

The Minimum Wiener Connector Problem

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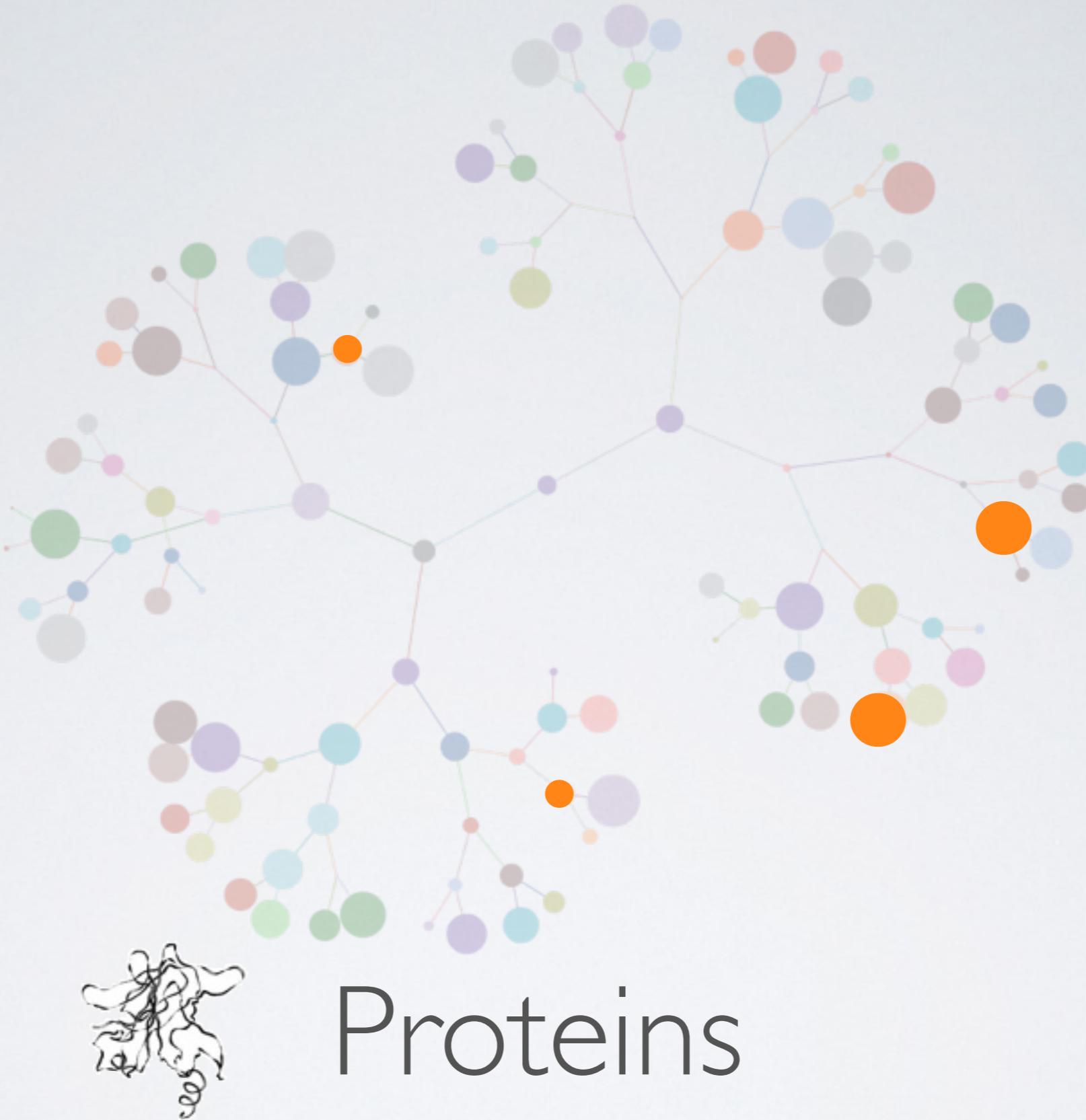






Infected Patients

Who is the culprit?



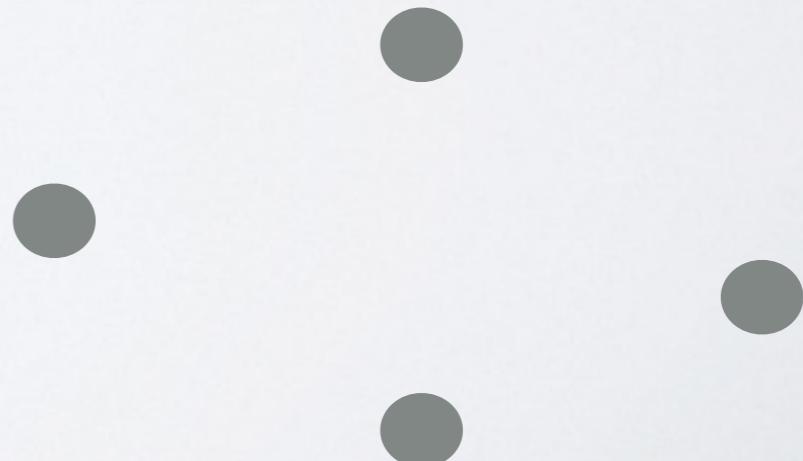
Proteins

Which other proteins participate in pathways?

General Problem

Given a graph $G = (V, E)$ and a set of query vertices $Q \subseteq V$,
find a *small* subgraph H of G that
“explains” the connections existing among Q .

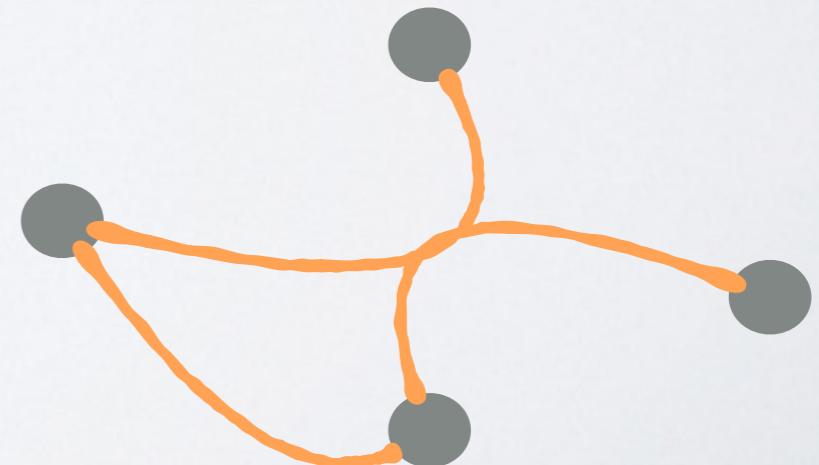
Call this query-dependent graph, H , a *connector*.



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Related Work

Random-walk

Run a random walk from each query node.

Identify a neighborhood of each node.

Combine neighborhoods.

Many parameters
Large solutions

Search

Search for a subgraph that best meets objective.

Steiner Tree

Find the smallest tree that connects query nodes.

No interpretation

Motivating Observation

A natural sense of closeness in graphs is captured by **short paths**.



Objective

We define a new problem where the objective is to:

minimize the sum of pairwise shortest-path-distances
between nodes **in the connector H .**

If $d(u, v)$ is the shortest-path distance, we want:

$$\text{minimize} \quad \sum_{(u,v) \in H} d(u, v)$$

In fact this quantity is called the **Wiener Index**.

Wiener Intuitions

Path is **largest**:



$$3+2+1+2+1+1 = 10$$

Clique/Star is **smallest**:



$$2+2+1+2+1+1 = 9$$

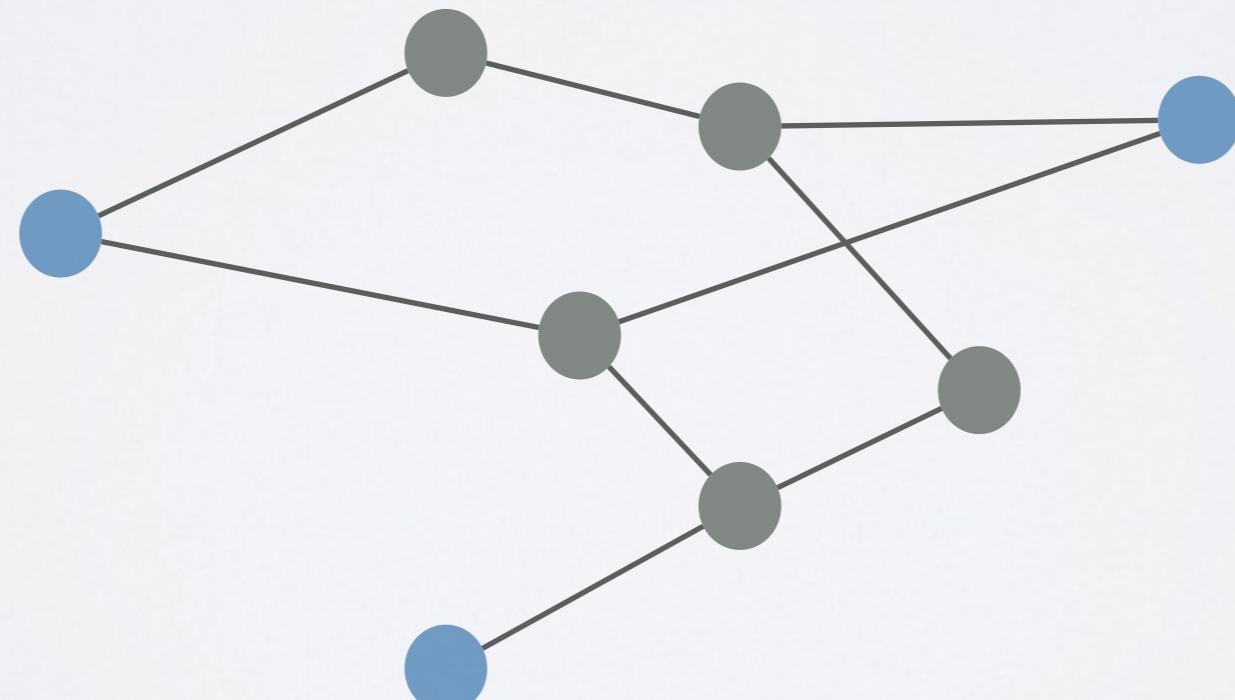
Favors star-shape, closeness.

