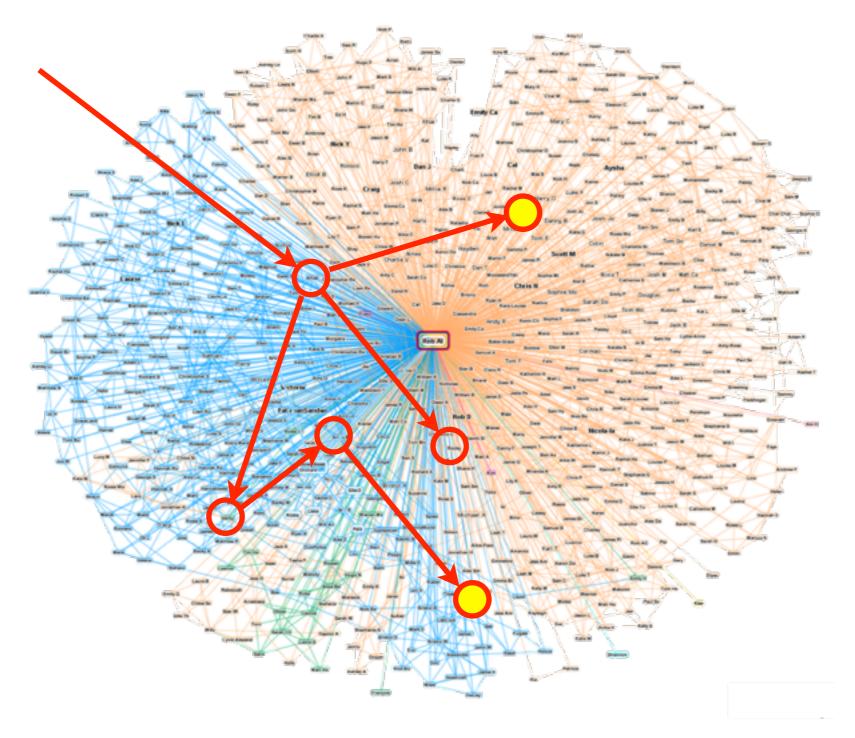
http://maxdemarzi.com/?s=facebook

\* Cypher query language example



Q

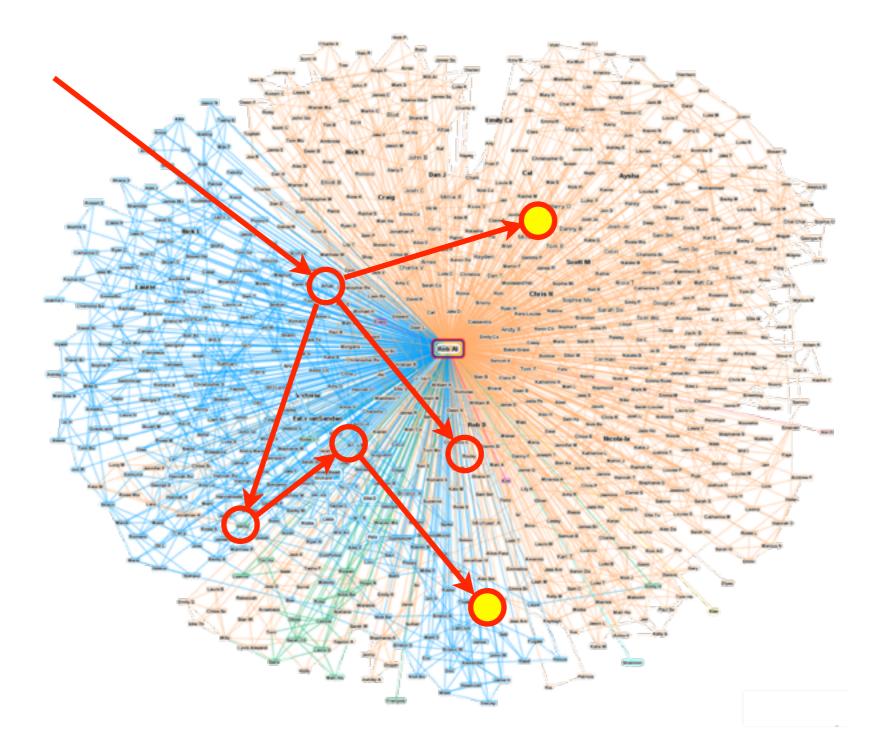




### Of course.. a graph is a graph is a graph



What drugs will bind to protein X and not interact with drug Y?





# Connected Query Performance



Query Response Time\* = f(graph density, graph size, query degree)

- Graph density (avg # rel's / node)
- Graph size (total # of nodes in the graph)
- Query degree (# of hops in one's query)

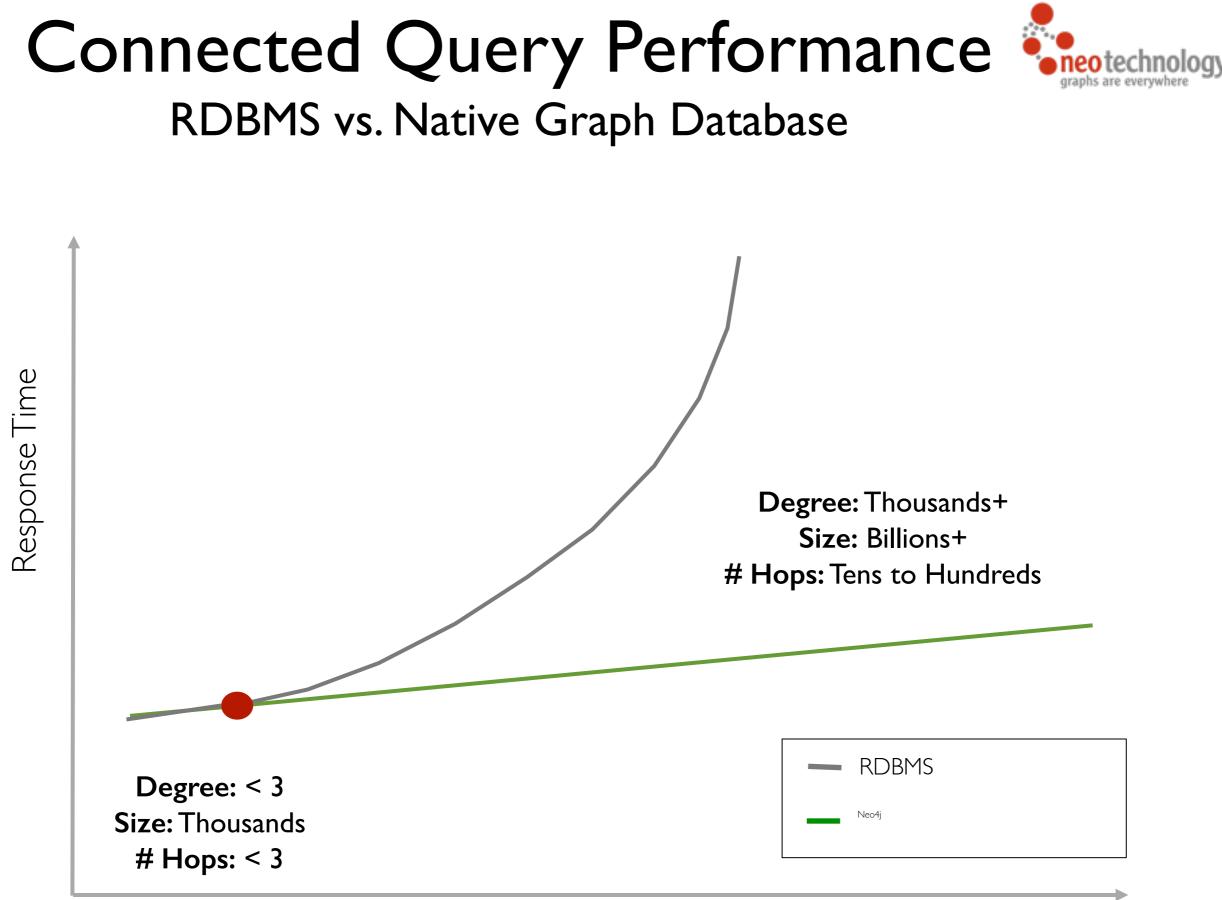
#### **RDBMS**:

>> exponential slowdown as each factor increases

#### Neo4j:

>> Performance remains constant as graph size increases

>> Performance slowdown is linear or better as density & degree increase



Connectedness of Data Set

# Graph db performance



• a sample social graph

- with ~1,000 persons
- average 50 friends per person
- pathExists(a,b) limited to depth 4
- caches warmed up to eliminate disk I/O

Database	# persons	query time
MySQL		
Neo4j		
Neo4j		

# Graph db performance



• a sample social graph

- with ~1,000 persons
- average 50 friends per person
- pathExists(a,b) limited to depth 4
- caches warmed up to eliminate disk I/O

