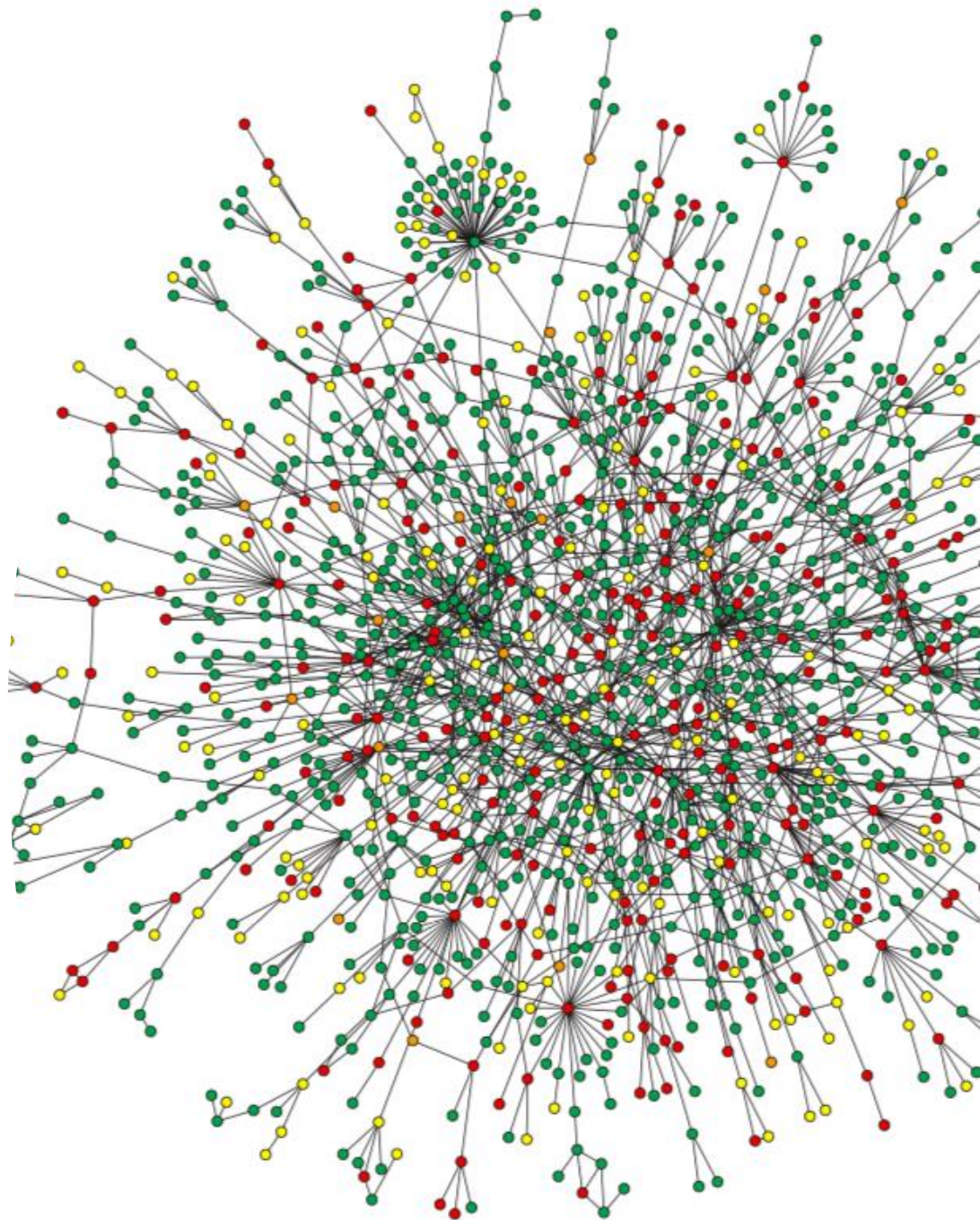


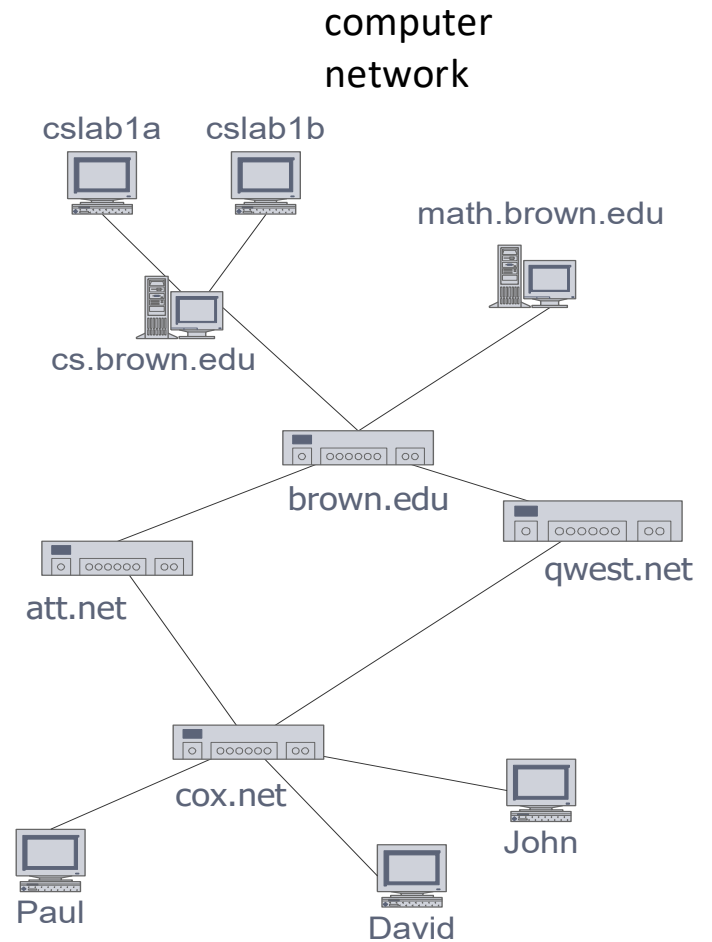
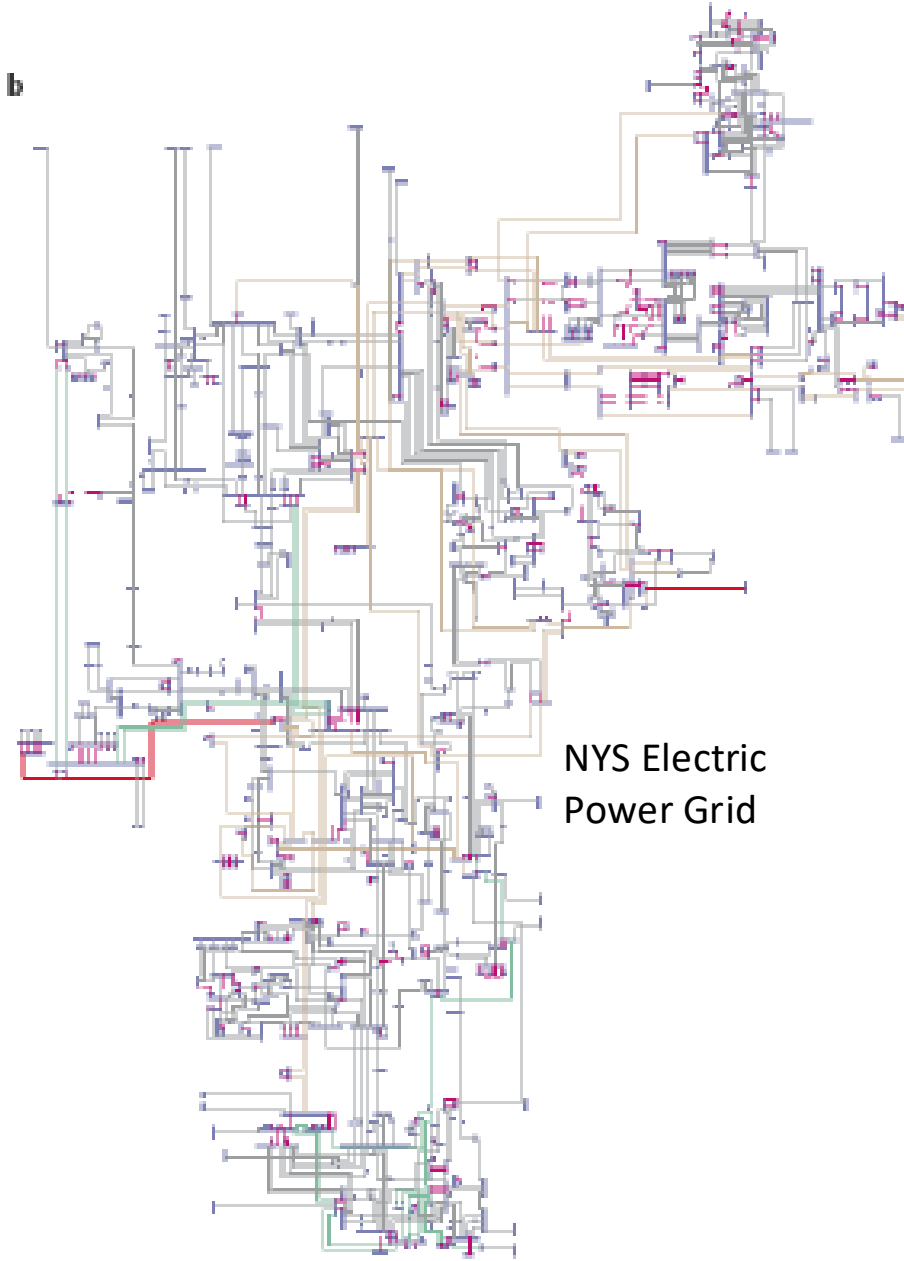
Geometric characteristics of real-world networks

Heather M. Guarnera
The College of Wooster



Why graph networks?