Provides a numerical feedback of connectedness.

Minimum Wiener Connector

Given a graph $G = (V, E)$ and a query set $Q \subseteq V$, find a connector H^* for Q in G with smallest Wiener index.

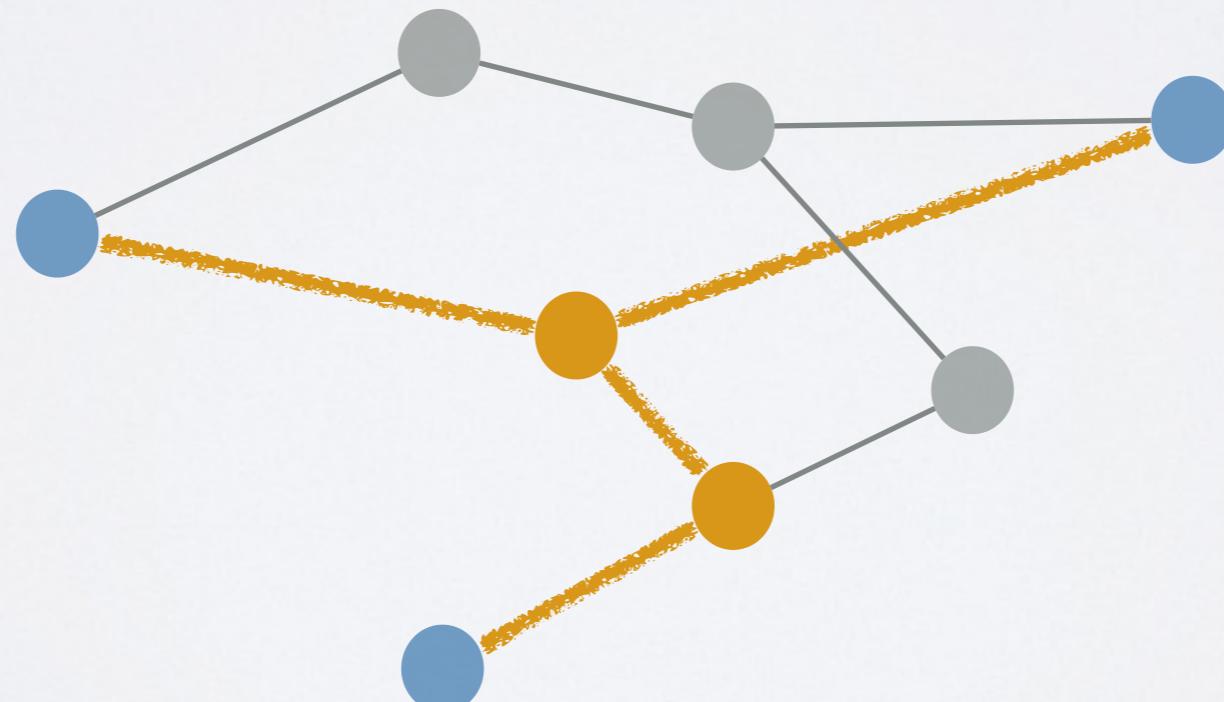
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Call H^* the **minimum Wiener connector**.

There is no explicit size constraint, but rewriting

$$W(H^*) = \sum_{\{u,v\} \subseteq V(H^*)} d_H(u, v) = \binom{|V(H^*)|}{2} * \text{average } d$$

uncovers a tradeoff between **size** and **average distance**

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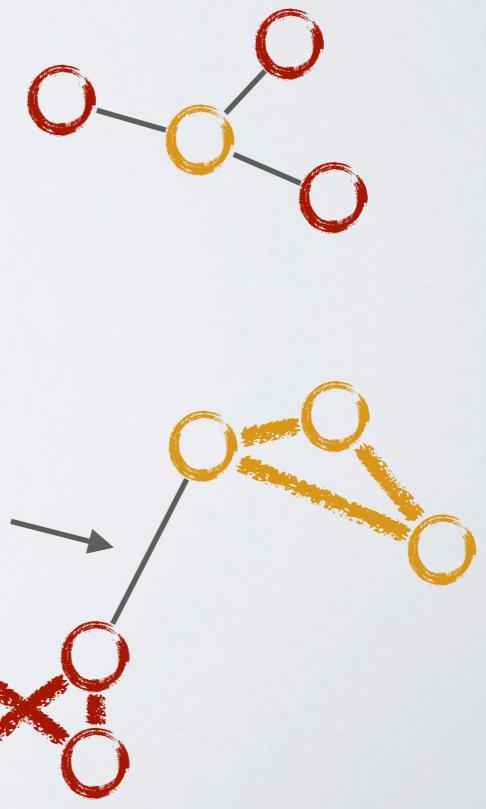
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Summary Of Results



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With the Wiener Index as our objective, we propose:

a **constant factor approximation** algorithm that runs in $\tilde{O}(|Q| |E|)$

Using this we find solutions that are aside from being **close to optimal**:
small, meaningful, and **amenable to visualization**.

For query nodes belonging to the **same** community:

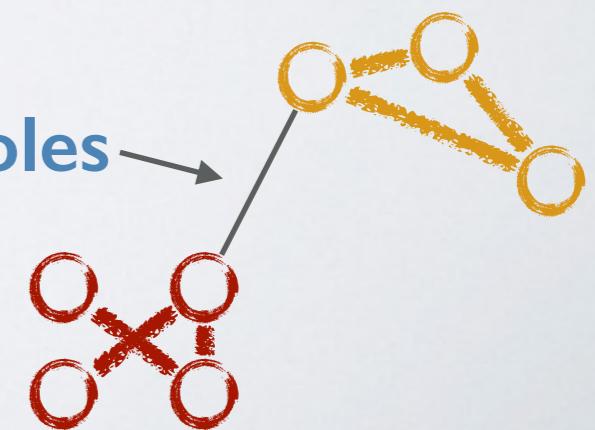
connector contains nodes of **high centrality**



For query nodes from **different** communities:

connector contains nodes that span **structural holes**

(incident to edges that bridge communities)



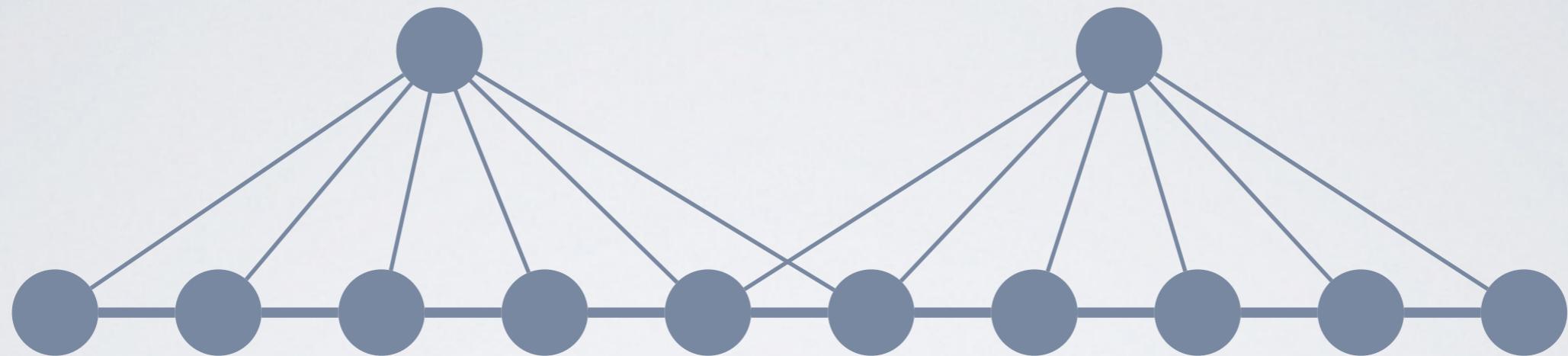
How Do We Find The
Minimum Wiener Connector?

No. Not The Steiner Tree

Steiner Tree: Given a graph $G = (V, E)$ and a set of query nodes (terminals) $Q \subseteq V$, find the smallest tree connecting all terminals.

Minimizing the number of edges will **not** necessarily result in the smallest Wiener Index!