Database	# persons	query time
MySQL	I,000	2,000 ms
Neo4j		
Neo4j		

# Graph db performance



• a sample social graph

- with ~1,000 persons
- average 50 friends per person
- pathExists(a,b) limited to depth 4
- caches warmed up to eliminate disk I/O

Database	# persons	query time
MySQL	I,000	2,000 ms
Neo4j	I,000	2 ms
Neo4j		

# Graph db performance



• a sample social graph

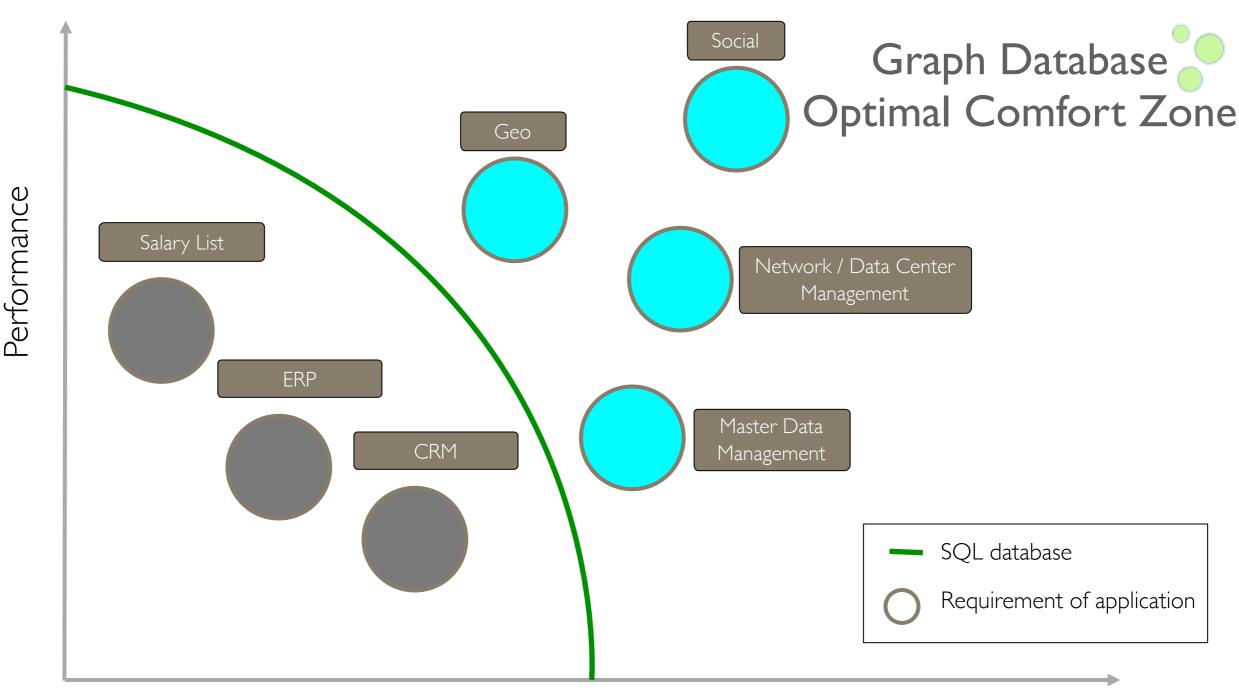
- with ~1,000 persons
- average 50 friends per person
- pathExists(a,b) limited to depth 4
- caches warmed up to eliminate disk I/O

Database	# persons	query time
MySQL	I,000	2,000 ms
Neo4j	I,000	2 ms
Neo4j	I,000,000	2 ms

\*Additional Third Party Benchmark Available in Neo4j in Action: http://www.manning.com/partner/

### The Zone of SQL Adequacy





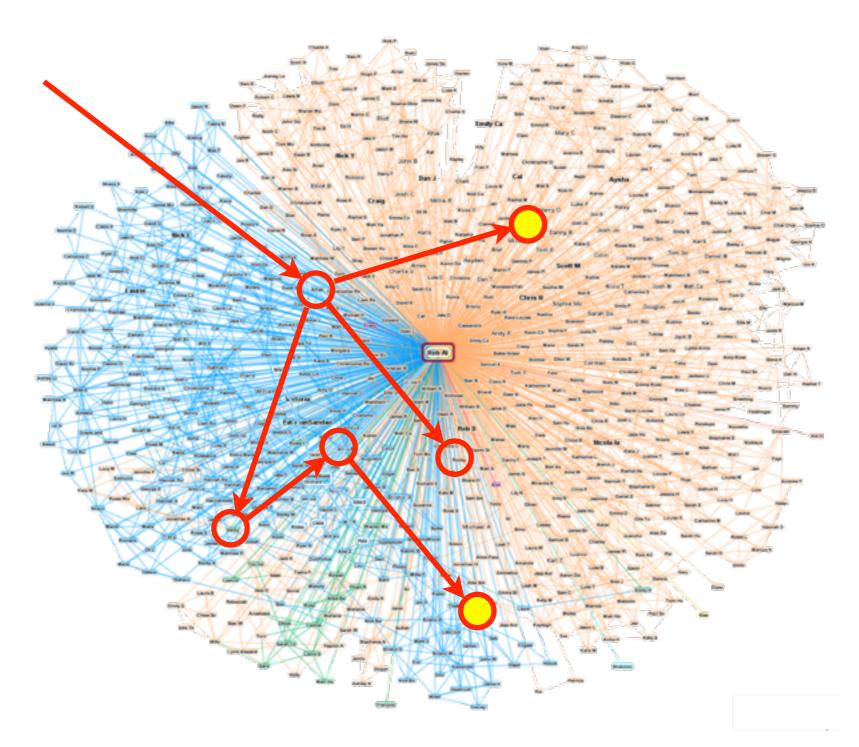
Connectedness of Data Set



# Graph Technology Ecosystem

### #I: Graph Local Queries





e.g. Recommendations, Friend-of-Friend, Shortest Path

#### What is a Graph Database



"A graph database... is an online database management system with CRUD methods that expose a graph data model"

- Two important properties:
  - Native graph storage engine: written from the ground up to manage graph data
  - Native graph processing, including index-free adjacency to facilitate traversals

1] Robinson, Webber, Eifrem. Graph Databases. O'Reilly, 2013. p. 5. ISBN-10: 1449356265

### Graph Databases are Designed to:



- I. Store inter-connected data
- 2. Make it easy to make sense of that data
- 3. Enable extreme-performance operations for:
  - Discovery of connected data patterns
  - Relatedness queries > depth I
  - Relatedness queries of arbitrary length
- 4. Make it easy to evolve the database

Top Reasons People Use Graph Databases



- I.Problems with Join performance.
- 2.Continuously evolving data set (often involves wide and sparse tables)
- 3.The Shape of the Domain is naturally a graph
- 4. Open-ended business requirements necessitating fast, iterative development.



### ABOUT THE VIADEO GROUP

The Viadeo Group forms the world's second largest professional social network with over 50 million members.

- N°1 in France with over 7 million members,
- N°1 in China with over 14 millions members.
- Strong presence in Europe and in emerging markets, strong growth in South America
- Presence in Russia since December 2011 through a joint venture with leading media group : Sanoma Independent Media
- April 2012 : record funding of 24 million euros (FSI, historical shareholders and new entrants)
- Headquarters in Paris and local offices on 5 continents
- 400 employees



• The Viadeo Group consists of:

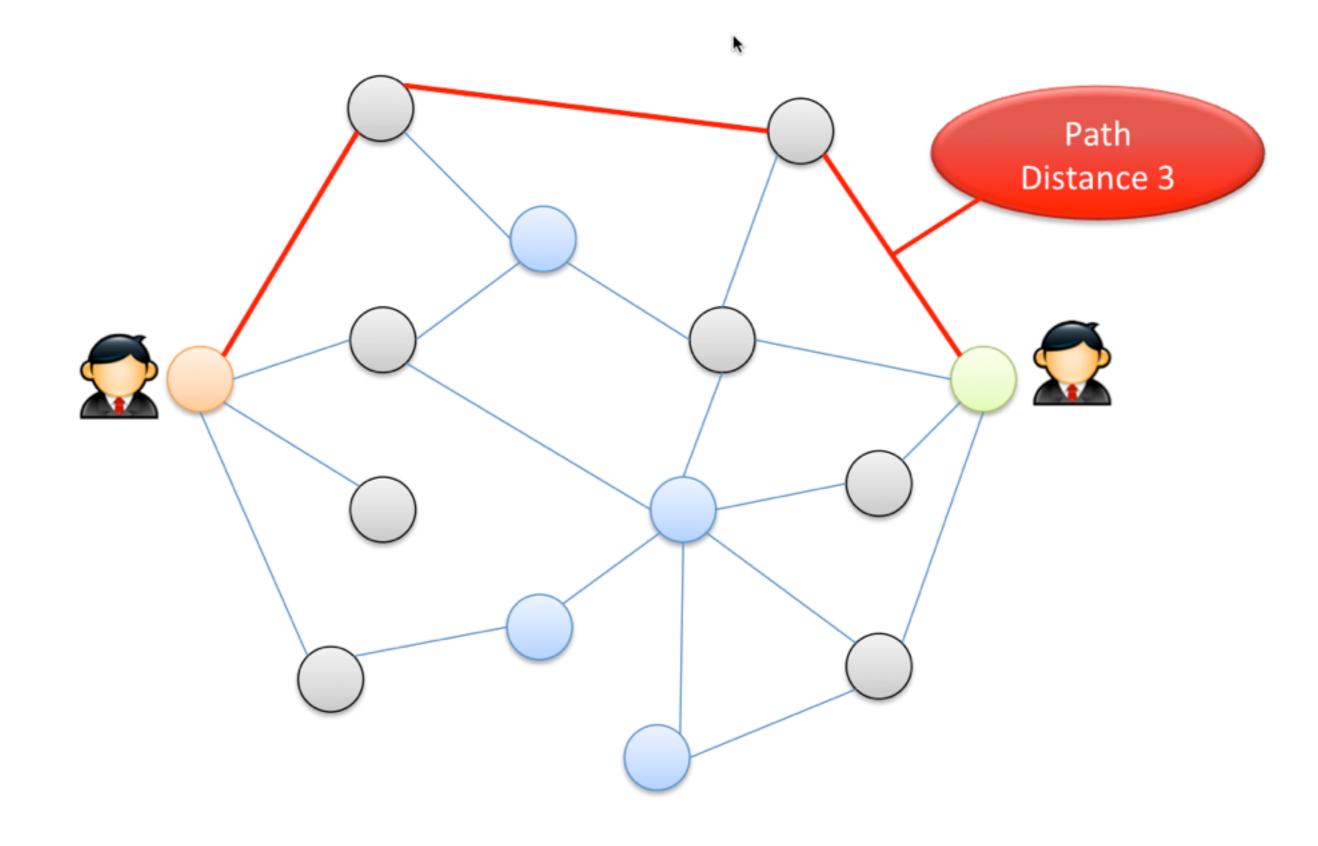
viadec

Apna**Circle.com** 











#### PREHISTORY 2006-2011

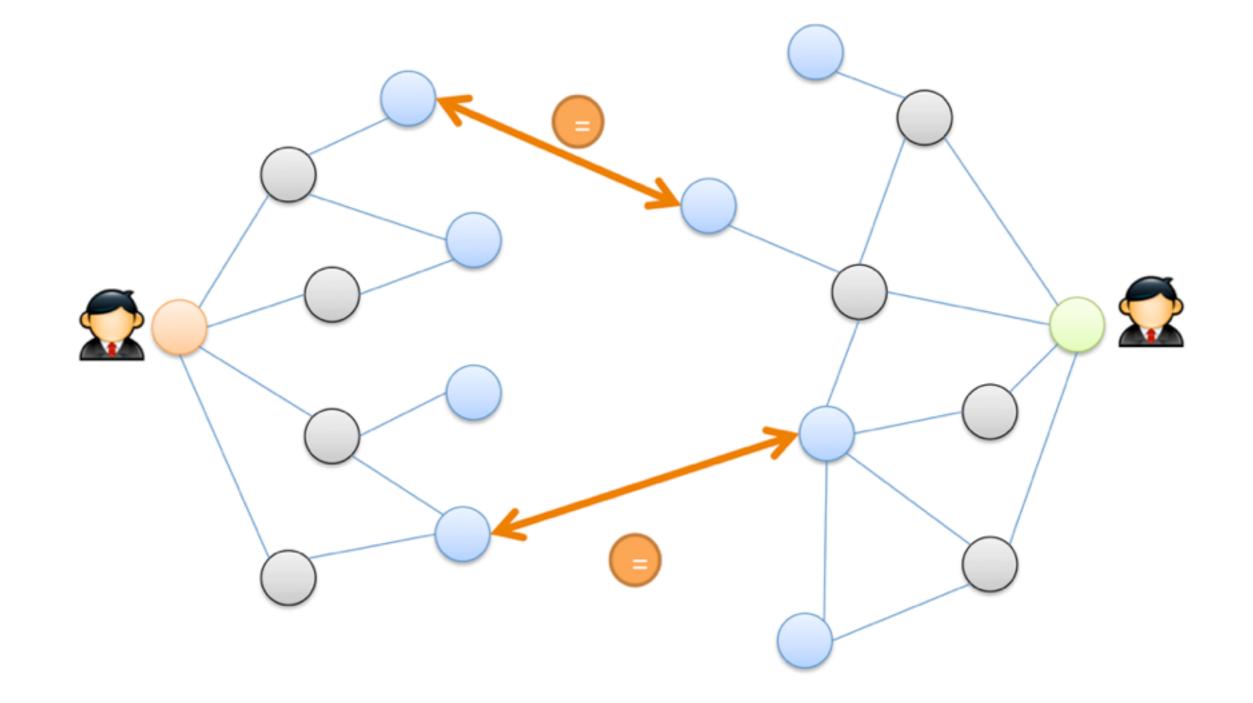
- In-house algorithm
- Network storage in MySQL Database

```
CREATE TABLE `Network` (
 `memberId` int(11) NOT NULL DEFAULT '0',
 `L1` mediumblob NOT NULL,
 `L2` mediumblob NOT NULL,
 PRIMARY KEY (`memberId`)
) ENGINE=InnoDB;
```



### PREHISTORY 2006-2011

k







#### LIMITATIONS

1) Important latency for complete update



2) Massive bandwidth impact for internal network

30%

3) 48 hours to restart from scratch



k





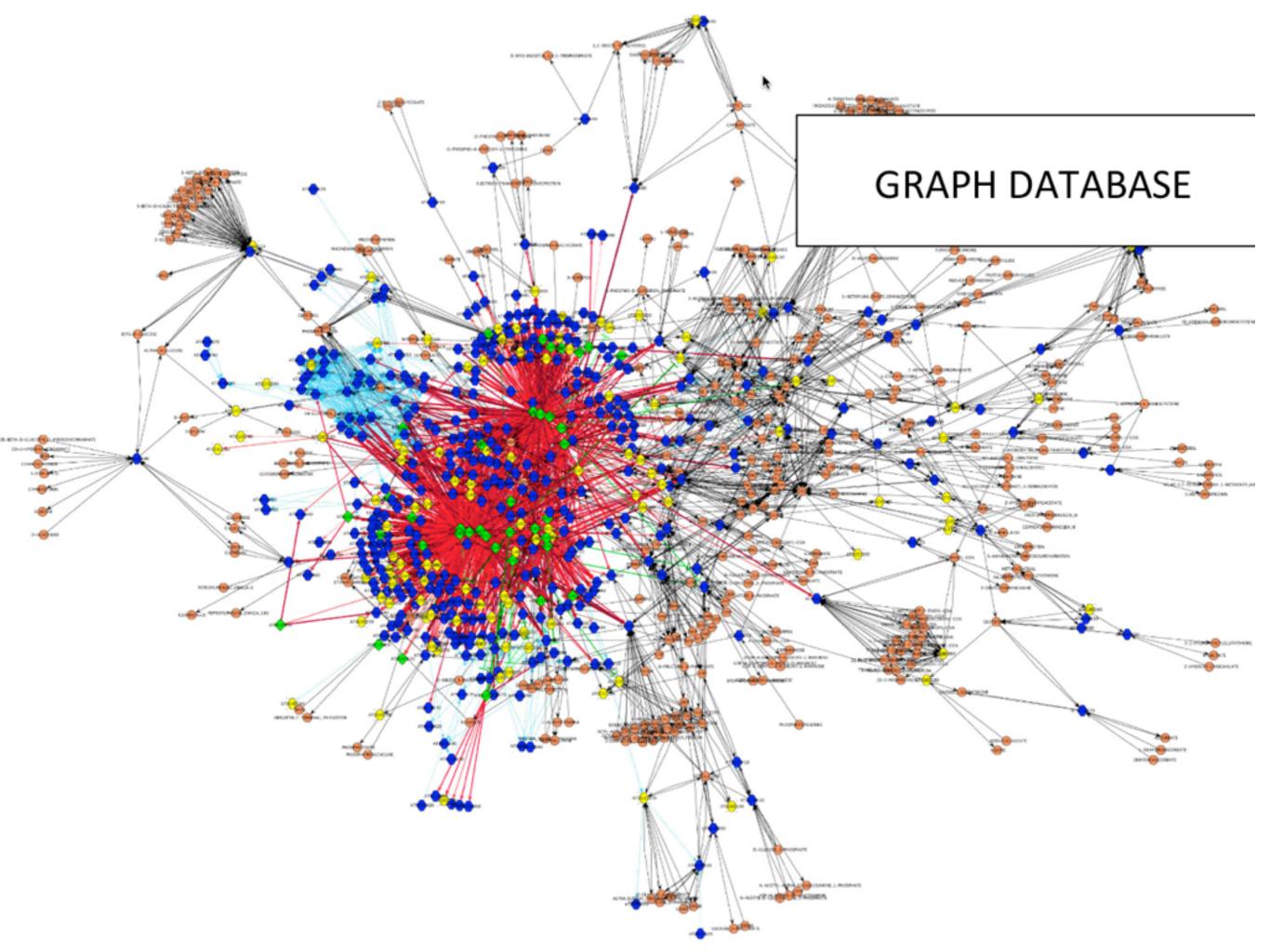
Update the network (old-fashioned style)