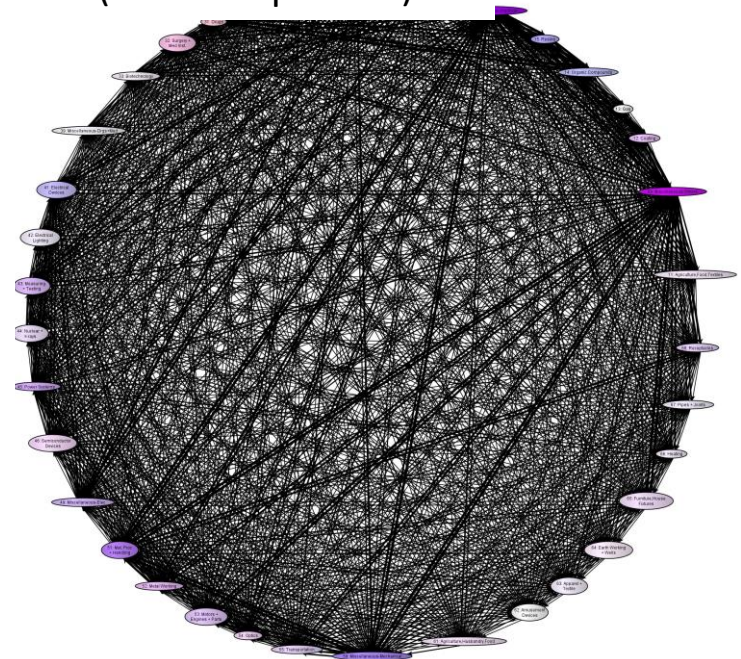
Graphs are everywhere



Graphs are everywhere



Utility Patent network
1972-1999
(3 Million patents)