Steiner vs Wiener

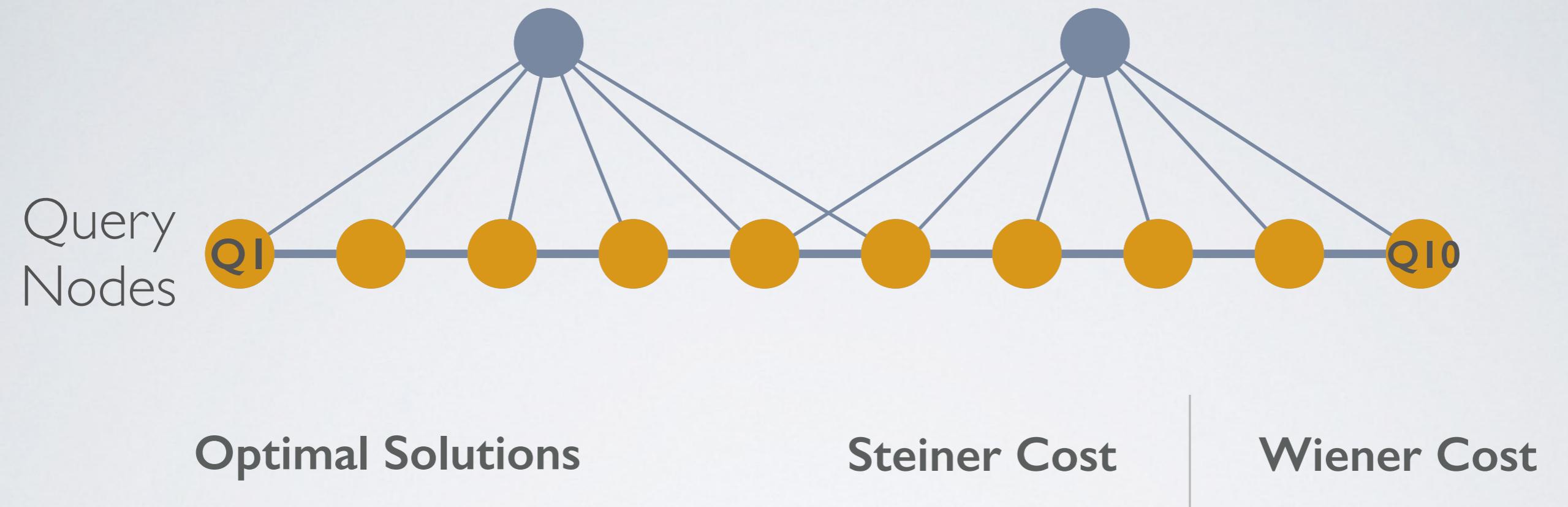


Optimal Solutions

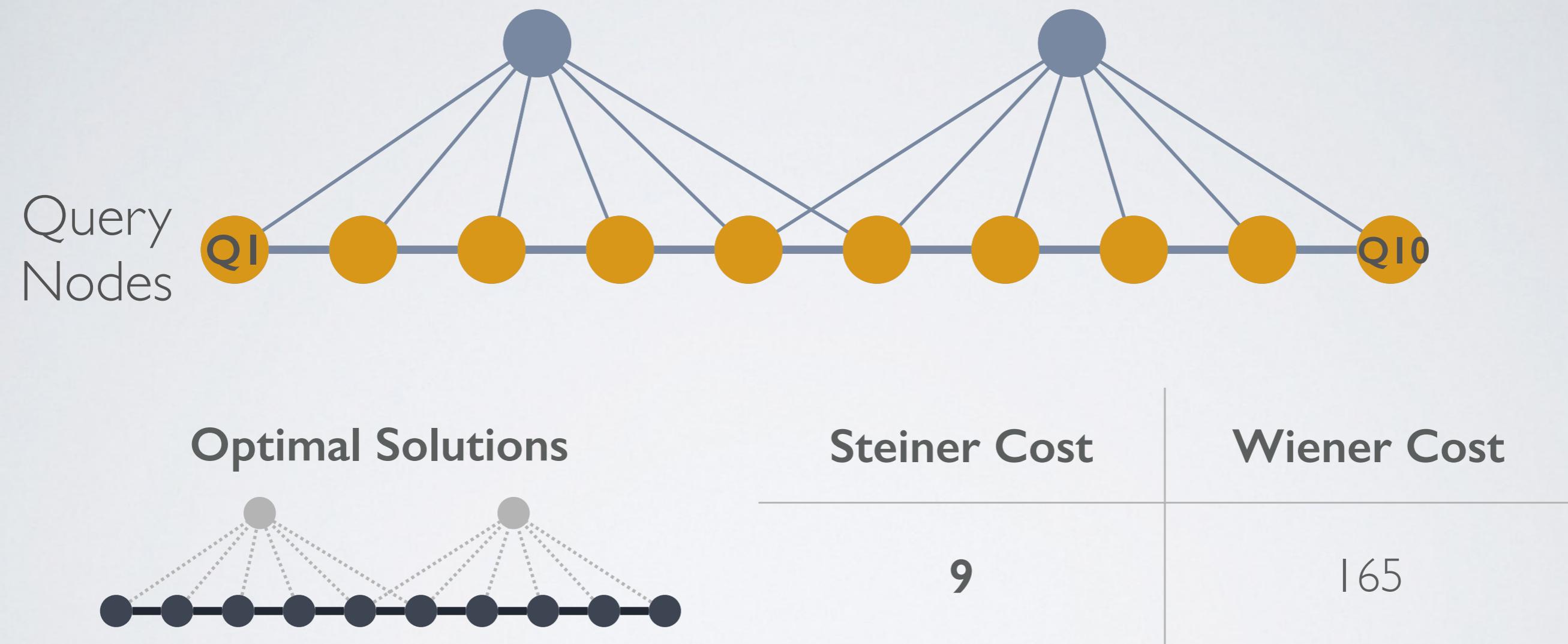
Steiner Cost

Wiener Cost

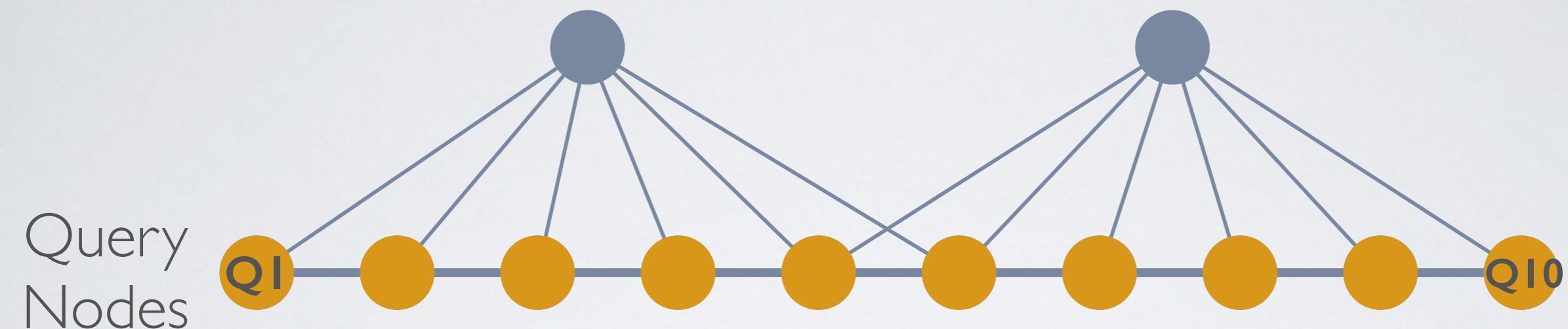
Steiner vs Wiener



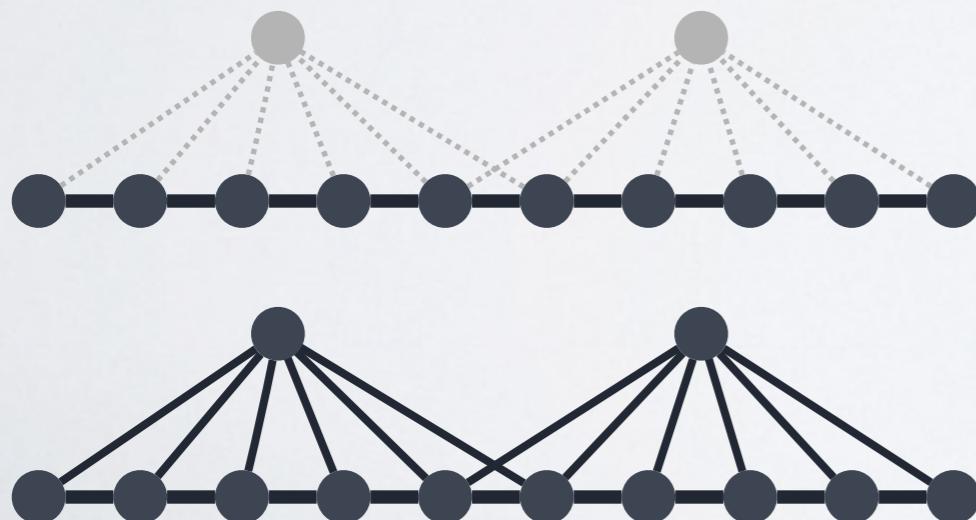
Steiner vs Wiener



Steiner vs Wiener



Optimal Solutions



Steiner Cost

9

21

Wiener Cost

165

142

Relaxations

Original Objective

Relaxed Objective

Reduce to classic Steiner Tree with carefully constructed edge weights.

Relaxations

Original Objective

All pairwise distances



Relaxed Objective

Distances from a root r

Reduce to classic Steiner Tree with carefully constructed edge weights.

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Relaxed Objective

Distances from a root r

Measure distance in H



Measure distance in G

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Measure distance in G

Product in objective



Linear objective

Reduce to classic Steiner Tree with carefully constructed edge weights.

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Product in objective



Linear objective

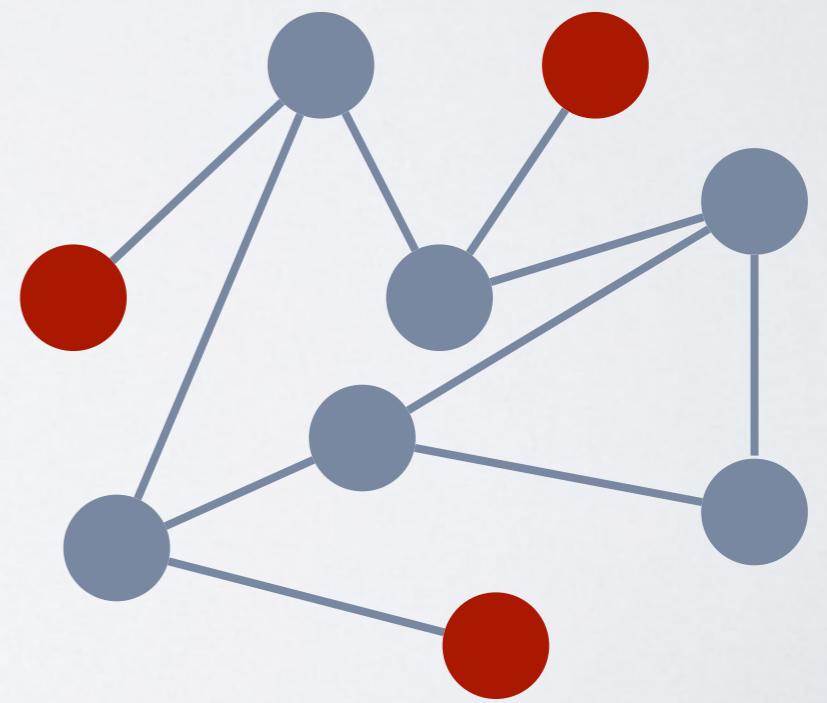
Node weights



Edge weights

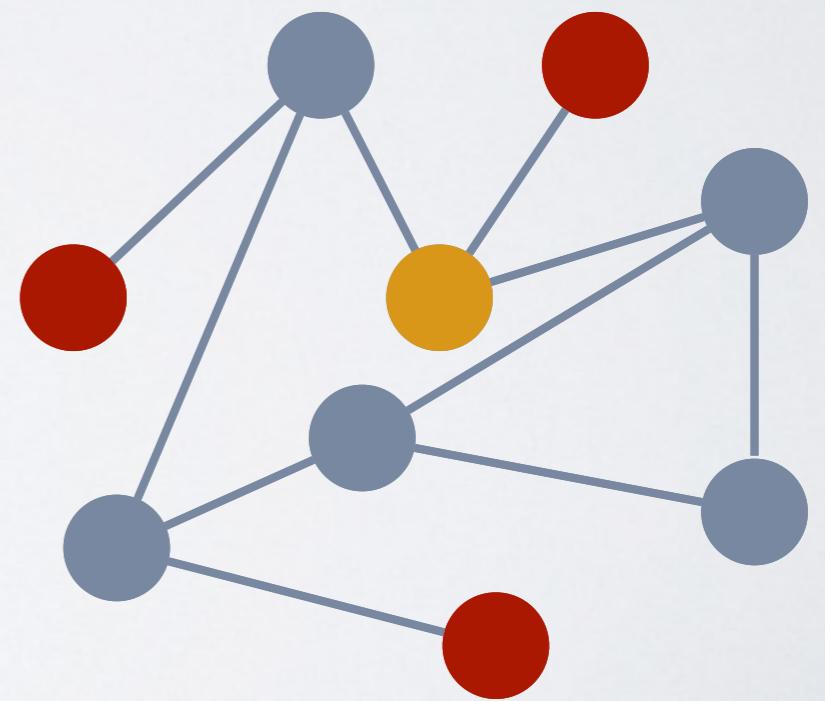
Reduce to classic Steiner Tree with carefully constructed edge weights.