Member A and Member B are now in contacts

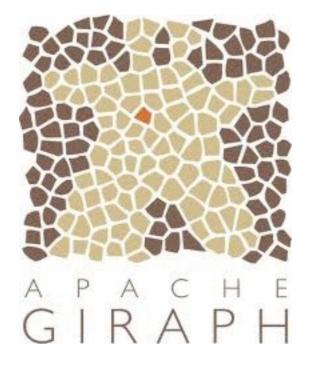
- Update of A.L1 + B.L1 and A.L2 + B.L2
- Retrieving A.L1 + B.L1 and update \*.L2

Example:

- A has 500 contacts
- B has 150 contacts
- → 500 + 150 + 2 = 652 updates!









#### Wait what?





# New Users? Real Time Updates?





#### BENEFITS



Very easy to integrate (less than 2 months)



Instantaneous graph updates

High Availability

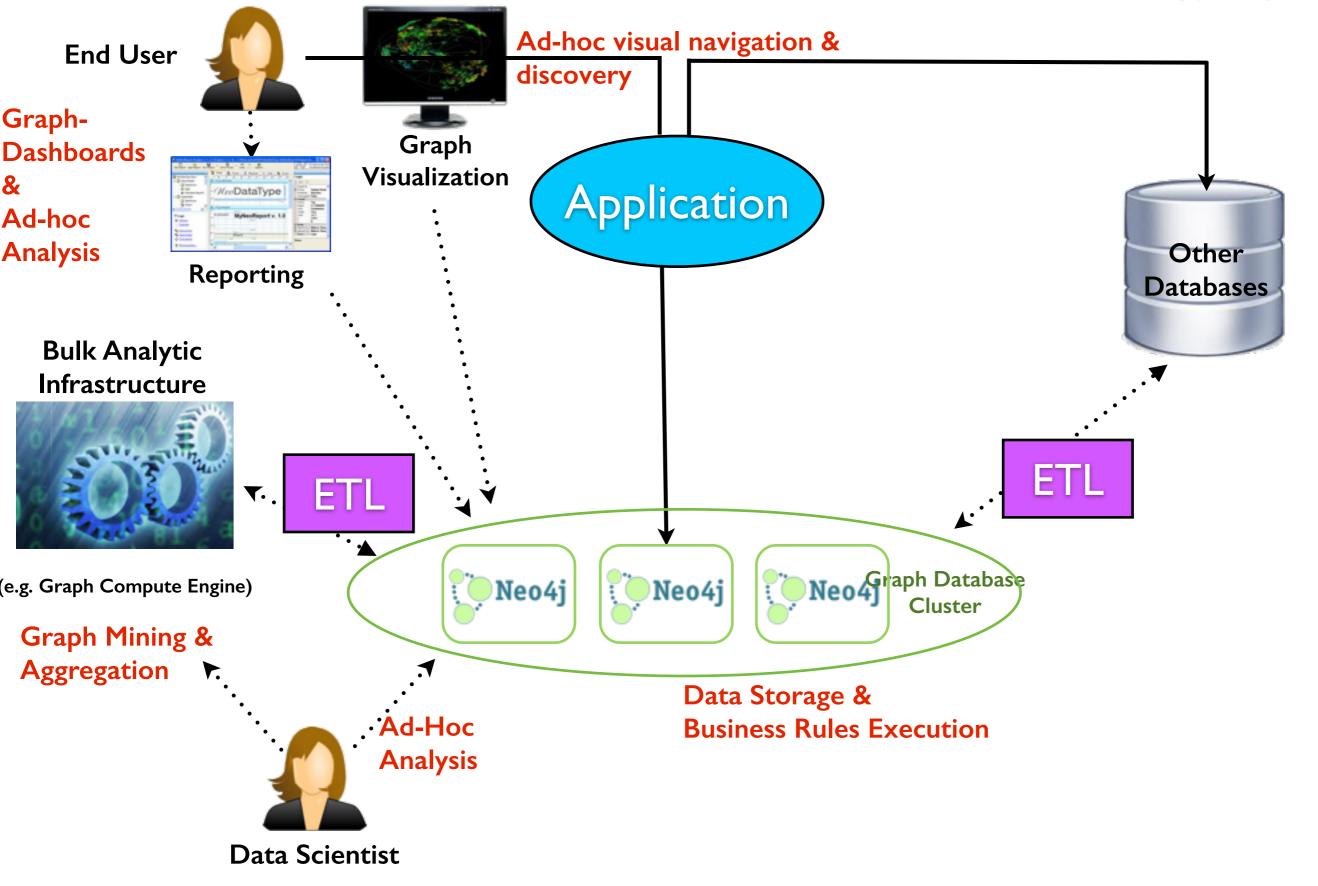




Backup / Restore

### Graph Database Deployment

o technology



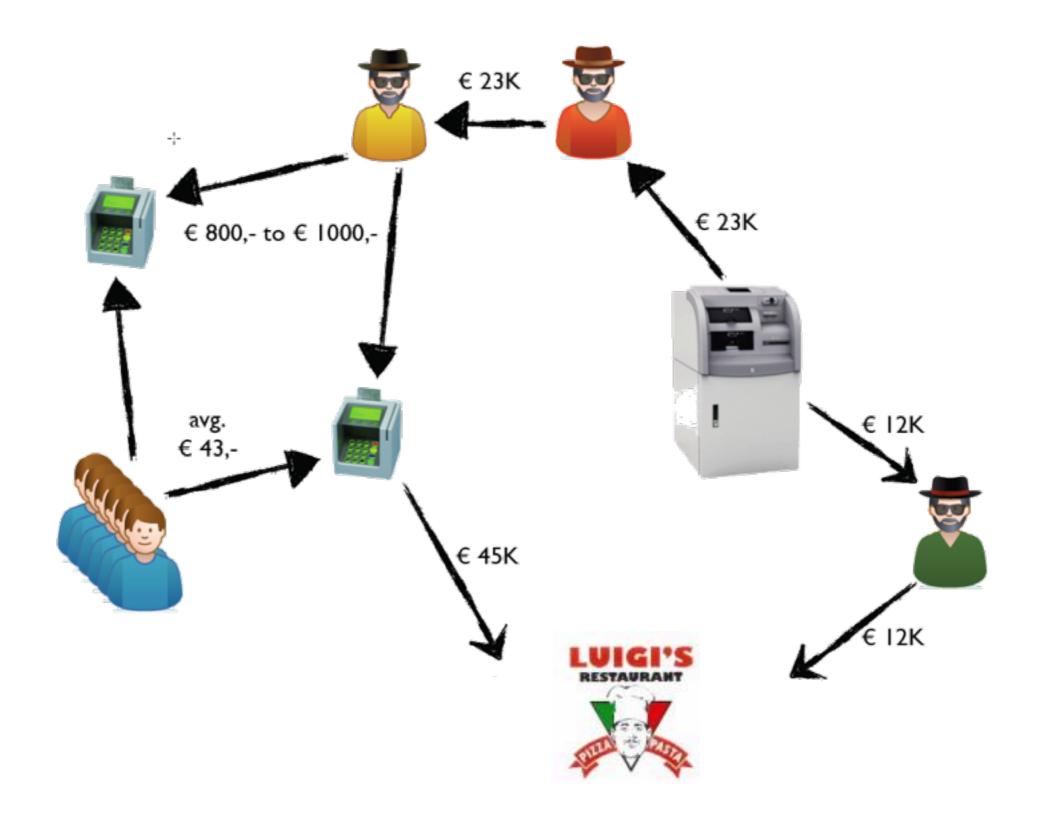


## Graph Dashboards

The Power of Visualization

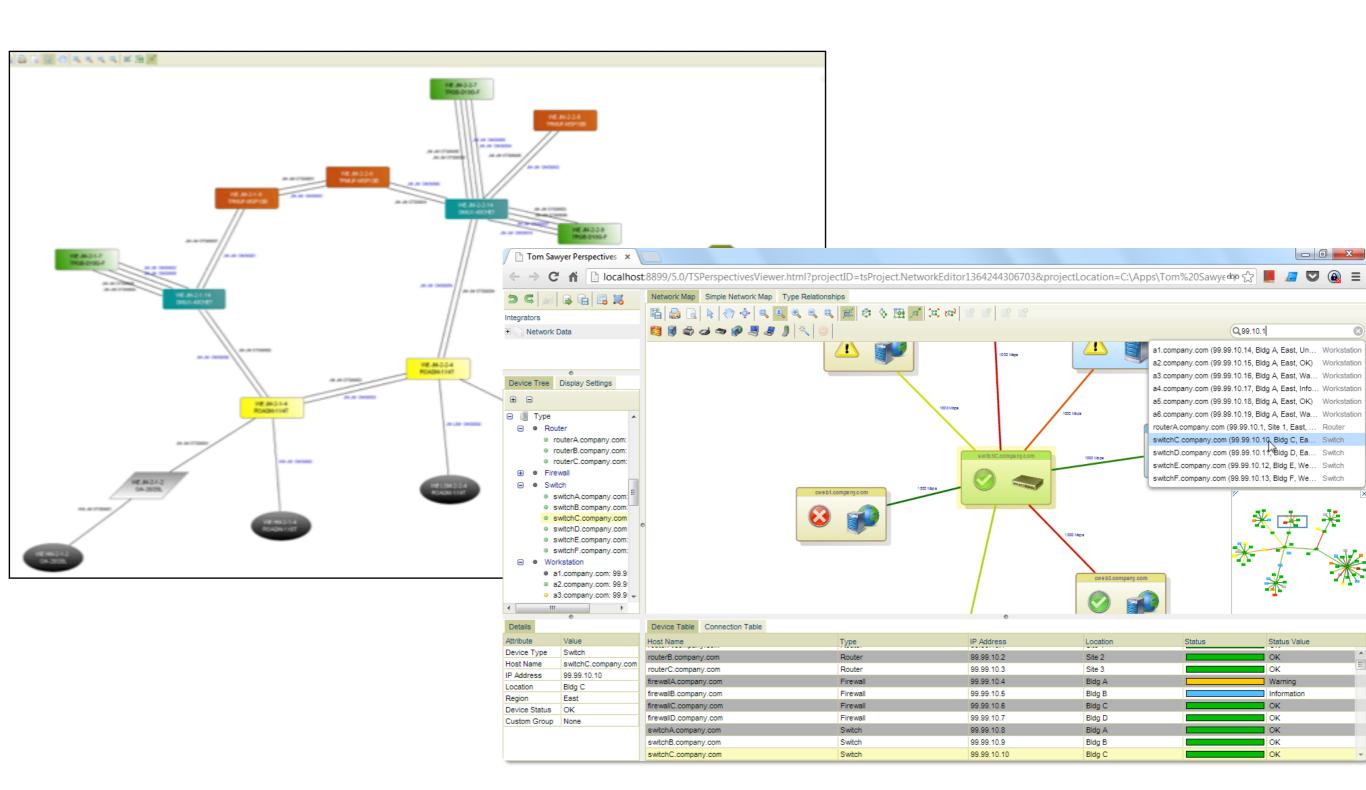
### Fraud Detection & Money Laundering

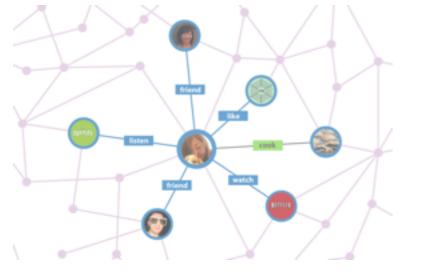




### **IT** Service Dependencies



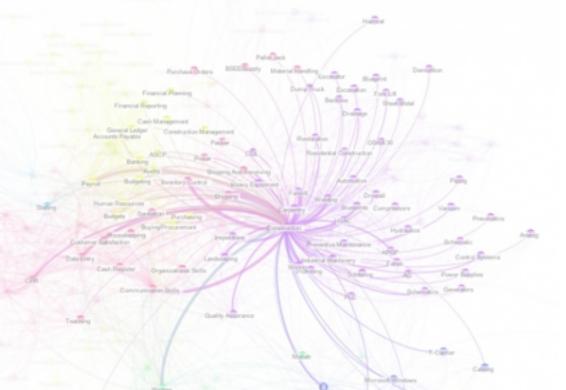


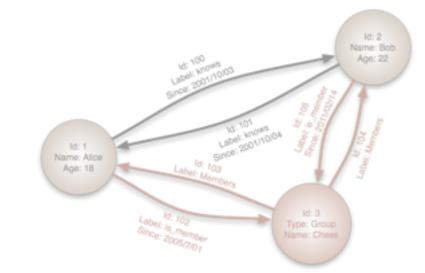




# Working with Graphs

### Case Studies & Working Examples







Cypher

#### ASCII art Graph Patterns



MATCH (A) -[:LOVES]-> (B)
WHERE A.name = "A"
RETURN B as lover



## Social Example

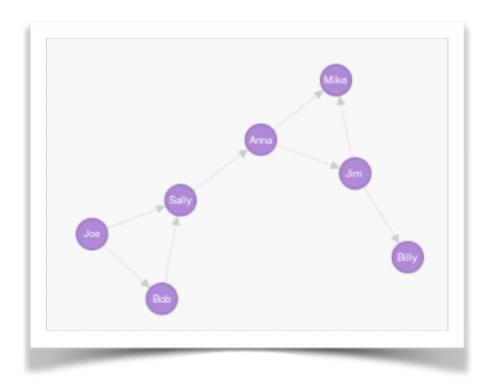