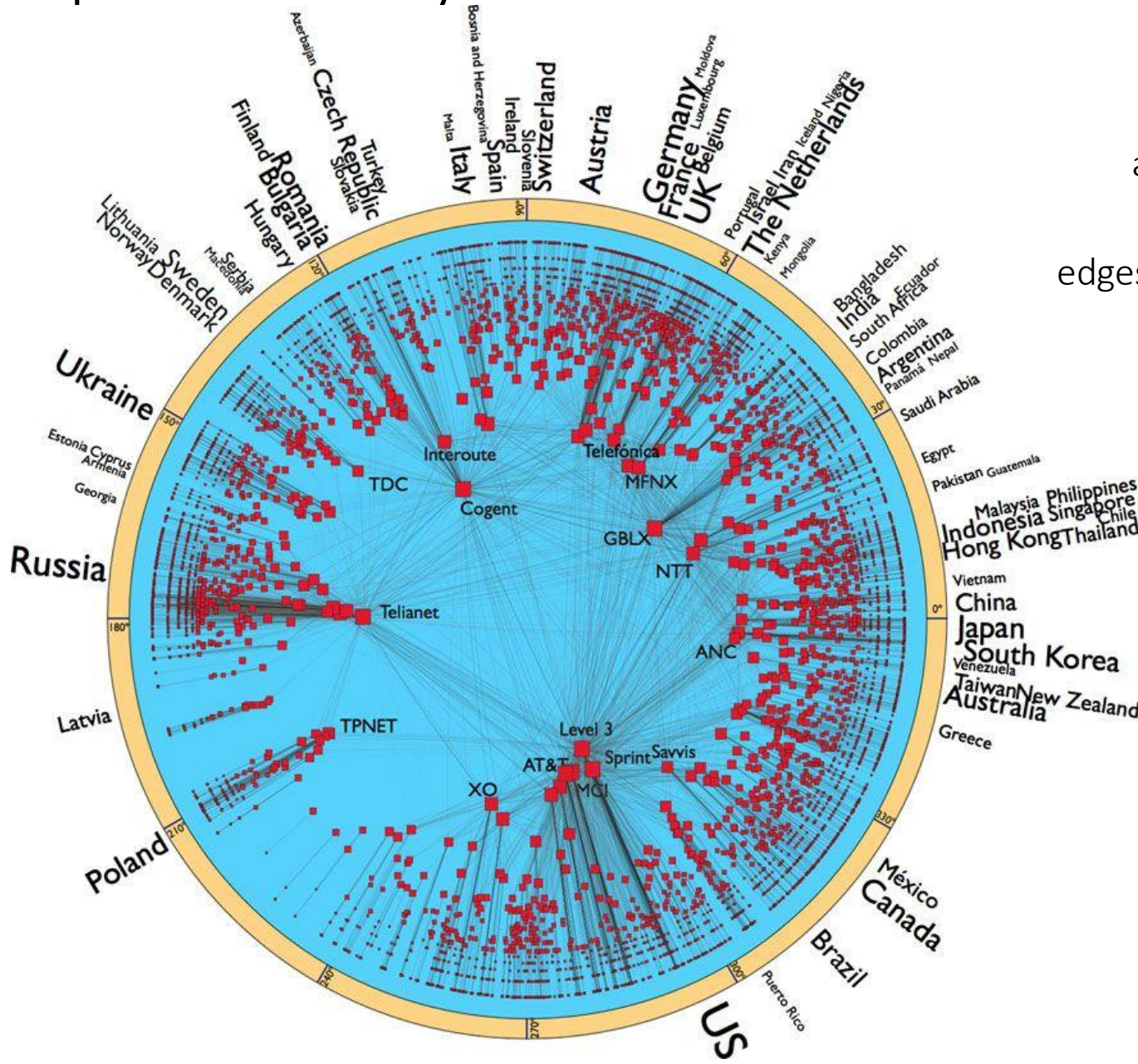


Graphs are everywhere

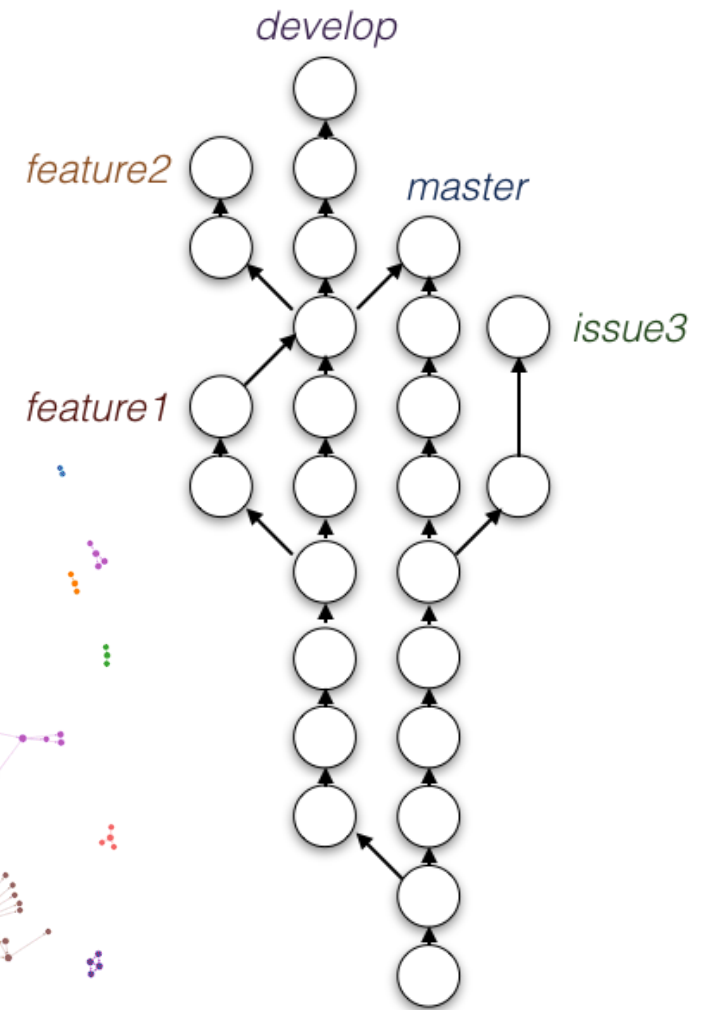
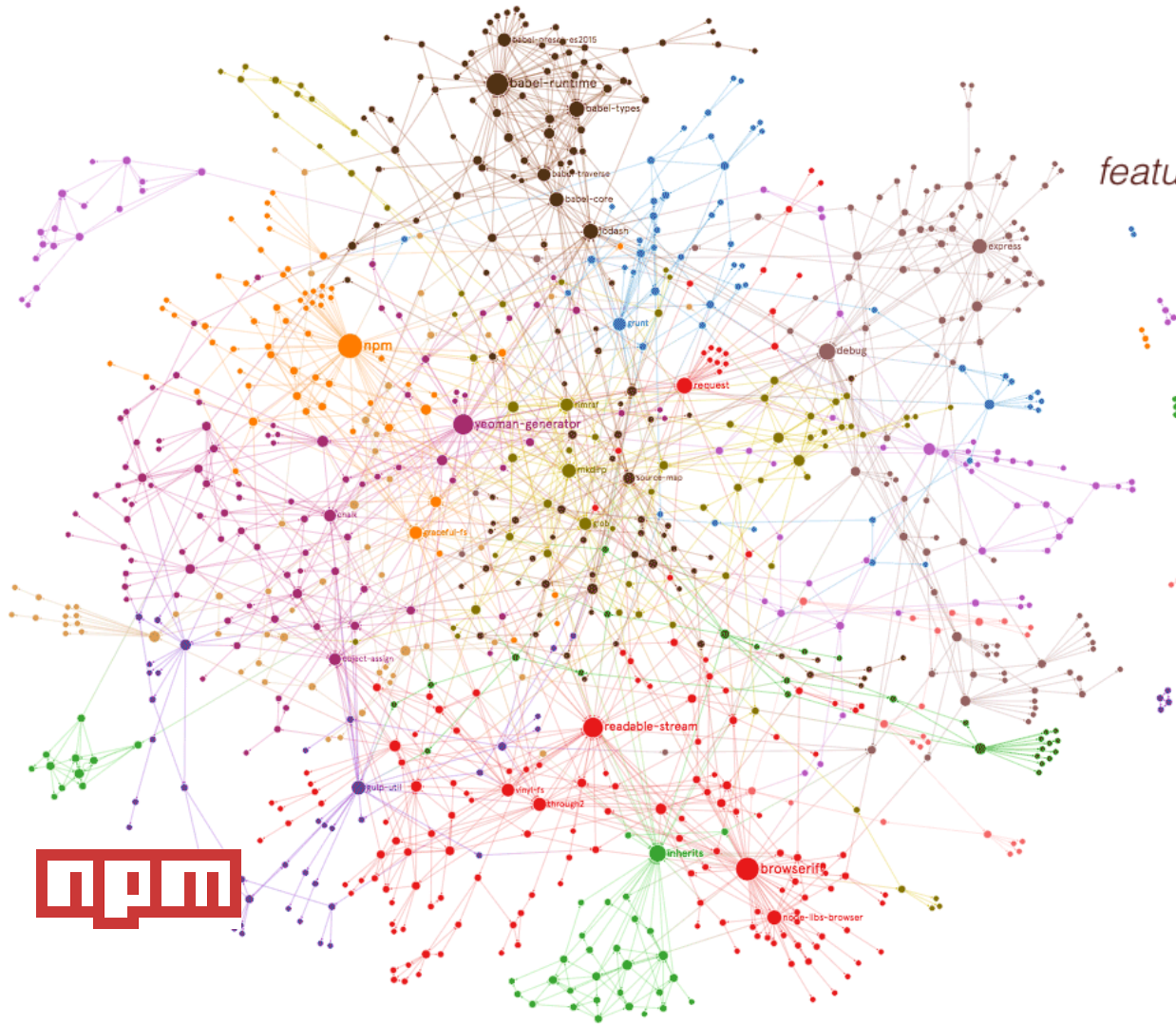
Internet (AS-level)