wienerSteiner(G, Q)



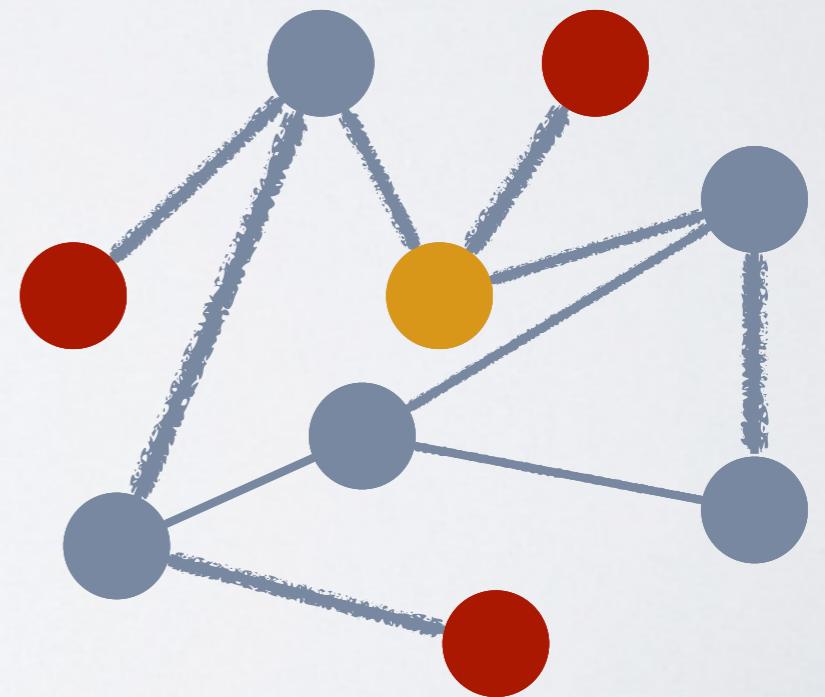
wienerSteiner(G, Q)

- For each vertex $r \in V$



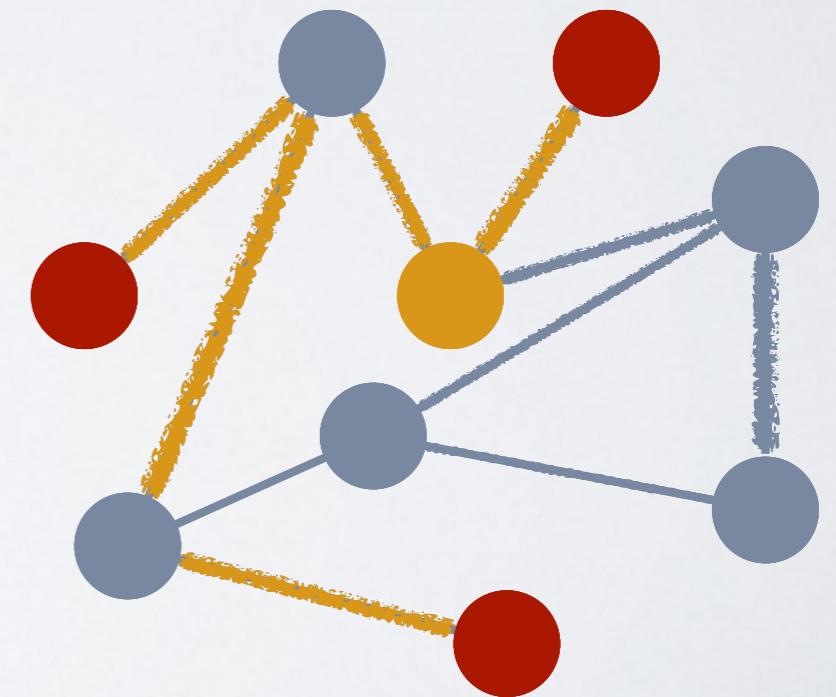
wienerSteiner(G, Q)

- For each vertex $r \in V$
 - I. Compute $d_G(u, v)$ from r to each vertex u
 2. Construct an edge-weighted graph



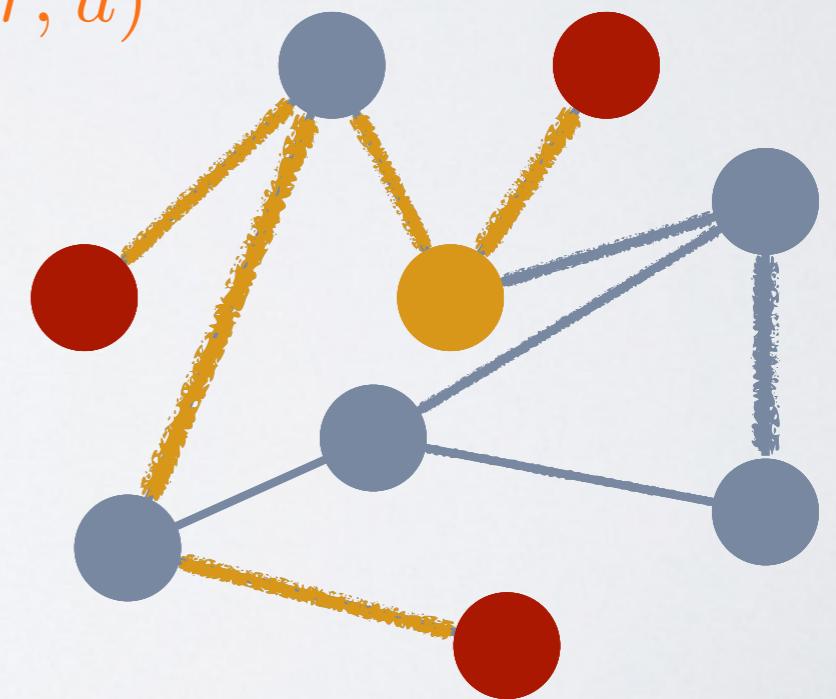
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- For each vertex $r \in V$
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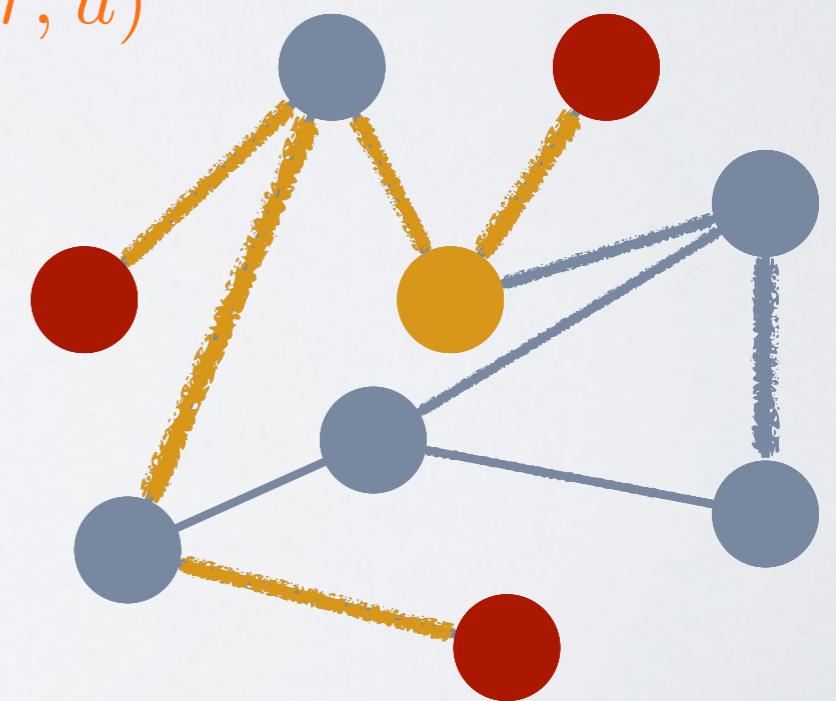
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 3. Find an approximate Steiner tree S_r^*
 4. Check for paths where $d_G(r, u) < d_{S^*}(r, u)$
- Pick S_r^* that minimizes $W(S^*)$