#### Practical Cypher Social Graph - Create



#### CREATE

```
(joe:Person {name:"Joe"}),
(bob:Person {name:"Bob"}),
(sally:Person {name:"Sally"}),
(anna:Person {name:"Anna"}),
(jim:Person {name:"Jim"}),
(mike:Person {name:"Mike"}),
(billy:Person {name:"Billy"}),
```

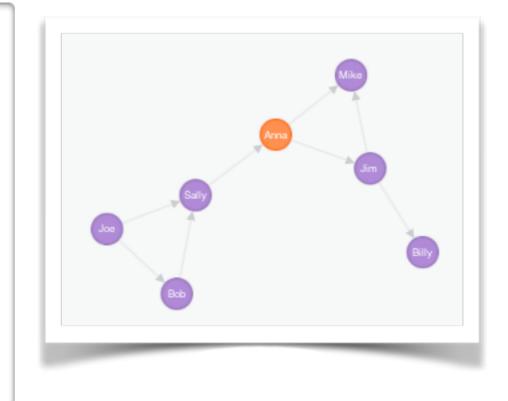
```
(joe)-[:KNOWS]->(bob),
(joe)-[:KNOWS]->(sally),
(bob)-[:KNOWS]->(sally),
(sally)-[:KNOWS]->(anna),
(anna)-[:KNOWS]->(jim),
(anna)-[:KNOWS]->(mike),
(jim)-[:KNOWS]->(mike),
(jim)-[:KNOWS]->(billy)
```





#### **Practical Cypher** Social Graph - Friends of Joe's Friends

```
MATCH (person)-[:KNOWS]-(friend),
    (friend)-[:KNOWS]-(foaf)
WHERE person.name = "Joe"
    AND NOT(person-[:KNOWS]-foaf)
RETURN foaf
```

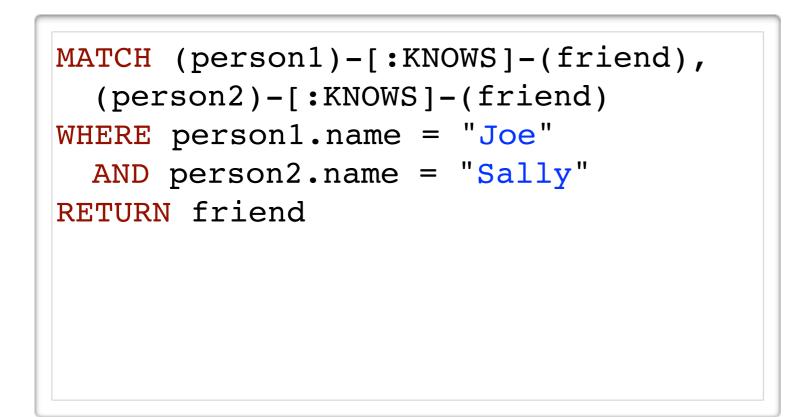


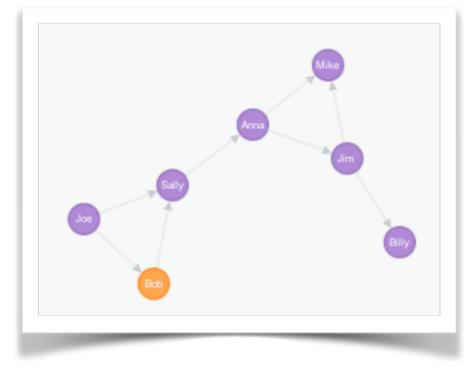
foaf

{name:"Anna"}



#### **Practical Cypher** Social Graph - Common Friends





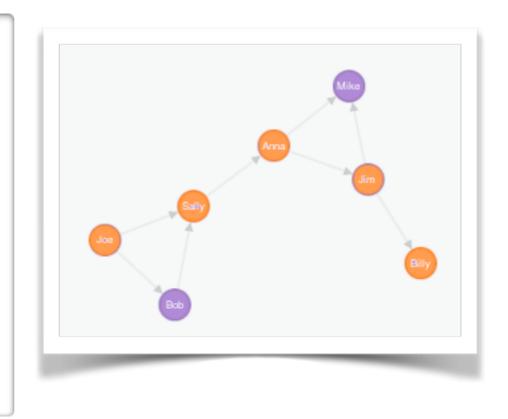
#### friend

{name:"Bob"}



#### **Practical Cypher** Social Graph - Shortest Path

```
MATCH path = shortestPath(
  (person1)-[:KNOWS*..6]-(person2)
)
WHERE person1.name = "Joe"
  AND person2.name = "Billy"
RETURN path
```