nodes $n = 23,752$
autonomous systems

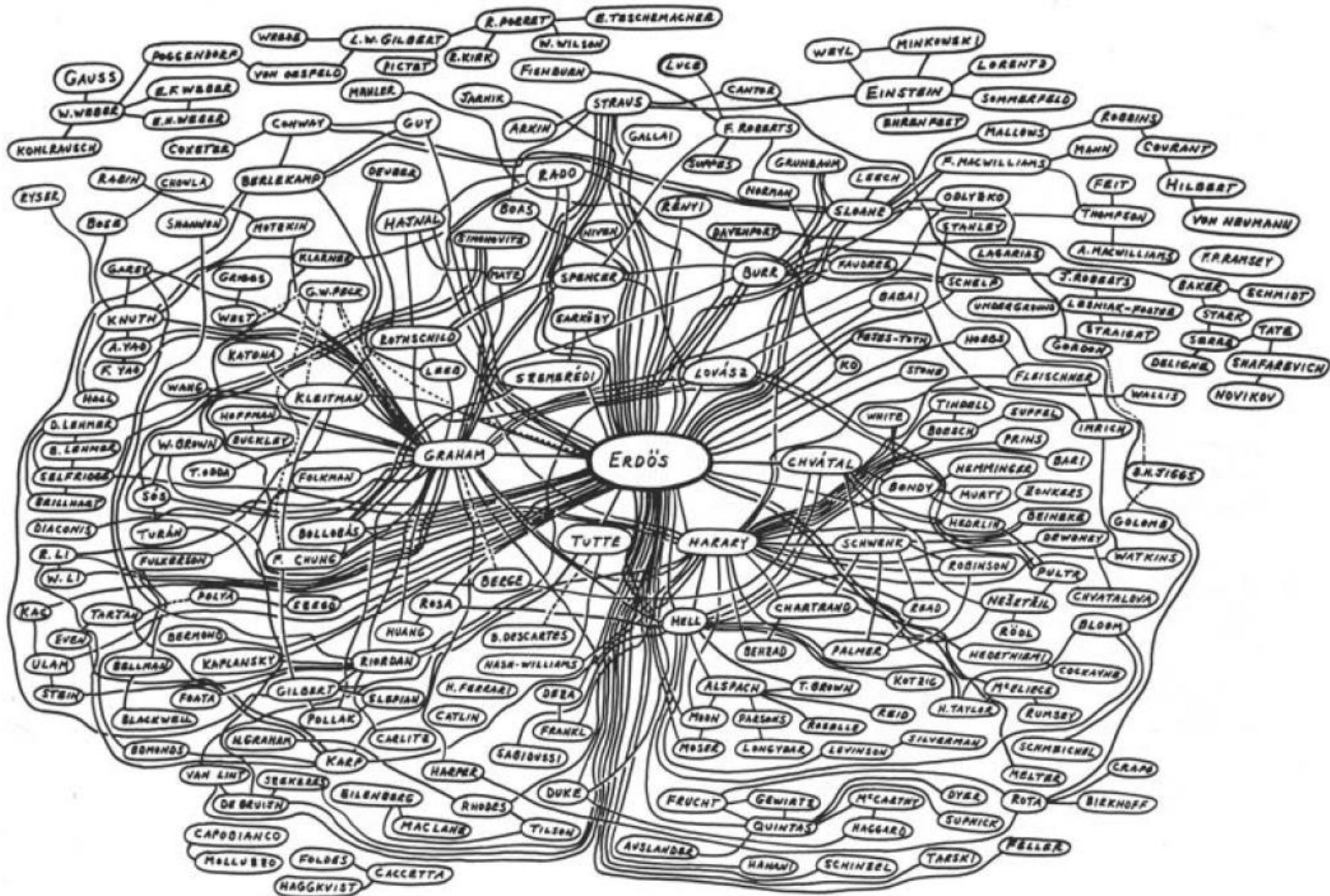
edges $m = 58,416$ AS links



Graphs are everywhere



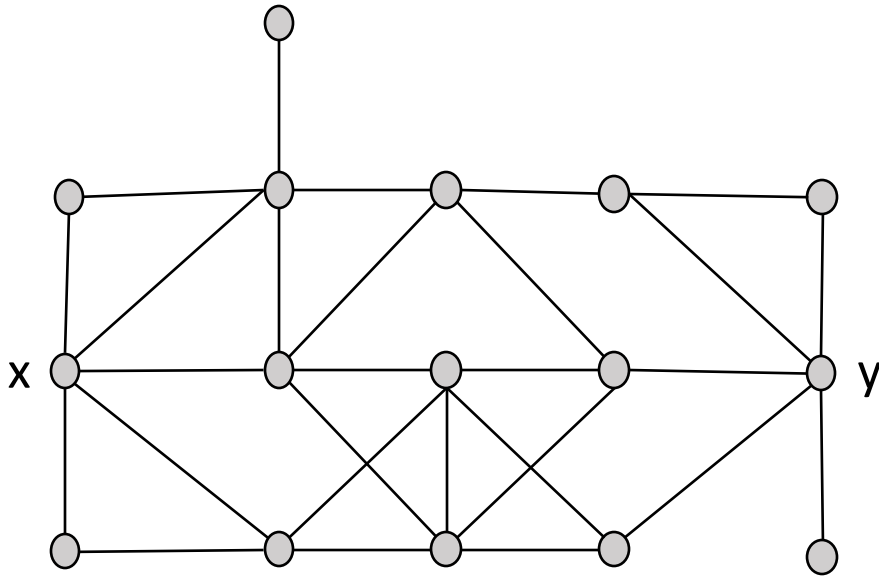
Graphs are everywhere



What is Fellow Travelers
Phenomenon?

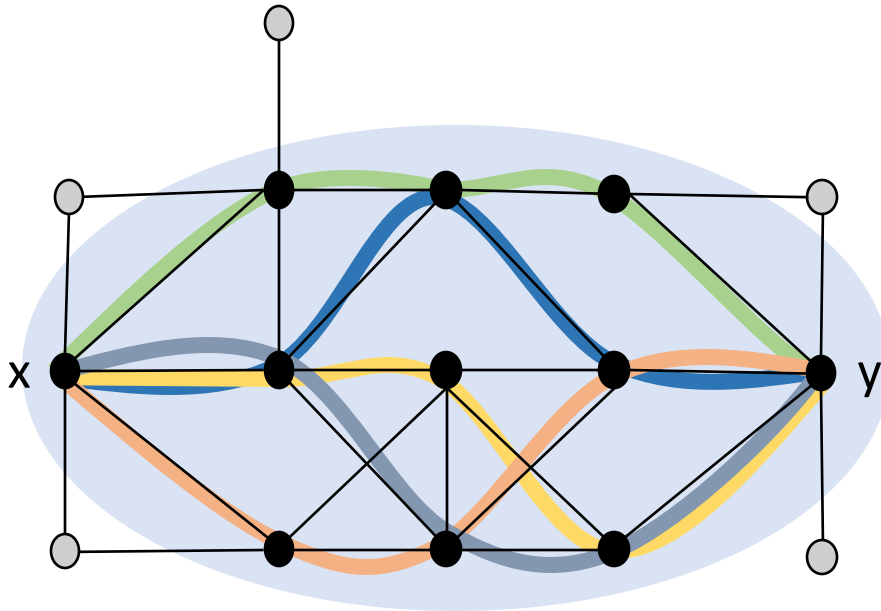
(Interval) Thinness of graphs

For any two x, y vertices on a graph $I(x, y) = \{z \in V : d(x, y) = d(x, z) + d(z, y)\}$ denotes the (metric) **interval**, i.e., **all vertices that lay on a shortest path between x and y .**



(Interval) Thinness of graphs

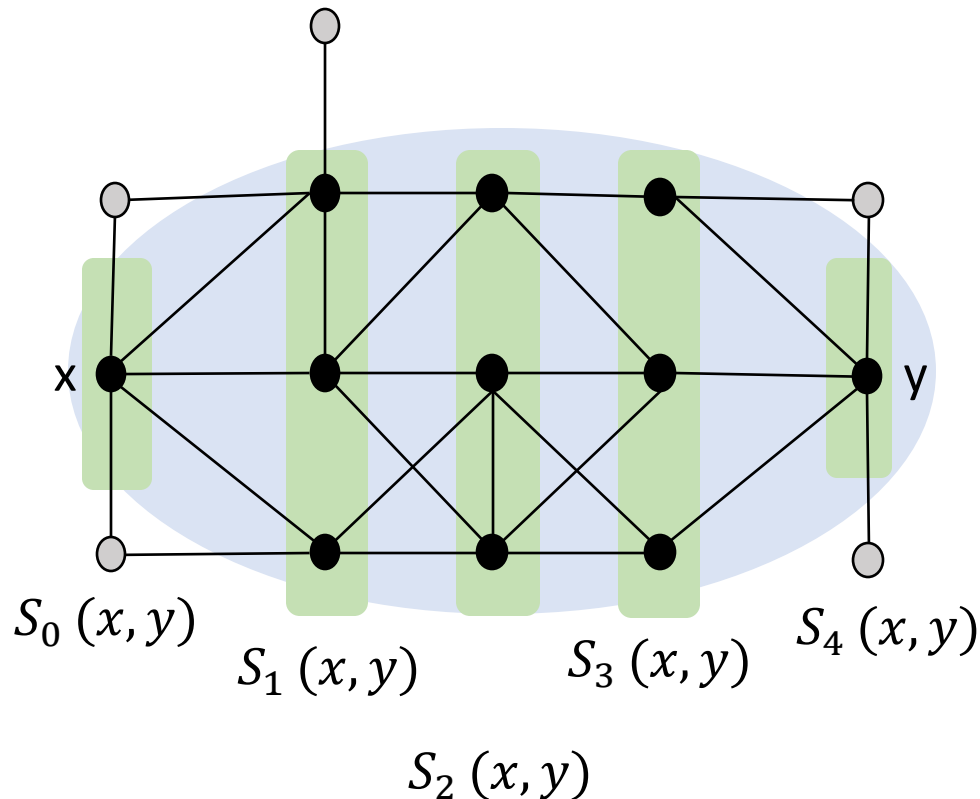
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The set $S_p(x, y) = \{z \in I(x, y) : d(z, x) = p\}$ is called a **slice** of the interval from x to y .

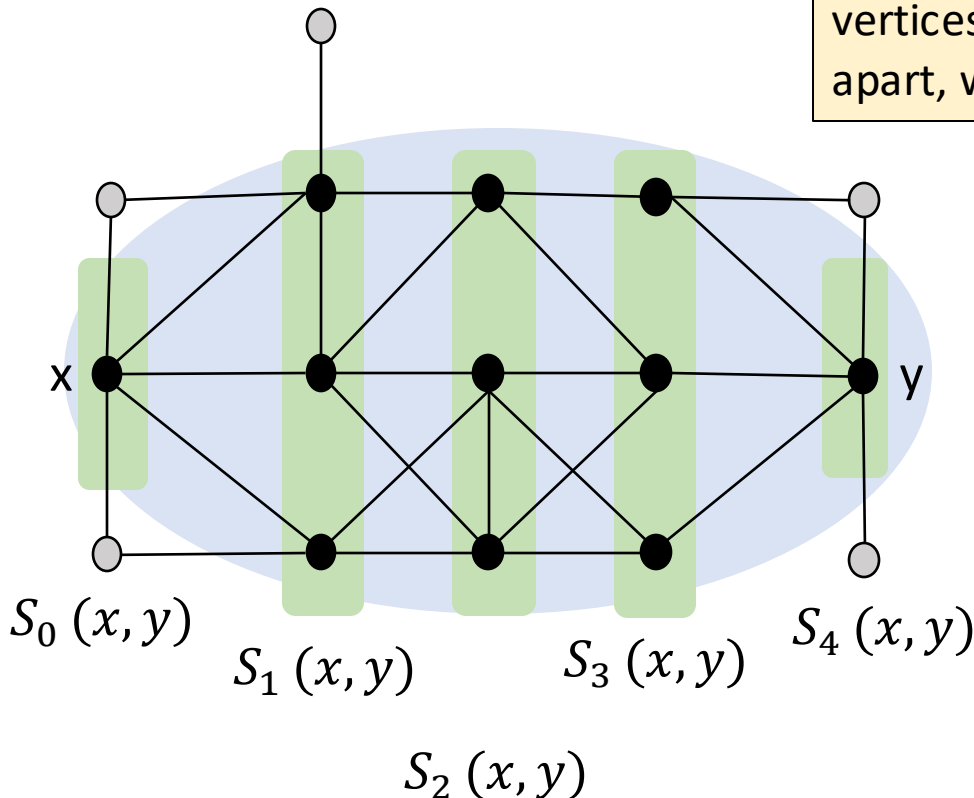


(Interval) Thinness of graphs

For any two x, y vertices on a graph $I(x, y) = \{z \in V : d(x, y) = d(x, z) + d(z, y)\}$ denotes the (metric) **interval**, i.e., **all vertices that lay on a shortest path between x and y** .