Case Studies

Case Study 1: Karate Club

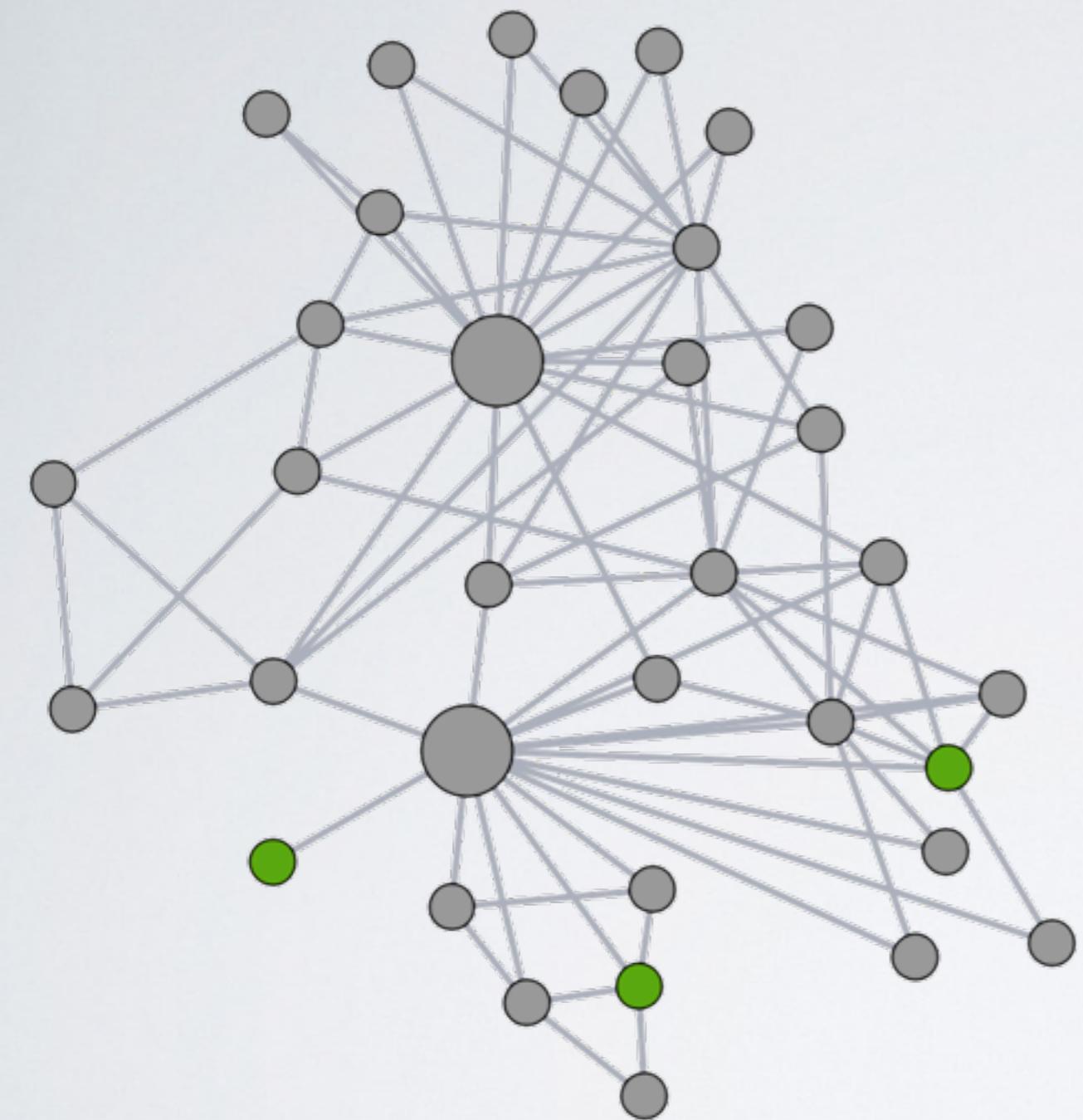


Two clusters around each
karate master.

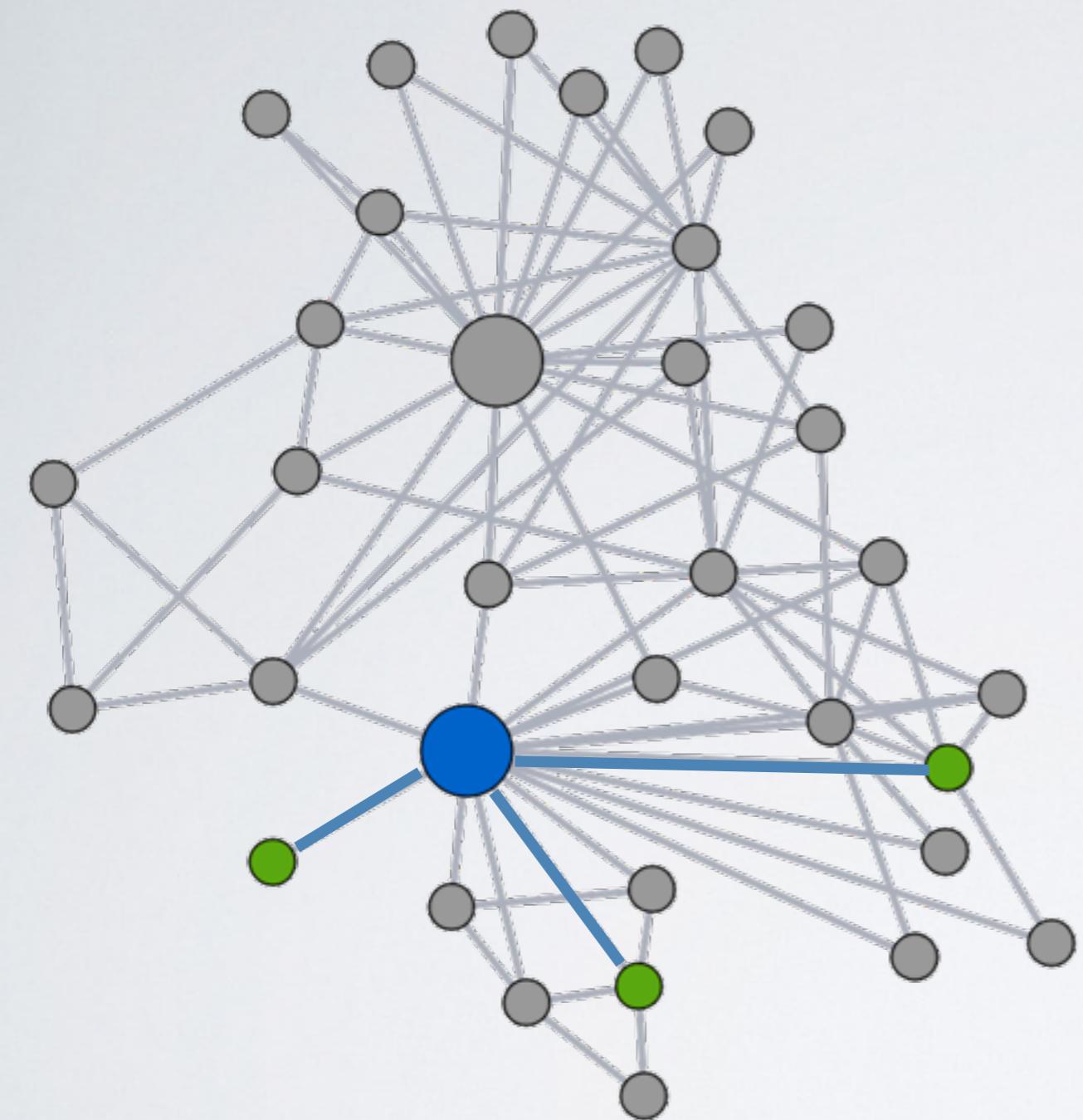
Few nodes with mixed loyalty.

By querying arbitrary nodes,
can we learn about their
loyalty without any outside
meta information?