#### path

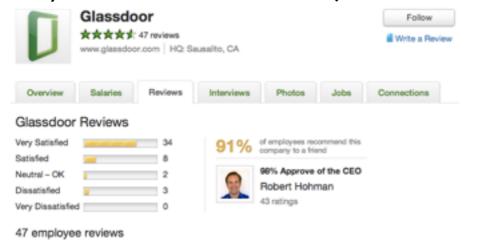
```
{start:"13759",
nodes:["13759","13757","13756","13755","13753"],
length:4,
relationships:["101407","101409","101410","101413"],
end:"13753"}
```

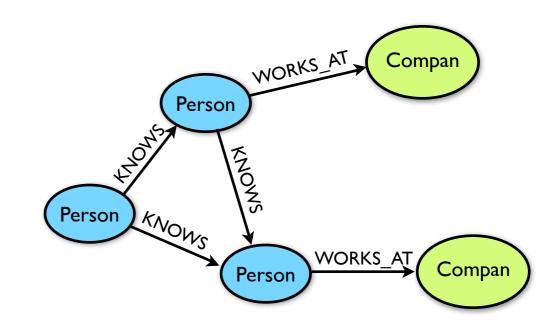
**Industry:** Online Job Search Use case: Social / Recommendations Sausalito, CA



#### Background

• Online jobs and career community, providing anonymized inside information to job seekers





#### **Business problem**

- Wanted to leverage known fact that most jobs are found through personal & professional connections
- Needed to rely on an existing source of social network data. Facebook was the ideal choice.
- End users needed to get instant gratification
- Aiming to have the best job search service, in a very competitive market

#### Solution & Benefits

- First-to-market with a product that let users find jobs through their network of Facebook friends
- Job recommendations served real-time from Neo4j
- Individual Facebook graphs imported real-time into Neo4j
- Glassdoor now stores > 50% of the entire Facebook social graph
- Neo4j cluster has grown seamlessly, with new instances being brought online as graph size and load have increased



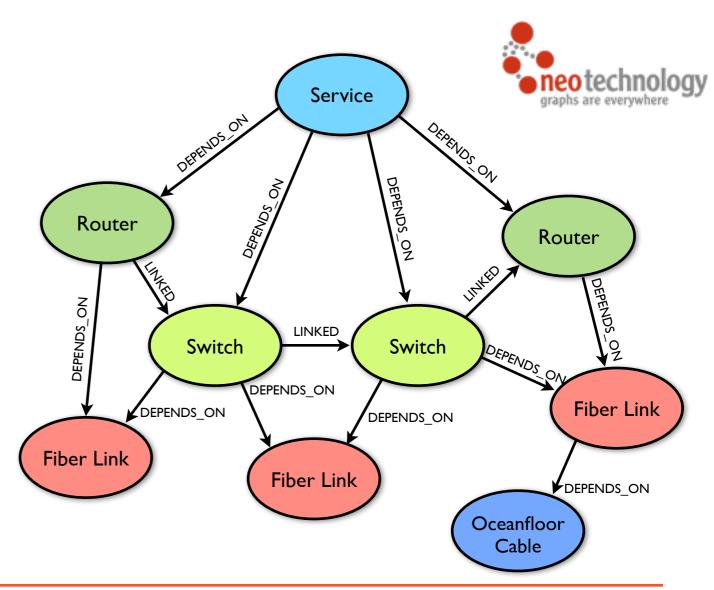
## Network Management Example



Industry: Communications Use case: Network Management Paris, France

#### Background

- Second largest communications company in France
- Part of Vivendi Group, partnering with Vodafone



#### **Business problem**

- Infrastructure maintenance took one full week to plan, because of the need to model network impacts
- Needed rapid, automated "what if" analysis to ensure resilience during unplanned network outages
- Identify weaknesses in the network to uncover the need for additional redundancy
- Network information spread across > 30 systems, with daily changes to network infrastructure
- Business needs sometimes changed very rapidly

#### Solution & Benefits

- Flexible network inventory management system, to support modeling, aggregation & troubleshooting
- Single source of truth (Neo4j) representing the entire network
- Dynamic system loads data from 30+ systems, and allows new applications to access network data
- Modeling efforts greatly reduced because of the near I:I mapping between the real world and the graph
- Flexible schema highly adaptable to changing business requirements



Industry: Web/ISV, Communications Use case: Network Management Global (U.S., France)

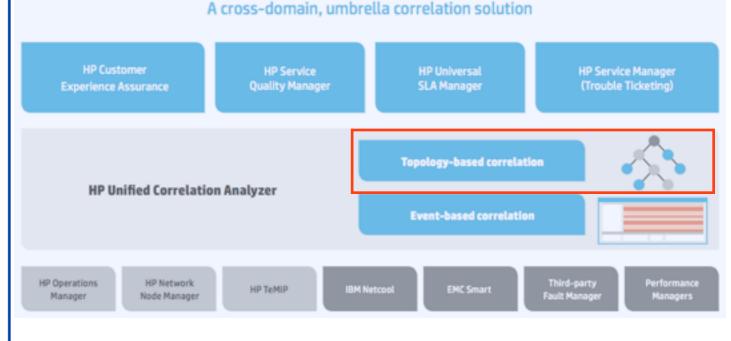


#### Background

- World's largest provider of IT infrastructure, software & services
- HP's Unified Correlation Analyzer (UCA) application is a key application inside HP's OSS Assurance portfolio
- Carrier-class resource & service management, problem determination, root cause & service impact analysis
- Helps communications operators manage large, complex and fast changing networks

#### **Business problem**

- Use network topology information to identify root problems causes on the network
- Simplify alarm handling by human operators
- Automate handling of certain types of alarms Help operators respond rapidly to network issues
- Filter/group/eliminate redundant Network Management System alarms by event correlation



#### Solution & Benefits

- Accelerated product development time
- Extremely fast querying of network topology
- Graph representation a perfect domain fit
- 24x7 carrier-grade reliability with Neo4j HA clustering
- Met objective in under 6 months

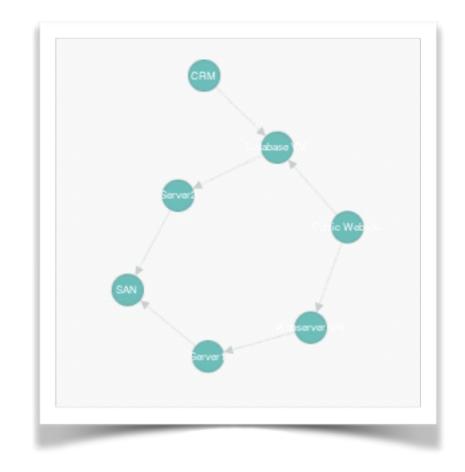


#### Practical Cypher Network Management - Create

#### CREATE

```
(crm {name:"CRM"}),
(dbvm {name:"Database VM"}),
(www {name:"Public Website"}),
(wwwvm {name:"Webserver VM"}),
(srv1 {name:"Server 1"}),
(san {name:"SAN"}),
(srv2 {name:"Server 2"}),
```