The set $S_p(x, y) = \{z \in I(x, y) : d(z, x) = p\}$ is called a **slice** of the interval from x to y .

An interval $I(x, y)$ is said to be **κ -thin** if any two vertices u, v of the slice $S_p(x, y)$ are at most κ apart, where integer p satisfies $0 \leq p \leq d(x, y)$.



Ex: $I(x, y)$ is 2-thin.

The smallest value κ for which all intervals of G are κ -thin is the **thinness of the graph**, denoted $\kappa(G)$.

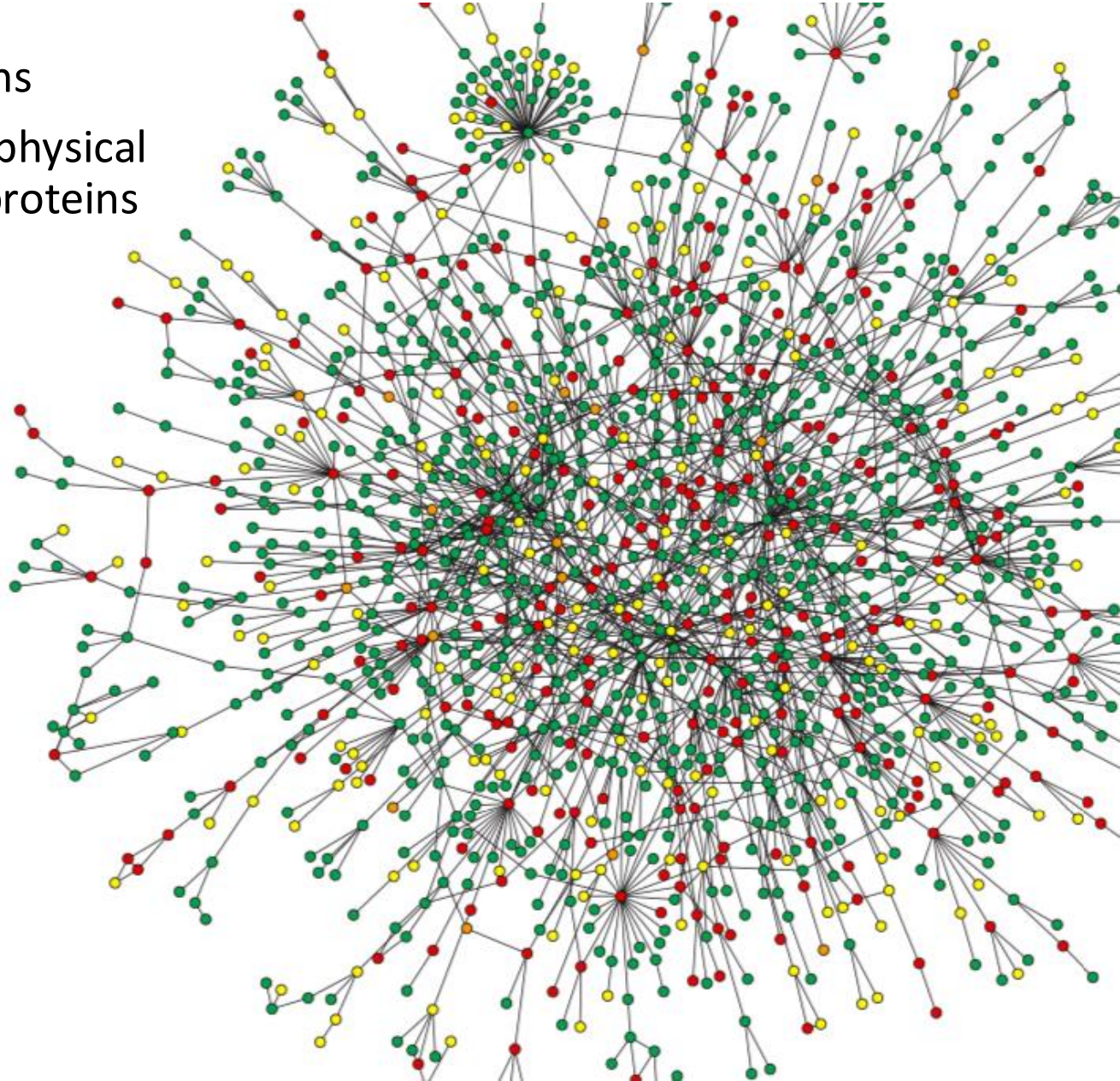
$\kappa(G)$ is a small constant in many real-world networks!

Ex: Protein Interaction Network

nodes $n = 1,870$ proteins

edges $m = 2240$ direct physical
interactions between proteins

$$\kappa(G) \leq 7$$



Ex: Other real-world networks with small thinness



- **Social networks** (subset of Facebook)
 - nodes $n = 293,501$ users
 - edges $m = 5,589,802$ friendships between users

$$\kappa(G) \leq 7$$

- **Web networks** (from Google)
 - nodes $n = 855,802$ websites
 - edges $m = 4,291,352$ hyperlinks connecting sites

$$\kappa(G) \leq 4$$



- **Peer-to-peer networks** (Gnutella)
 - nodes $n = 62,561$ hosts
 - edges $m = 147,878$ connections between hosts

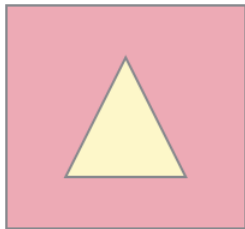
$$\kappa(G) \leq 5$$

Fellow travelers phenomenon is attributed to the negative curvature of the graph

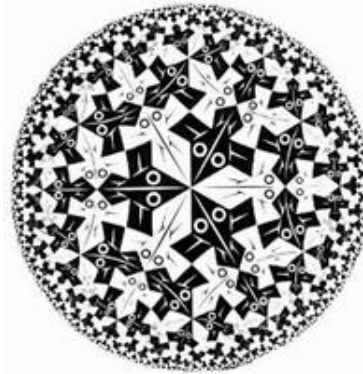
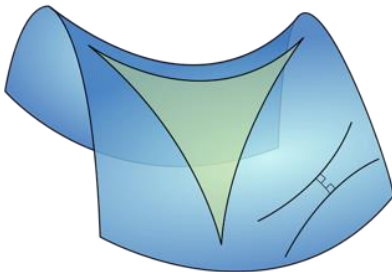
Geometric characteristics of real-world networks

- Surge of recent empirical and theoretical work analyzes geometric characteristics
- One important property: **negative curvature**
 - causes traffic between vertices to pass through a relatively small core of the network – as if the shortest paths between them were curved inwards
 - measured in many different (**somewhat equivalent**) ways

Zero Curvature



Negative Curvature



Geometric characteristics of real-world networks

- Surge of recent empirical and theoretical work analyzes geometric characteristics
- One important property: **negative curvature**
 - causes traffic between vertices to pass through a relatively small core of the network – as if the shortest paths between them were curved inwards
 - measured in many different (**somewhat equivalent**) ways
- Measures of negative curvature
 - κ Interval thinness
 - τ Geodesic triangle thinness
 - δ **Gromov Hyperbolicity**
 - ς Slimness
 - ι Rooted Insize

How can this geometric
information be applied?

Parameterized complexity/approximation factor

- **Goal:** create algorithms which solve problems utilizing these geometric properties
- Example: Consider δ hyperbolicity, which is known to be small in many real-world networks.
 - Solve a problem in $O(f(\delta) m)$ time
 - Compute a $f(\delta)$ approximation
- Some problems this has been applied to:
 - Covering/packing problems
 - Computing the diameter/radius
 - Facility location problems
 - Network analysis
 - Vertex pursuit games on graphs
 - Traveling salesman problem

Parameterized complexity/approximation factor

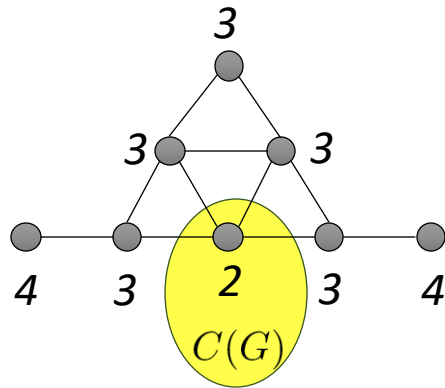
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1. F. Dragan and **H. Guarnera**. Helly-gap of a graph and vertex eccentricities. Theoretical Computer Science, 867:68-84, 2021.
2. F. Dragan and **H. Guarnera**. Eccentricity function in distance-hereditary graphs. Theoretical Computer Science, 833: 26-40, 2020.
3. F. Dragan and **H. Guarnera**. Eccentricity terrain of δ -hyperbolic graphs. Journal of Computer and System Sciences, 112: 50-56, 2020.
4. F. Dragan, G. Ducoffe, **H. Guarnera**. Fast deterministic algorithms for computing all eccentricities in (hyperbolic) Helly graphs, the 17th Algorithms and Data Structures Symposium (WADS'21), 2021.
5. Mohammed, F. Dragan, **H. Guarnera**. Fellow Travelers Phenomenon Present in Real-World Networks, Complex Networks & Their Applications, 2022.

