

1-11-2019

Latent Space Models for Temporal Networks

Jasper Alt
Portland State University

Follow this and additional works at: https://pdxscholar.library.pdx.edu/systems_science_seminar_series



Part of the [Dynamic Systems Commons](#)

Let us know how access to this document benefits you.

Recommended Citation

Alt, Jasper, "Latent Space Models for Temporal Networks" (2019). *Systems Science Friday Noon Seminar Series*. 73.

https://pdxscholar.library.pdx.edu/systems_science_seminar_series/73

This Book is brought to you for free and open access. It has been accepted for inclusion in Systems Science Friday Noon Seminar Series by an authorized administrator of PDXScholar. Please contact us if we can make this document more accessible: pdxscholar@pdx.edu.

Latent Space Models for Temporal Networks

Jasper Alt

Department of Computer Science
Portland State University

References

- **This talk based on work (with Rajesh Venkatachalapathy) submitted to NetSci 2019.**
- **A fuller manuscript (with Rajesh) is in progress.**
- **Peres et. al. “Mobile Geometric Graphs” (2010)**
- **Holme, P. “Modern Temporal Network Theory: A Colloquium”**

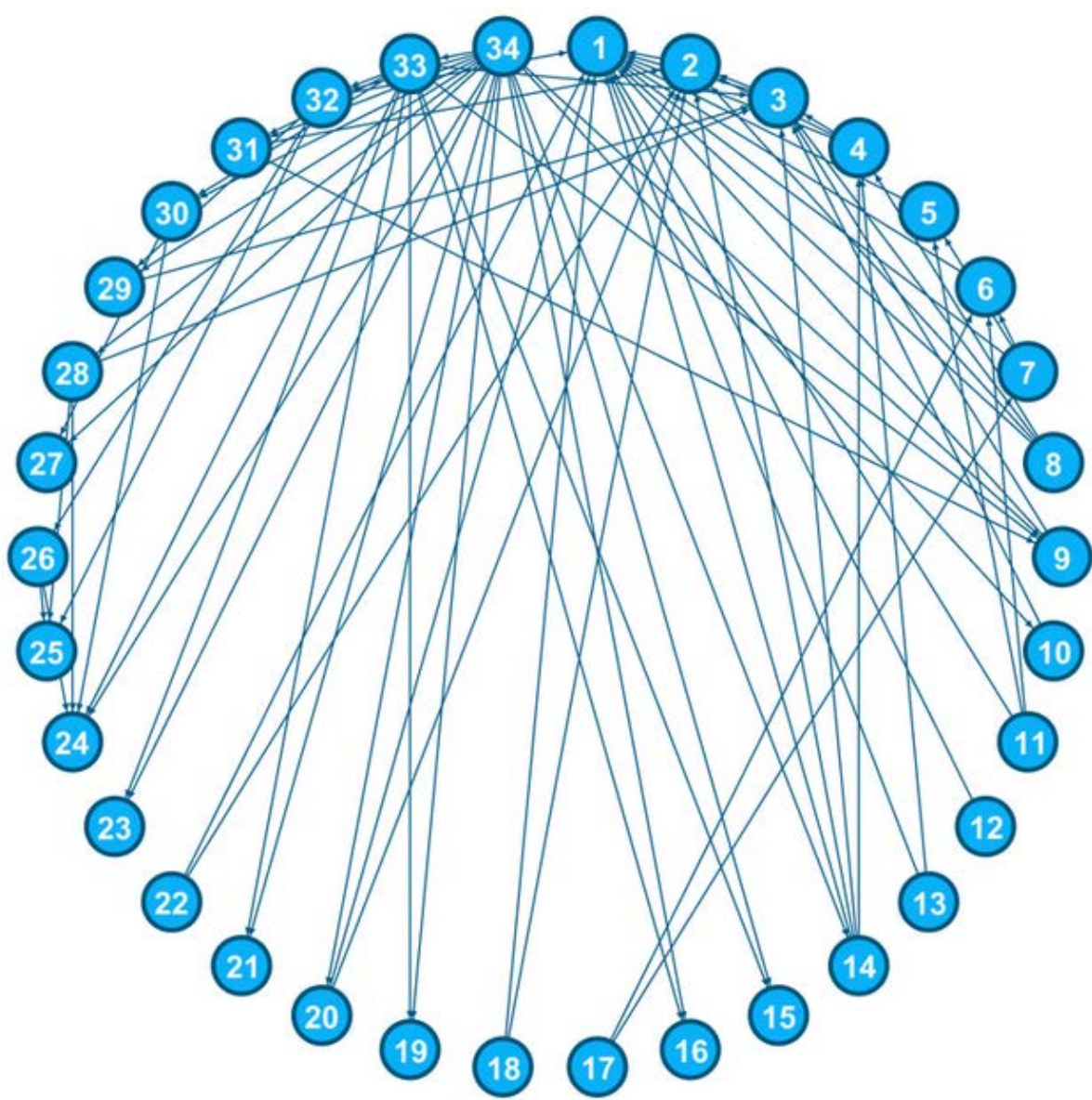
(2015)