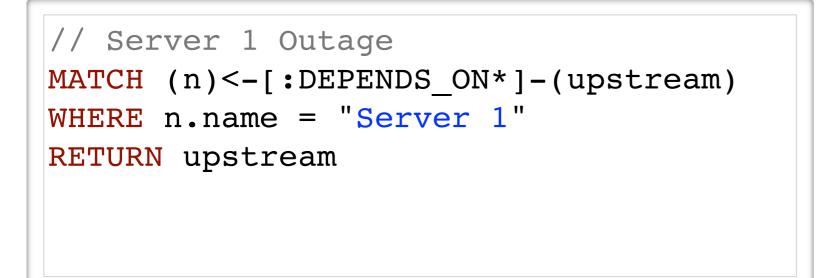
```
(crm)-[:DEPENDS_ON]->(dbvm),
(dbvm)-[:DEPENDS_ON]->(srv2),
(srv2)-[:DEPENDS_ON]->(san),
(www)-[:DEPENDS_ON]->(dbvm),
(www)-[:DEPENDS_ON]->(wwwvm),
(wwwvm)-[:DEPENDS_ON]->(srv1),
(srv1)-[:DEPENDS_ON]->(san)
```

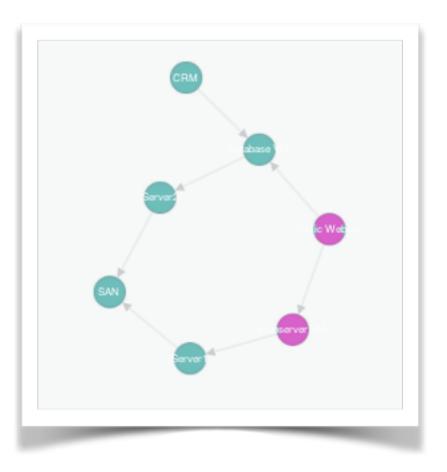




## Practical Cypher

#### Network Management - Impact Analysis



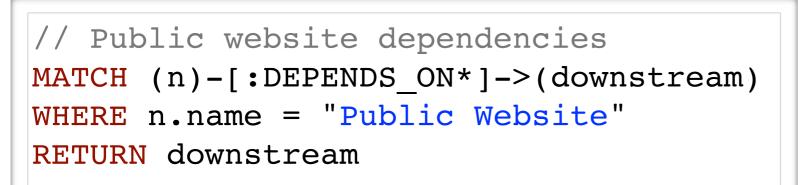


upstream

{name:"Webserver VM"}

{name:"Public Website"}

#### **Practical Cypher** Network Management - Dependency Analysis





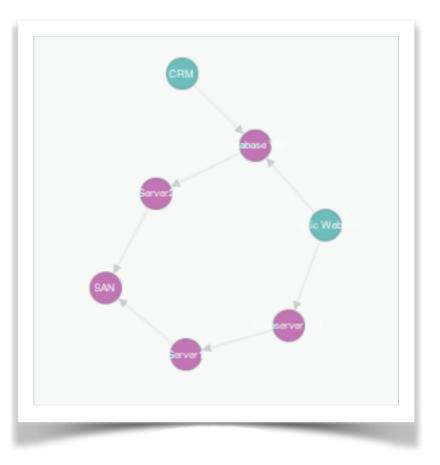
{name:"Database VM"}

{name:"Server 2"}

{name:"SAN"}

{name:"Webserver VM"}

{name:"Server 1"}



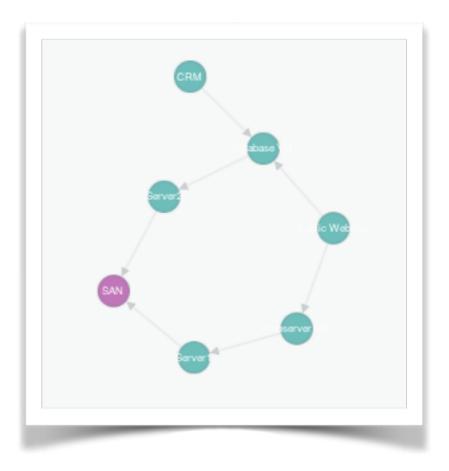




### **Practical Cypher** Network Management - Statistics

```
// Most depended on component
MATCH (n)<-[:DEPENDS_ON*]-(dependent)
RETURN n,
count(DISTINCT dependent)
    AS dependents
ORDER BY dependents DESC
LIMIT 1</pre>
```

n	dependents
{name:"SAN"}	6





# More Graphs in the Real World





#### Background

- One of the world's largest logistics carriers
- Projected to outgrow capacity of old system
- New parcel routing system
  - Single source of truth for entire network
  - B2C & B2B parcel tracking
  - Real-time routing: up to 5M parcels per day



#### **Business problem**

- 24x7 availability, year round
- Peak loads of 2500+ parcels per second
- Complex and diverse software stack
- Need predictable performance & linear scalability
- Daily changes to logistics network: route from any point, to any point

- Neo4j provides the ideal domain fit:
  - a logistics network is a graph
- Extreme availability & performance with Neo4j clustering
- Hugely simplified queries, vs. relational for complex routing
- Flexible data model can reflect real-world data variance much better than relational
- "Whiteboard friendly" model easy to understand



Industry: Communications Use case: Recommendations San Jose, CA

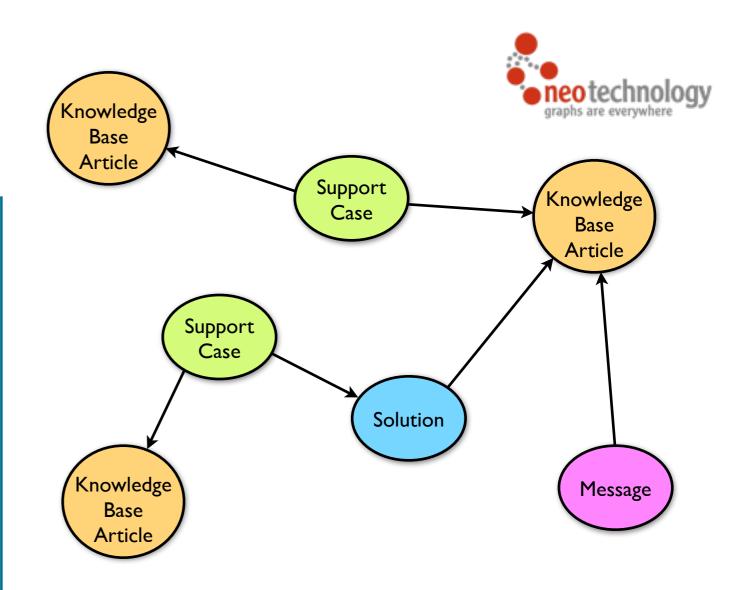
#### Cisco.com

#### Background

- Cisco.com serves customer and business customers with Support Services
- Needed real-time recommendations, to encourage use of online knowledge base
- Cisco had been successfully using Neo4j for its internal master data management solution.
  - Identified a strong fit for online recommendations

#### **Business problem**

- Call center volumes needed to be lowered by improving the efficacy of online self service
- Leverage large amounts of knowledge stored in service cases, solutions, articles, forums, etc.
- Problem resolution times, as well as support costs, needed to be lowered



- Cases, solutions, articles, etc. continuously scraped for cross-reference links, and represented in Neo4j
- Real-time reading recommendations via Neo4j
- Neo4j Enterprise with HA cluster
- The result: customers obtain help faster, with decreased reliance on customer support



Deutsche Telekom Industry: Communications Use case: Social gaming Frankfurt, Germany



#### Background

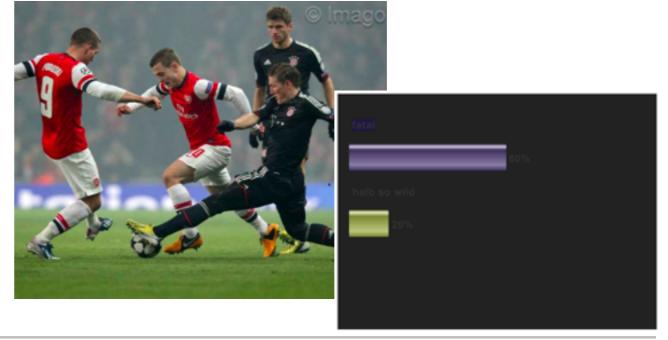
- Europe's largest communications company
- Provider of mobile & land telephone lines to consumers and businesses, as well as internet services, television, and other services

> 236,000 Employees worldwide in 2011





#### Interactive Television Programming



#### Business problem

- The Fanorakel application allows fans to have an interactive experience while watching sports
- Fans can vote for referee decisions and interact with other fans watching the game
- Highly connected dataset with real-time updates
- Queries need to be served real-time on rapidly changing data
- One technical challenge is to handle the very high spikes of activity during popular games

- Interactive, social offering gives fans a way to experience the game more closely
- Increased customer stickiness for Deutsche Telekom
- A completely new channel for reaching customers with information, promotions, and ads
- Clear competitive advantage