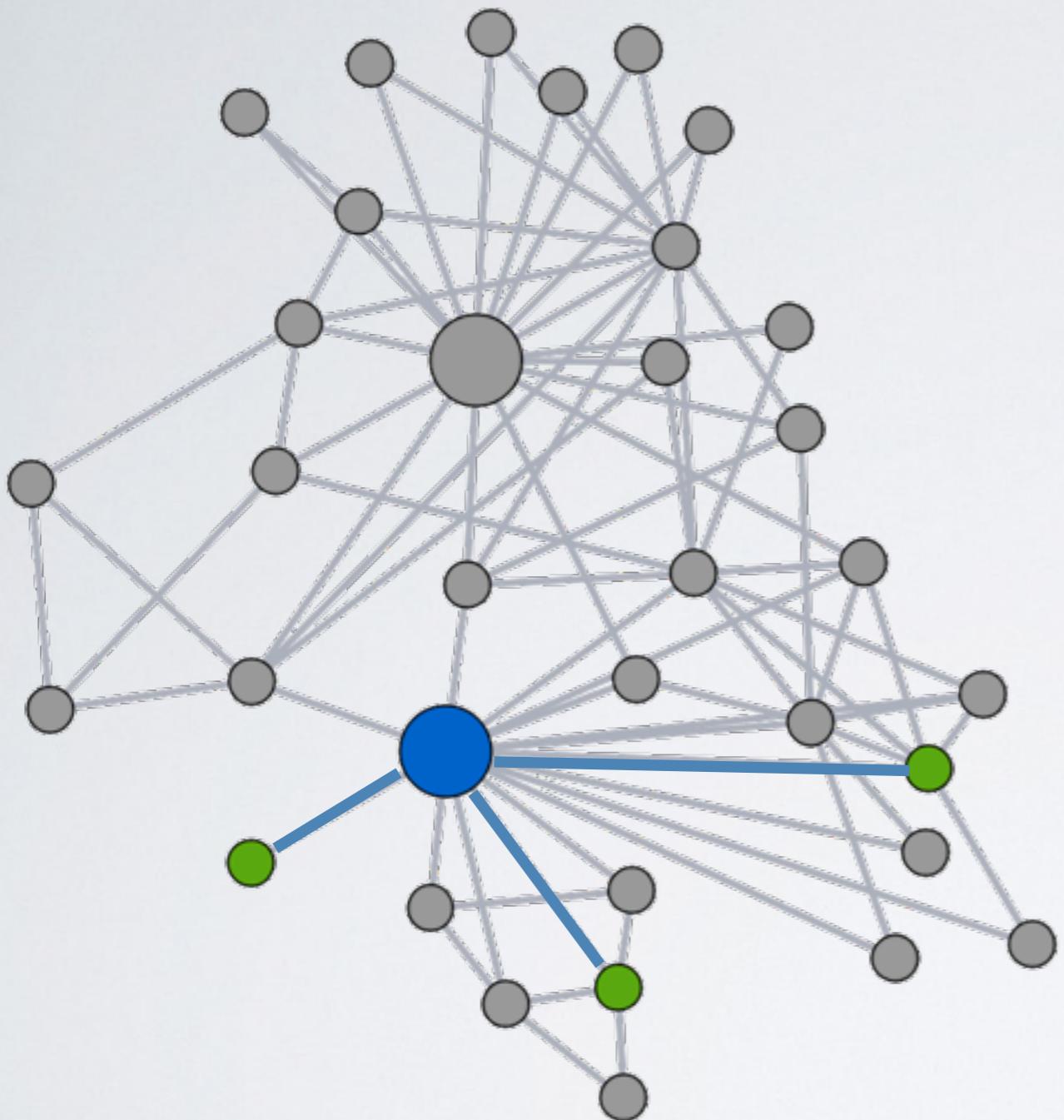
Same Community



Same Community



Same Community



Different Communities

Same Community



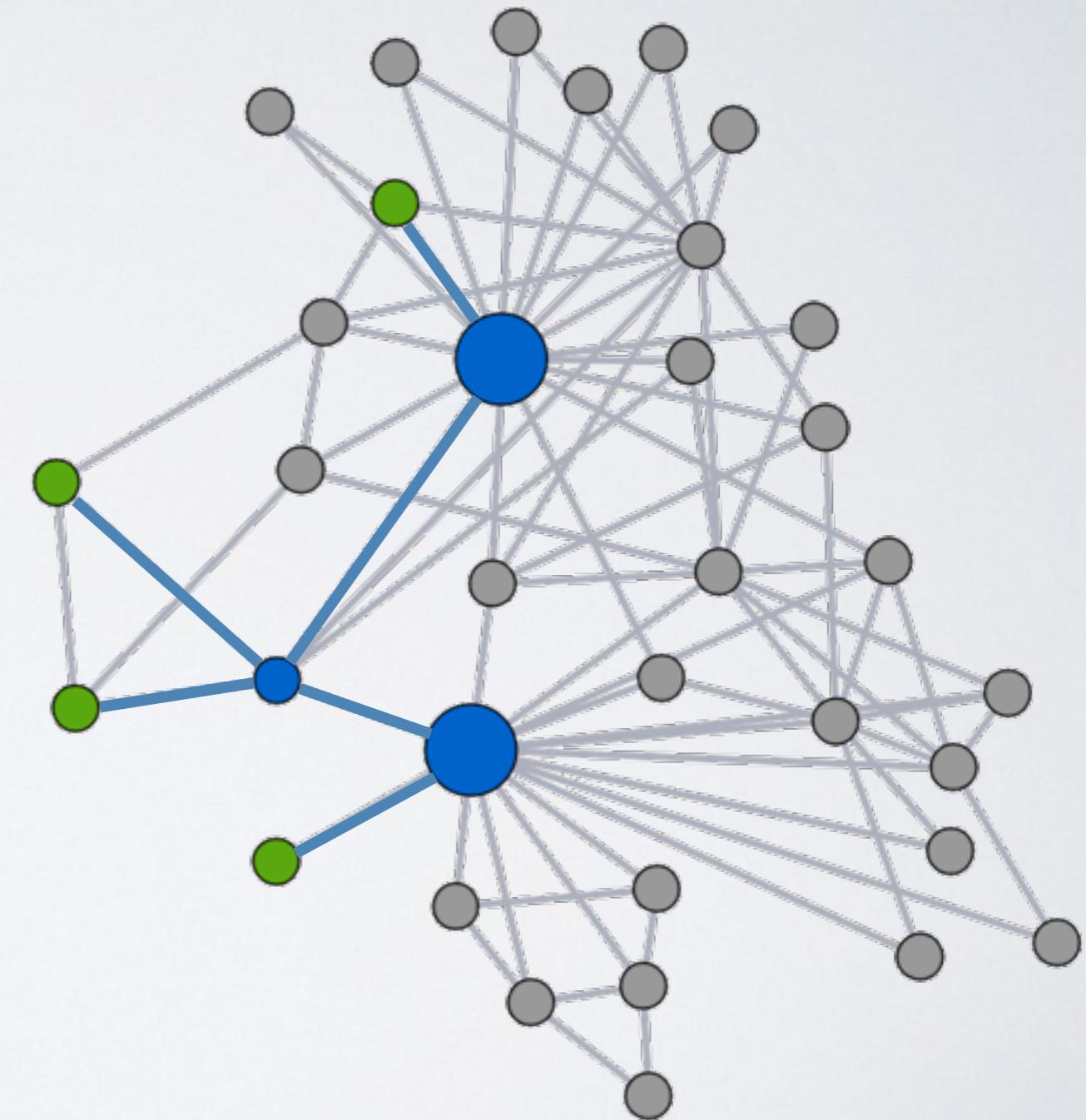
Different Communities



Same Community



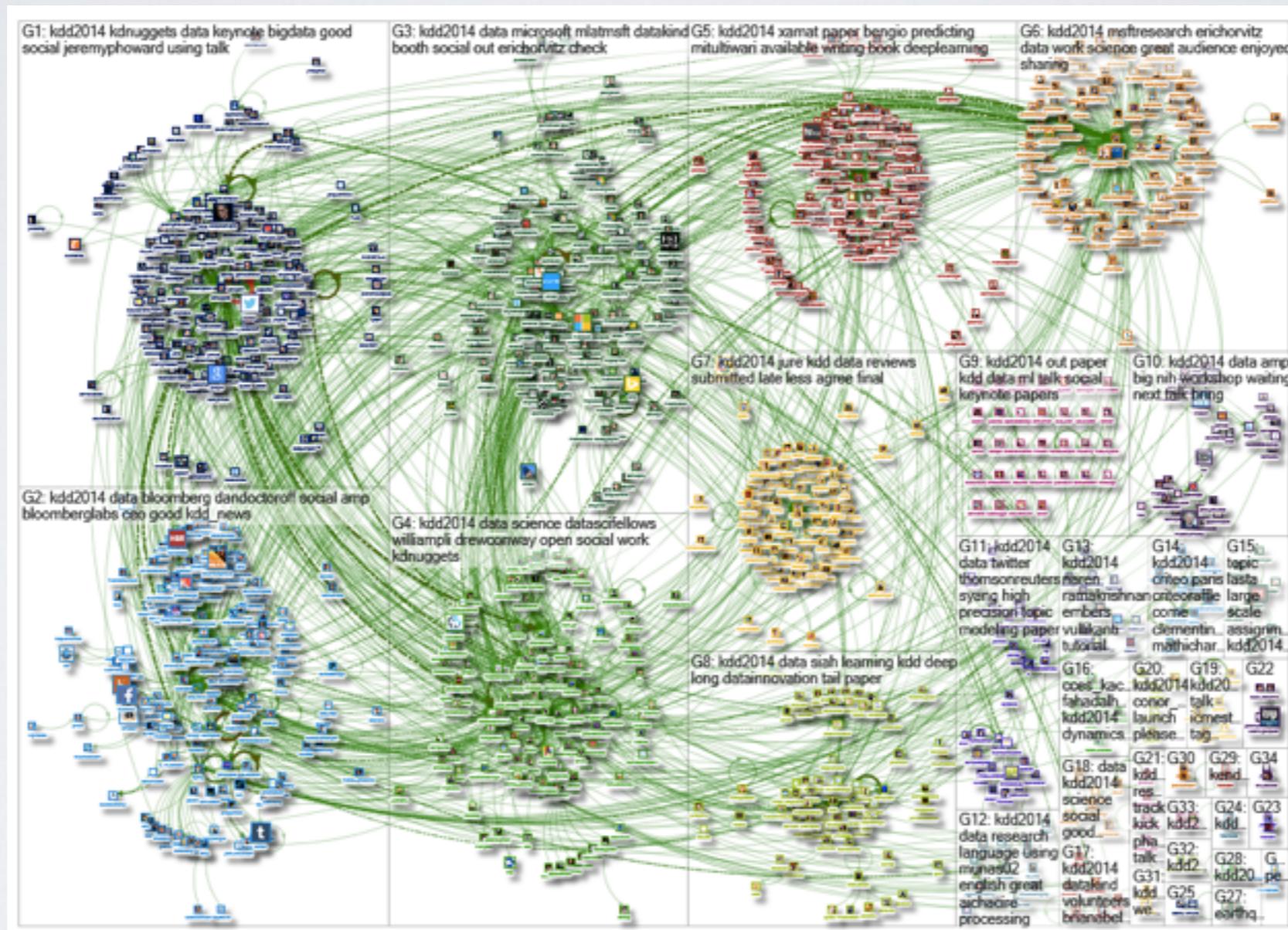
Different Communities



Case Study 2: KDD Tweets

KDD 2014 Tweets

Graph of Twitter users taking part in KDD 2014,
with an edge between replies or mentions.



Clustered into 10 communities.

jonkleinberg

thrillscience

destrin

Group 13

jonkleinberg

gizmonaut

Group 10

destrin

irescuapp

Group 1

thrillscience

jromich

kdnuggets

drewconway

Group 13

jonkleinberg

gizmonaut

Top tweeter in
G13

Group 10

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Group 13

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Top tweeter in
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irescuapp

Group 1

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drewconway

Top replied-to
in **G1**

Group 13

jonkleinberg

gizmonaut

Top tweeter in
G13

Group 10

destrin

irescuapp

Top mentioned
in **G10**

Group 1

thrillscience

kdnuggets

drewconway

jromich

Top replied-to
in **G1**

Group 13

jonkleinberg

gizmonaut

Group 10

destrin

irescuapp

Group 1

Top mentioned/
betweenness/word
in **entire graph**

thrillscience

kdnuggets

drewconway

Top mentioned/
replied-to in
entire graph

jromich

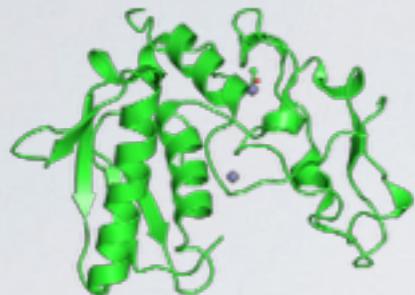
Case Study 3: PPI Network

Biology Dataset

Protein-Protein-Interaction (PPI) network collected from BioGrid3 with 15 312 vertices.

- Do they interact?
- How are they related?
- Which disease are they associated with?
- Which well-known proteins are ‘closest’ to each?