Example: eccentricity function and centers

The **eccentricity** $e(x)$ of a vertex x is the distance to a furthest u vertex to x

$$e(x) = \max_{u \in V} d(x, u)$$



The minimum and maximum eccentricities are called the **radius** $rad(G)$ and **diameter** $diam(G)$ of the graph, respectively

The **center** of a graph $C(G)$ is the set of vertices with minimum eccentricity

$$C(G) = \{v \in V : e(v) = rad(G)\}$$

Applications:

- Measure the importance of a node (centrality indices)
- Facility location problems
- Detecting small-world networks (degrees of freedom)

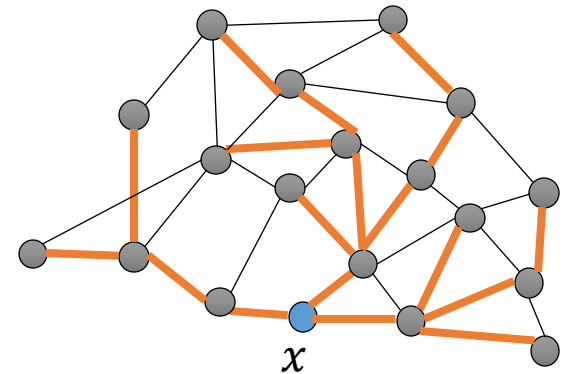
Computing vertex eccentricities straightforwardly.

The **eccentricity** $e(x)$ of a vertex x is the distance to a furthest u vertex to x

$$e(x) = \max_{u \in V} d(x, u)$$

Take a connected graph with n vertices and m edges.

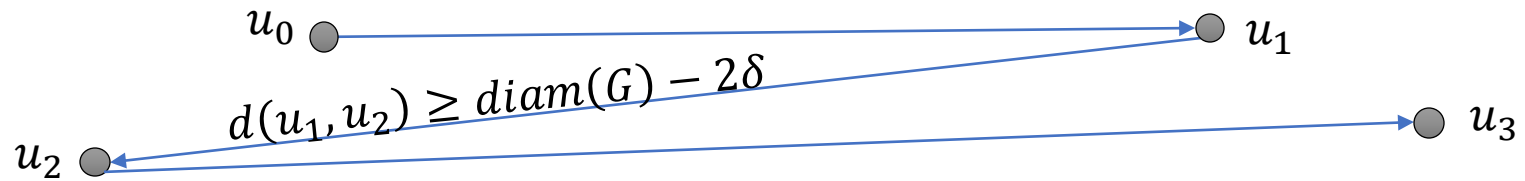
- A single Breadth-First Search (BFS) from a vertex x
 - runs in $O(m)$ time
 - yields $e(x)$
- Call BFS for each of the n vertices
- Total $O(nm)$ runtime



This is prohibitively expensive on many real-world networks, as they are huge!

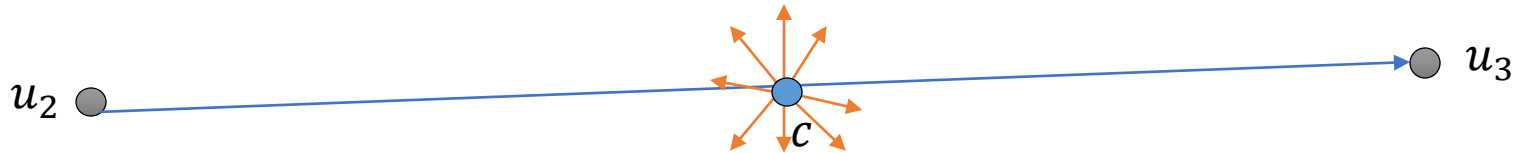
Efficient eccentricity approximation via eccentricity approximating spanning tree

- Find a long path in $O(m)$ time



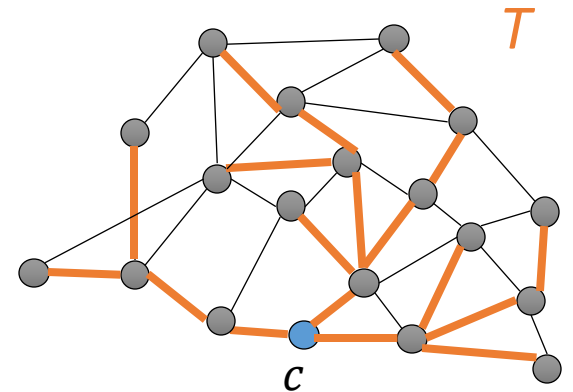
Efficient eccentricity approximation via eccentricity approximating spanning tree

- Find a long path in $O(m)$ time



- Run breadth-first search (BFS) from the middle vertex c between u_2u_3
- We show $e_T(v) \leq e_G(v) \leq e_T(v) + 6\delta$

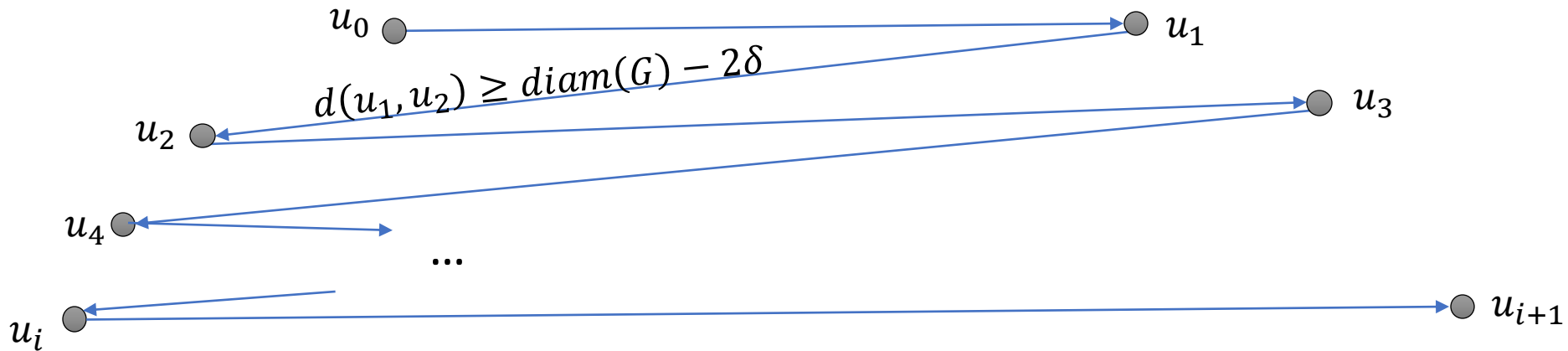
Theorem [2]: There is a 6δ approximation of all eccentricities in total $O(m)$ time



Efficient eccentricity approximation via eccentricity approximating spanning tree

- Find a **mutually distant** pair of vertices x, y in $O(\delta m)$ time

$$e(x) = d(x, y) = e(y)$$

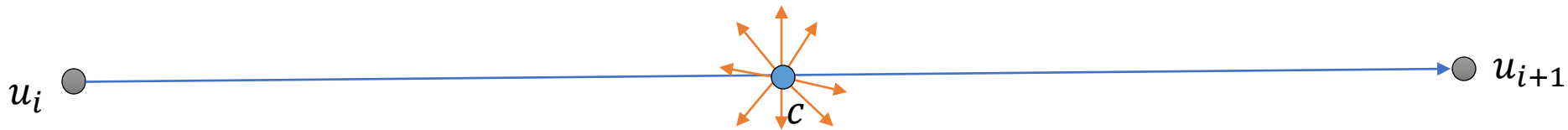


Efficient eccentricity approximation via eccentricity approximating spanning tree

- Find a **mutually distant** pair of vertices x, y in $O(\delta m)$ time

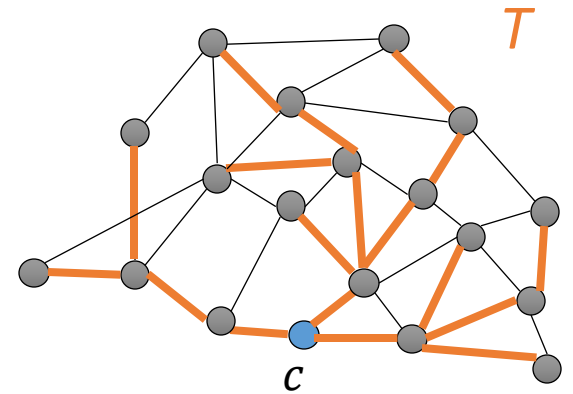
$$e(x) = d(x, y) = e(y)$$

- Run breadth-first search (BFS) from the middle vertex c of the mutually distant pair