Industry: Social Network Use case: Social / Recommendations Seattle, WA

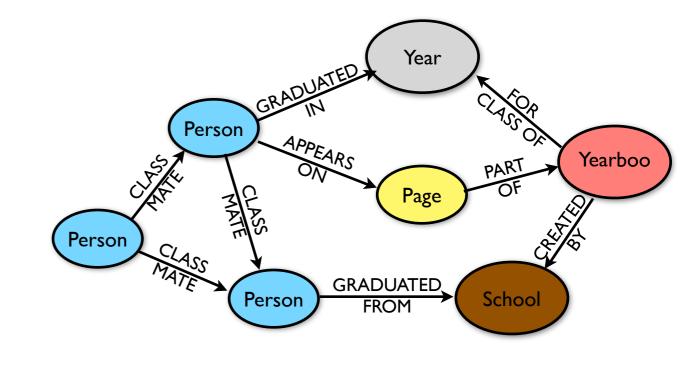


#### Background

- Memory Lane, Inc. was founded in 1995 and based in Seattle, Washington. Subsidiary of United Online, Inc.
- Classmates.com, operates an online yearbook that connects members in the United States and Canada with friends and acquaintances from school, work, and the military.
- Evolving toward more sophisticated social networking capability

#### **Business problem**

- Develop new Social capabilities to help monetize Yearbook-related offerings
  - Show me all the people I know in a yearbook
  - Show me yearbooks my friends appear in most often (i.e. "Top yearbooks to look at")
  - Show me sections of a yearbook that your friends appear most in (i.e. "8 of your friends are on page 12 with the football team)
  - Show me other high schools that my friends went to (i.e. friends you made in other schools) Neo Technology Confidential



- 3-Instance Neo4j Cluster with Cache Sharding + Disaster-Recovery Cluster
- Neo4j provides 18 ms response time for the top 4 queries
- Initial graph size: 100M nodes and 600M relationships
  People, Images, Schools, Yearbooks, Yearbook Pages
- Projected to grow to IB nodes & 6B relationships

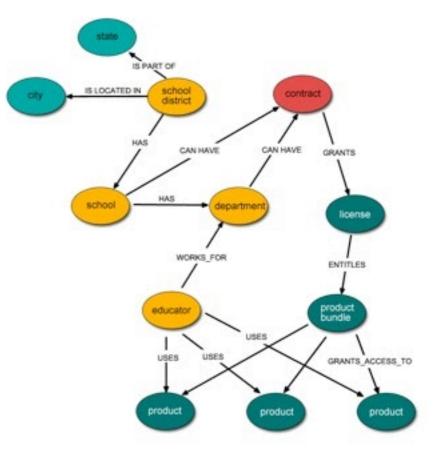


Industry: Education Use case: Resource Authorization & Access Control San Francisco, CA



#### Background

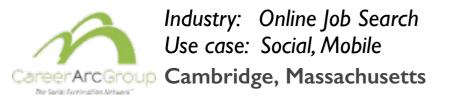
- Teachscape, Inc. develops online learning tools for K-12 teachers, school principals, and other instructional leaders.
- Teachscape evaluated relational as an option, considering MySQL and Oracle.
- Neo4j was selected because the graph data model provides a more natural fit for managing organizational hierarchy and access to assets.



#### **Business problem**

- Neo4j was selected to be at the heart of a new architecture.
- The user management system, centered around Neo4j, will be used to support single sign-on, user management, contract management, and end-user access to their subscription entitlements.

- Domain and technology fit
  - simple domain model where the relationships are relatively complex. Secondary factors included support for transactions, strong Java support, and well-implemented Lucene indexing integration
- Speed and Flexibility
  - The business depends on being able to do complex walks quickly and efficiently. This was a major factor in the decision to use Neo4j.
- Ease of Use
  - accommodate efficient access for home-grown and commercial off-the-shelf applications, as well as ad-hoc use.
  - Extreme availability & performance with Neo4j clustering
  - Hugely simplified queries, vs. relational for complex routing
  - Flexible data model can reflect real-world data variance much better than relational
  - "Whiteboard friendly" model easy to understand





#### Background

- Own several job search properties, including Internships.com, TweetMyJobs, and CareerBeam
- CareerArc Group ecosystem now serves
   30 million annual visitors, 40,000 companies
   (including more than 50% of the Fortune 100),
   and over 350 universities



#### **Business problem**

- Launching new application:"Who do you know on Facebook"
- Existing SQL Server application difficult to develop against, and slowing down as user volumes increased
- Looking for a solution that could easily manage a large amount of interrelated data

- Neo4j's was a natural fit for the social graph underlying CareerArcGroup's professional social network
- Significantly accelerated R&D, overcoming the agility problems the team had been facing
- Strong, reliable performance on EC2, running reliably inside of existing EC2 architecture
- Initial go-live included 10 million nodes, 30 million relationships and 235 million properties



Industry: Communications Use case: Social, Mobile Hong Kong



#### Background

- Hong Kong based telephony infrastructure provider (aka M800 aka Pop Media)
- Exclusive China Mobile partner for international toll-free services. SMS Hub & other offerings
- 2012 Red Herring Top 100 Global Winner



#### **Business problem**

- Launched a new mobile communication app "Maaii" allowing consumers to communicate by voice & text (Similar to Line, Viber, Rebtel, VoxOx...)
- Needed to store & relate devices, users, and contacts
- Import phone numbers from users' address books. Rapidly serve up contacts from central database to the mobile app
- Currently around 3M users w/200M nodes in the graph

- Quick transactional performance for key operations:
  - friend suggestions ("friend of friend")
  - updating contacts, blocking calls, etc.
  - etc.
- High availability telephony app uses Neo4j clustering
- Strong architecture fit: Scala w/Neo4j embedded



#### %onefinestay

Industry: Hospitality Use case: Social Mobile London, England

#### Background

- London-based 'A-list' home listing & booking
- Web & mobile app w/visible floor plans
- Luxury benefits & home convenience in international cities (NYC, London) 24/7 phone services, suggested eateries/events/hot spots
- Travel agency partners
- 2010 had 6 apts 2012, = 700 & growing
- Bloomberg Business "Like living in a catalog"

#### **Business problem**

- Needed to serve up structure & floor plan details, and related spatial uses: "where's linen?"
- Relational database couldn't unlock the power of data & balance features with support a
- Needed complex structural data modeling with speed, accuracy and flexibility
- Serving thousands of listings & rapidly growing





- 6-week implementation, integrated w/framework
- optimized workflow w/user interface = easy entry
- adds operation speed/efficiency, saves time/money
- scales easily, supports growth, strong community
- Best NoSQL alternative, w/Neo4j query performance & admin interface



Industry: Health Care Use case: Recommedatations Newton, Massachusetts



#### Background

- Founded in 1999.Widely considered the industry leader in patient management for discharges & referrals
- Manage patient referrals for more than 4600 health care facilities
- Connects providers, payers and suppliers via secure electronic patient-transition networks, and web-based patient management platform

## No other patient management platform is this connected to results.



#### **Business problem**

- Satisfy complex "Graph Search" queries by discharge nurses and intake coordinators, e.g.: "Find a skilled nursing facility within n miles of a given location, belonging to health care group XYZ, offering speech therapy and cardiac care, and optionally Italian language services"
- Real-time Oracle performance not satisfactory
- New functionality called for more complexity, including granular role-based access control

- Fast real-time performance needs now satisfied
- Queries span multiple hierarchies, including provider graph & employee permissions graph
- Graph data model provided a strong basis for adding more dimensions to the data, such as insurance networks, service areas, and ACOs (Accountable Care Organizations)
- ty, Some multi-page SQL statements have been turned into one simple function with Neo4j



Industry: Health Care Use case: Bioinformatics **Cambridge, Massachusetts** 

#### Background

- Clinical diagnostics company specializing in genetic carrier screening for inherited diseases
- Founded in 2008 by Harvard Business School & Harvard Medical School graduates
- Two sides of the business: Clinical and R&D
- Particularly strong in the detection of rare alleles and measuring frequency in the population





#### **Business problem**

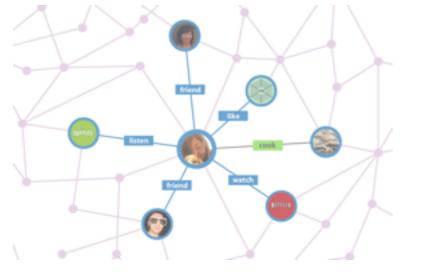
- Clinical data split across several operational databases that are not structured for discovery
- Needed an easy query mechanism for scientists who are not data scientists."Graph search" for bioinformatics.
- Much in Bioinformatics remains unknown: having to specifying a schema ahead of time can range from difficult to impossible.

- New R&D database build atop Neo4j to support information discovery by scientists
- Lightweight web front end allows simple Cypher queries to be constructed ad hoc
- Raw VCF sequence data imported into Neo4j, along with clinical data from Oracle database
- Time to answer new questions went from days of ad-hoc information gathering to hours or minutes

# Gartner's "5 Graphs" Consumer Web Giants Depends on Five Graphs **Interest Graph Intent Graph Social Graph**

**Payment Graph** 

Mobile Graph





## Questions ?