BMP1



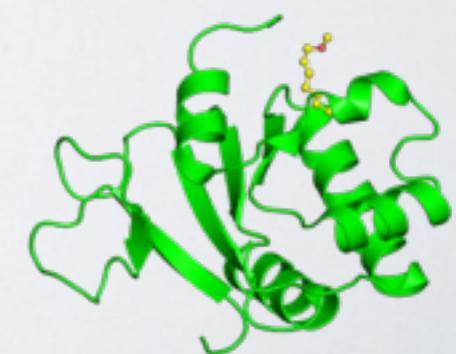
JAK2

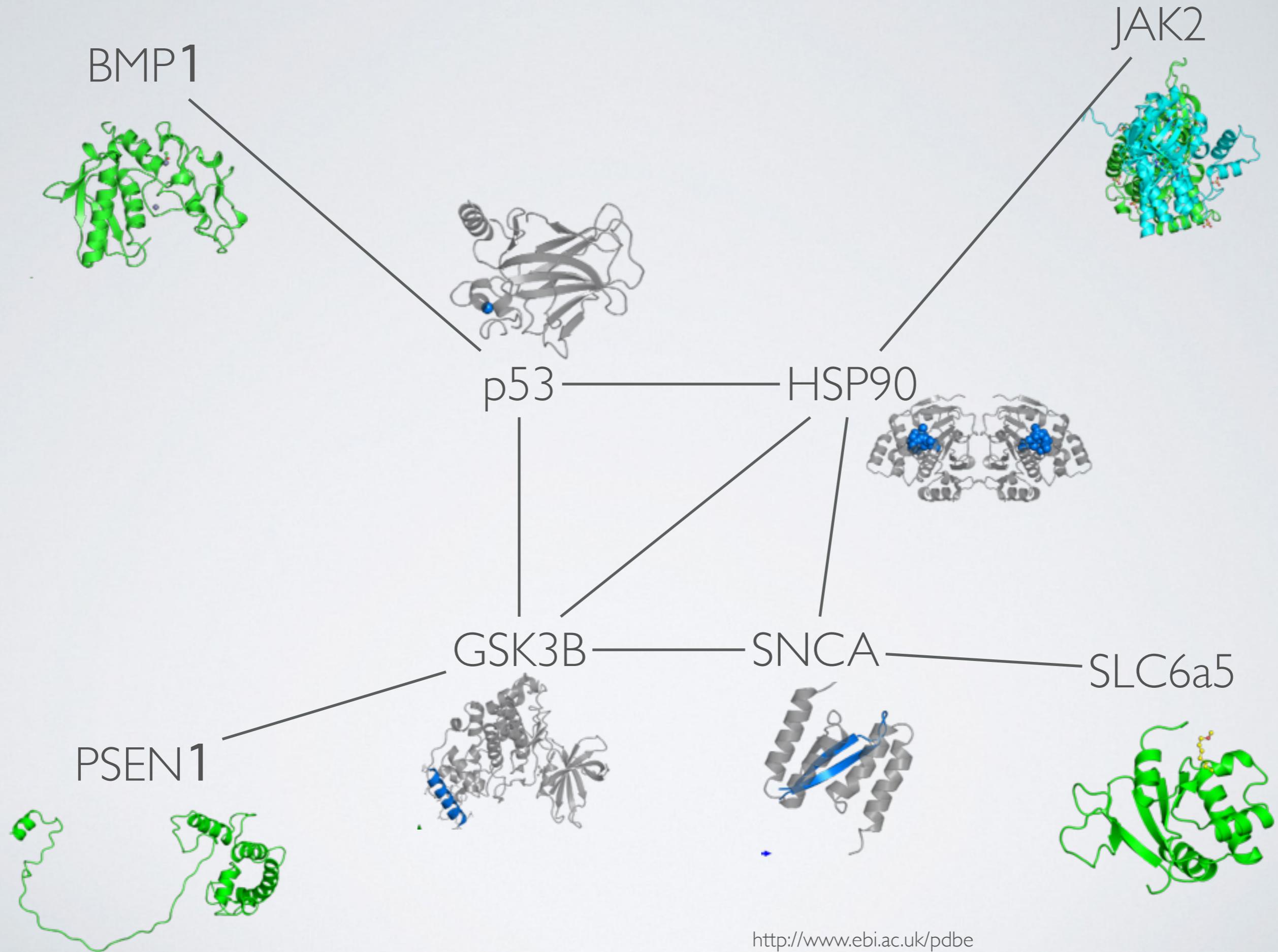


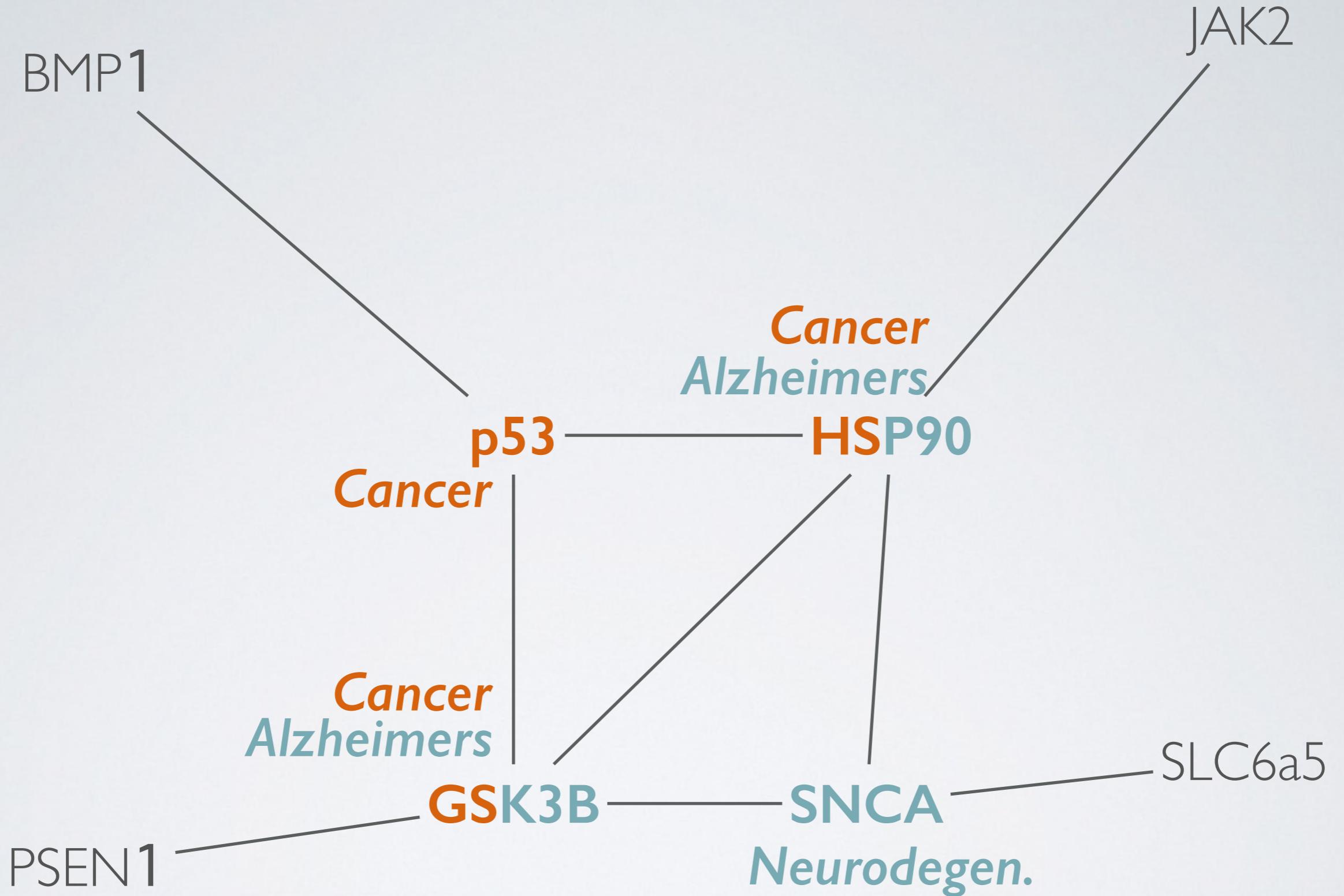
PSEN1

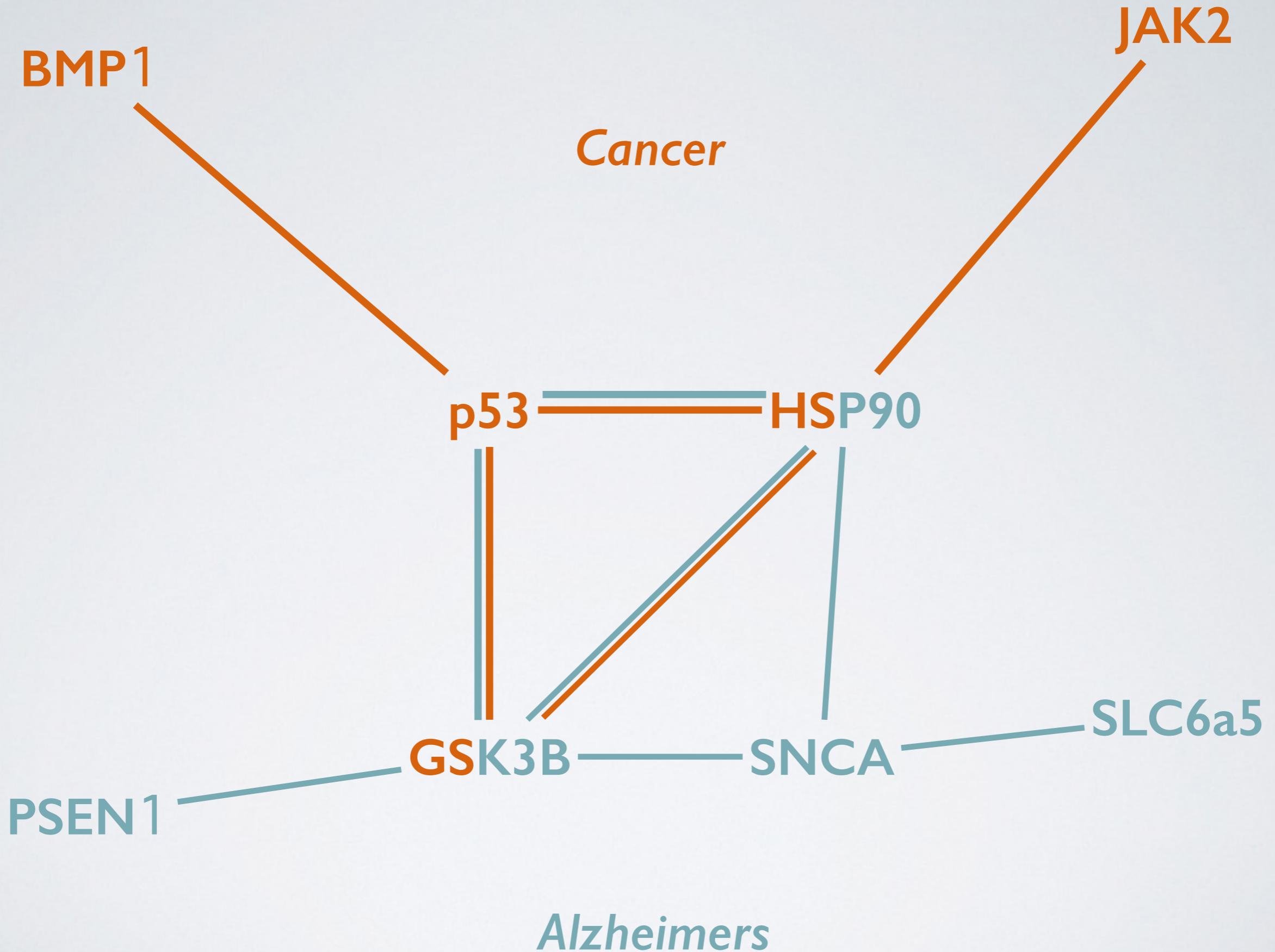


SLC6a5



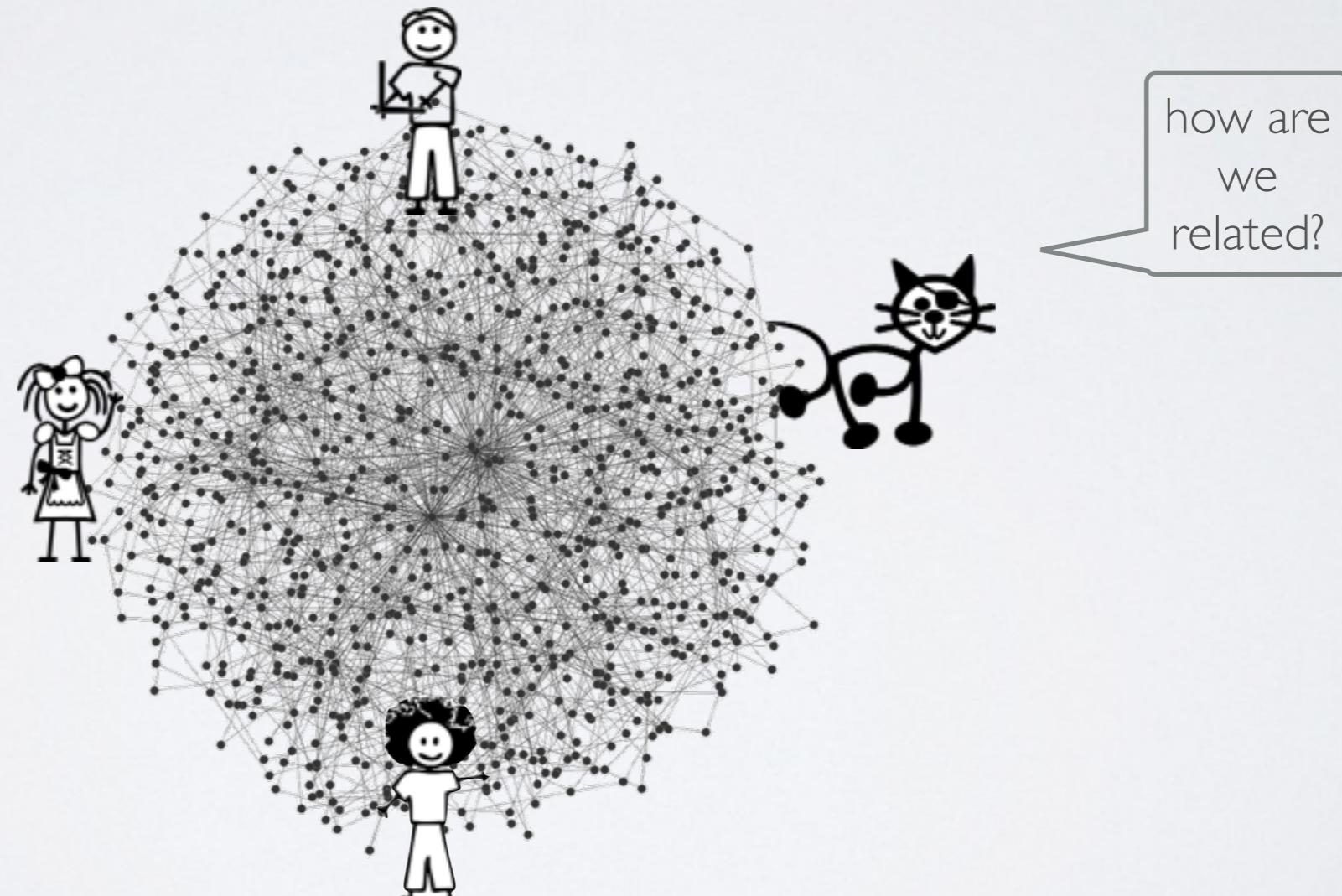






What Was The Point?

Finding a **connector for a set of query nodes** in a graph is an interesting and relevant problem.



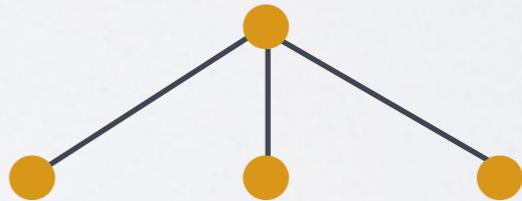
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Finding a **connector for a set of query nodes** in a graph is an interesting and relevant problem.

The Wiener Index is the **sum of shortest-path distances**, which is intuitive graph measure of closeness.