- We show $e_T(v) \leq e_G(v) \leq e_T(v) + 4\delta$

Theorem [2]: There is a 4δ approximation of all eccentricities in total $O(\delta m)$ time



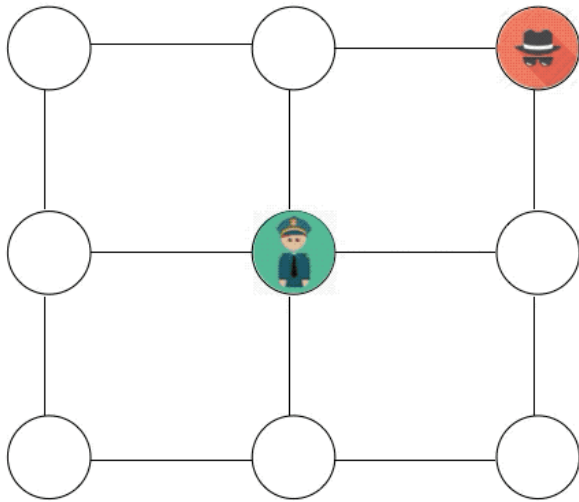
Conclusion

- Many real world networks exhibit the fellow travelers property
 - Biological networks
 - Communication networks
 - Social networks
 - Software ecosystems
- We can take advantage of this nice geometric property to solve problems faster on these networks
 - Ex: computing vertex eccentricities

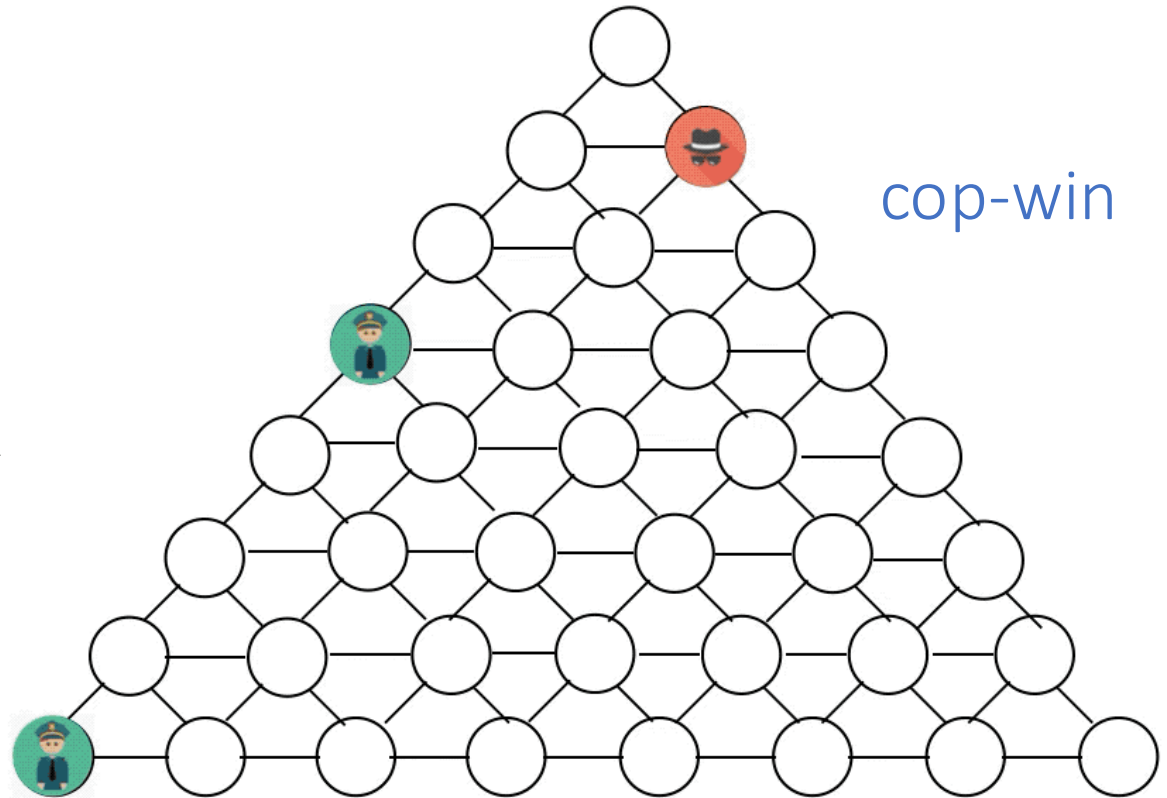
Conclusion and future work

- Many real world networks exhibit the fellow travelers property
 - Biological networks
 - Communication networks
 - Social networks
 - Software ecosystems
 - **What else?**
- We can take advantage of this nice geometric property to solve problems faster on these networks
 - Ex: computing vertex eccentricities
 - **What else? Ex: vertex pursuit games**
- **Routes:**
 - **Theoretical**
 - **Applied**

Games on graphs: cops vs. robbers



robber-win



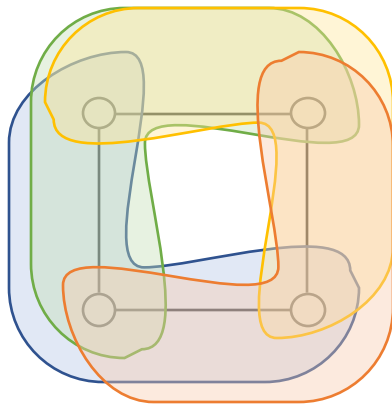
cop-win

Paths for Future Work (Theoretical)

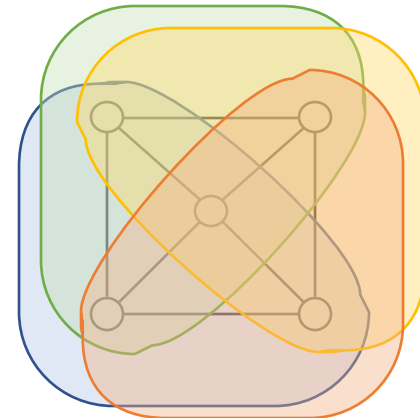
Every graph G can be isometrically embedded into the smallest Helly graph $\mathcal{H}(G)$.

A family F of sets has the **Helly property** if for every subfamily S of F the following hold: if the elements of S pairwise intersect, then the intersection of all elements of S is also non-empty.

A graph is called **Helly** if its family of disks satisfies the Helly property.



Not Helly



Helly

Paths for Future Work (Theoretical)

Every graph G can be isometrically embedded into the smallest Helly graph $\mathcal{H}(G)$.

$\mathcal{H}(G)$ is called the **injective hull** of G [Isbell 1964, Dress 1984].

- $\mathcal{H}(G)$ preserves hyperbolicity
- If G is δ -hyperbolic, then any vertex of $\mathcal{H}(G)$ is within 2δ to a vertex of G [Lang 2013]
- **[1]** Any vertex of $\mathcal{H}(G)$ is within $\alpha(G)$ to a vertex of G , where $\alpha(G)$ is the Helly-gap.

This motivates finding solutions to problems in $\mathcal{H}(G)$ which can lead to approximate solutions in G .

Paths for Future Work (Theoretical)

$\mathcal{H}(G)$ can be constructed efficiently [6] for some graph classes (e.g., distance-hereditary graphs).

- It is computationally difficult to compute for other (even basic) graph classes [6]
- Investigate other graph classes (alpha-i metric graphs, chordal graphs, etc.)

The existence of $\mathcal{H}(G)$ is a powerful tool to gain insight into a graph class from various perspectives

- structurally what other classes are closed under Hellification?
- metrically what parameters are preserved?
- algorithmically what else can be solved?

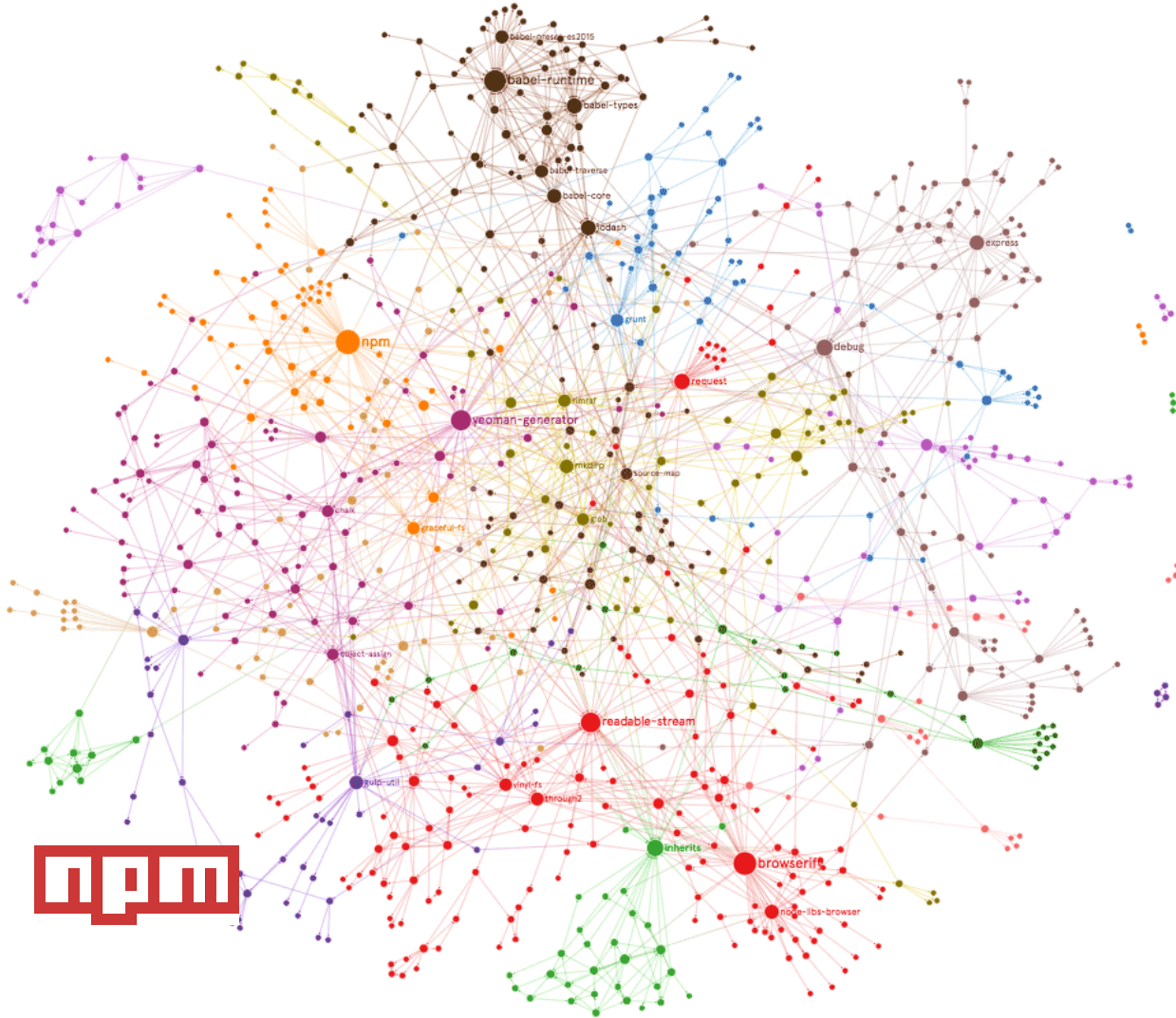
It lends itself nicely to approximation algorithms dependent on $\alpha(G)$.

- Apply those algorithms to specific graph classes

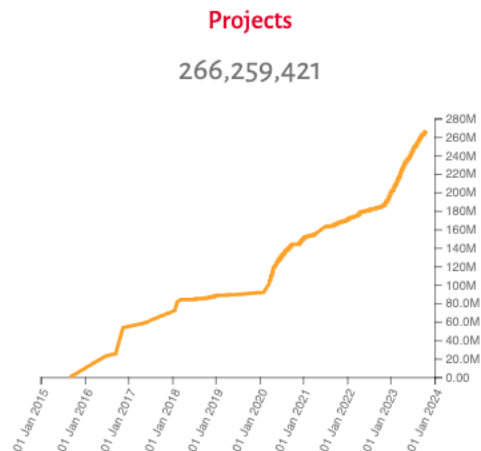
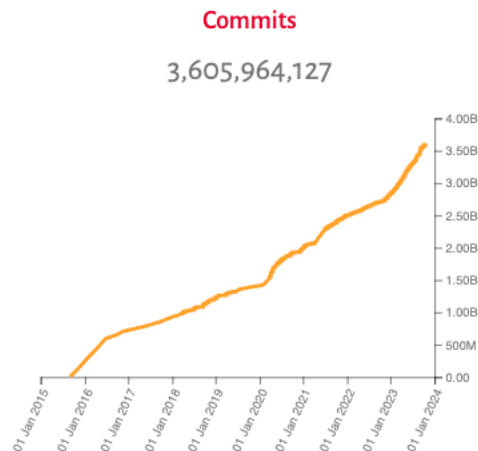
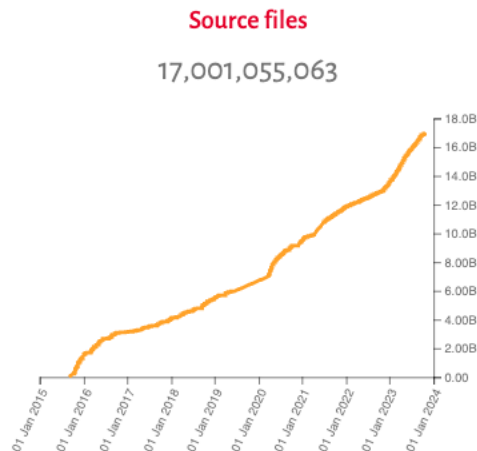
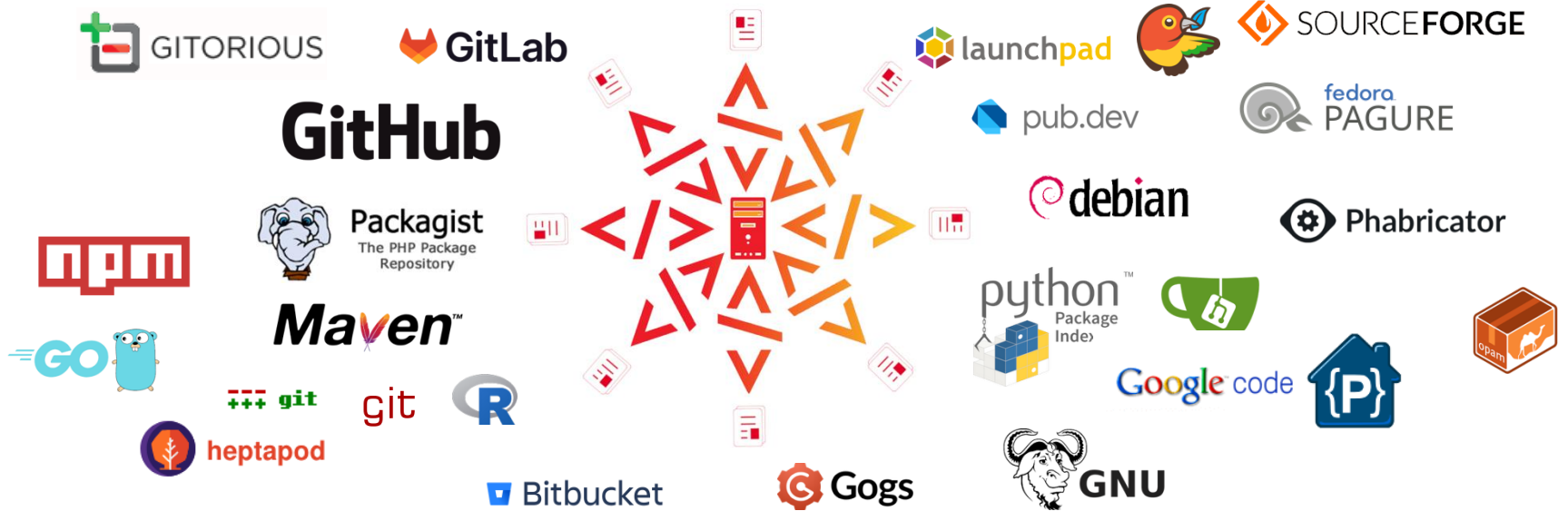
Paths for Future Work (Applied)

- Analyze existing real-world networks
 - Biological networks
 - Communication networks
 - Social networks
 - **Software ecosystems**
- Compute values of graph parameters on existing networks
 - Form conjectures for theoretical work
- Optimize algorithms to run efficiently on enormous networks
- Mining software repositories

Network analysis: software ecosystems



Software Heritage Project



Directories
13,682,190,925

Authors
66,817,476

Releases
77,841,991

Thank you! Questions?