high



low

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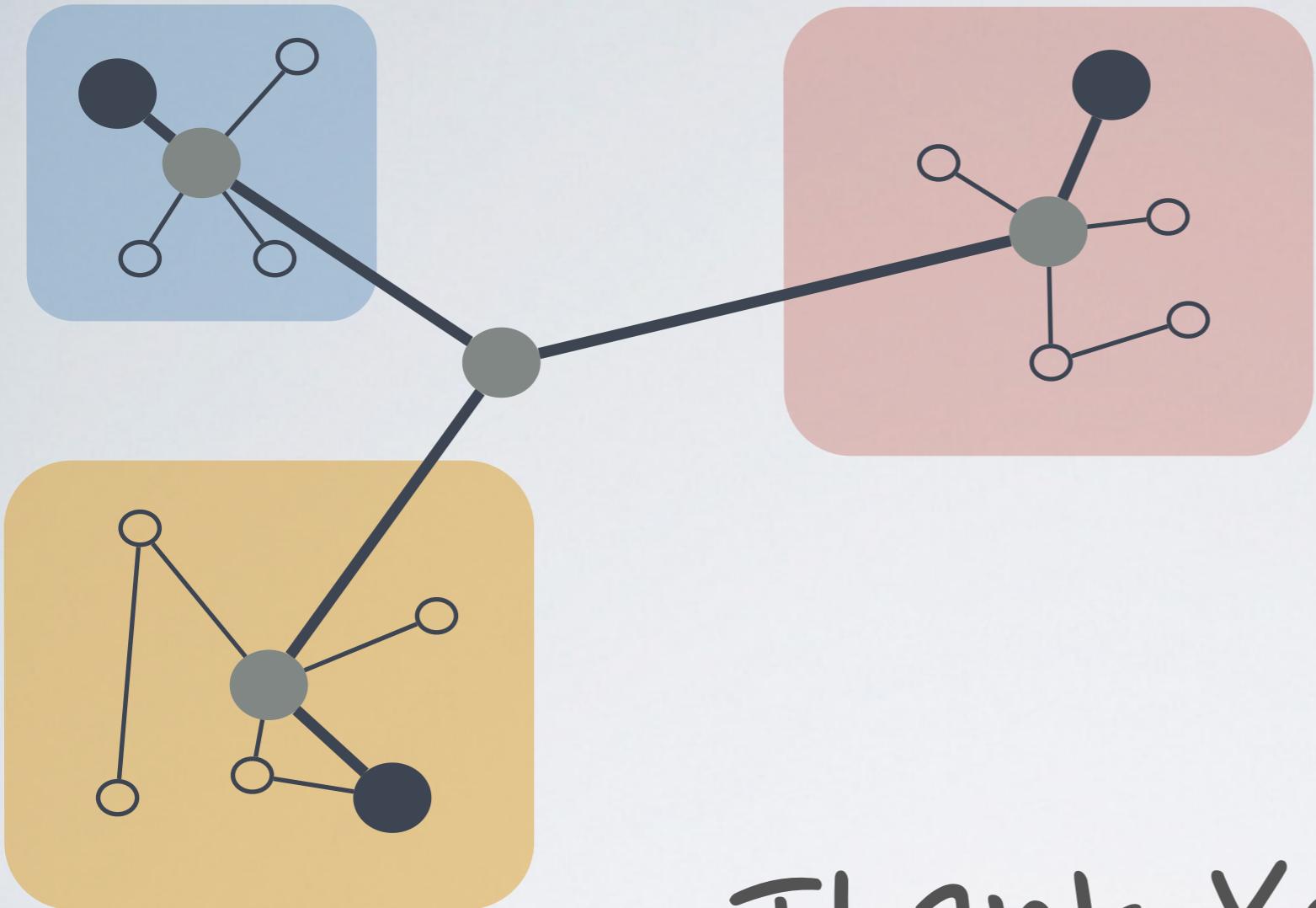
Proposed a constant factor approximation algorithm, that

- finds **small solutions**
- that are **easy to visualize**
- contain **important, central nodes**
- that **convey the relationship** among query nodes
- in a small amount of time.

Further Experiments

- Scalability
- Ground Truth Communities
- Steiner Tree Benchmark Datasets (DIMACS Challenge 2015)
- Comparison to Integer Program
- (and proofs)

Read the paper!



Thank You.

https://en.wikipedia.org/wiki/Wiener_Connector