Outline

- **Homophily and latent space**
- **Random geometric graphs and Brownian motion**
- **Mobile geometric graphs**
- **Implementation**
- **Observations**
- **Current and future work**

0. Social Networks

- **Maps of relationships between actors**
- **What do the actors tell us about the networks?**
- **What do the networks tell us about the actors?**



A social network (Wikipedia)

1. Homophily

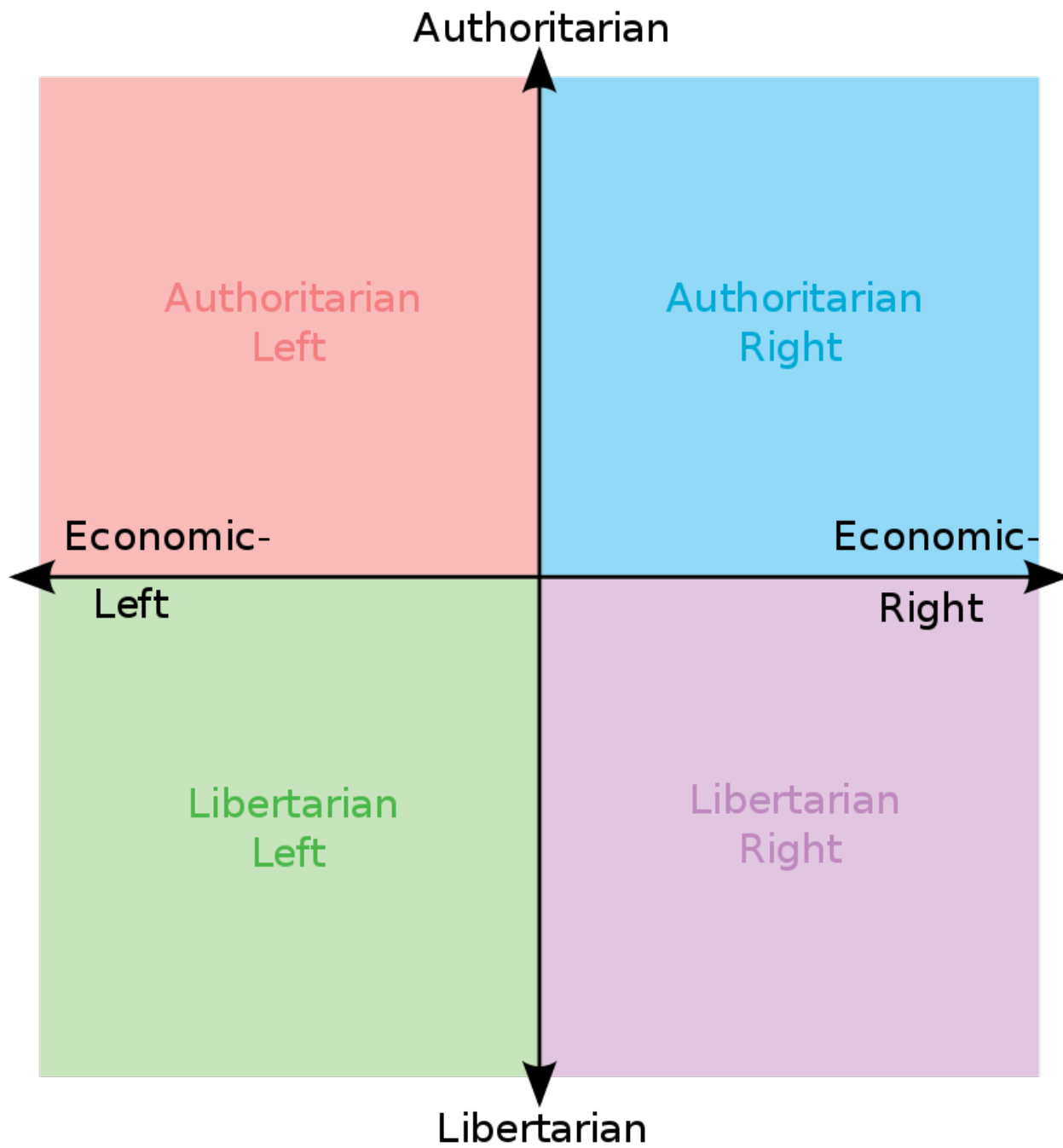
“Birds of a feather flock together”

- **Actors with common attributes are more likely to form ties**
- **Ties between actors may indicate common attributes**
- **Typically expressed in an informal way**

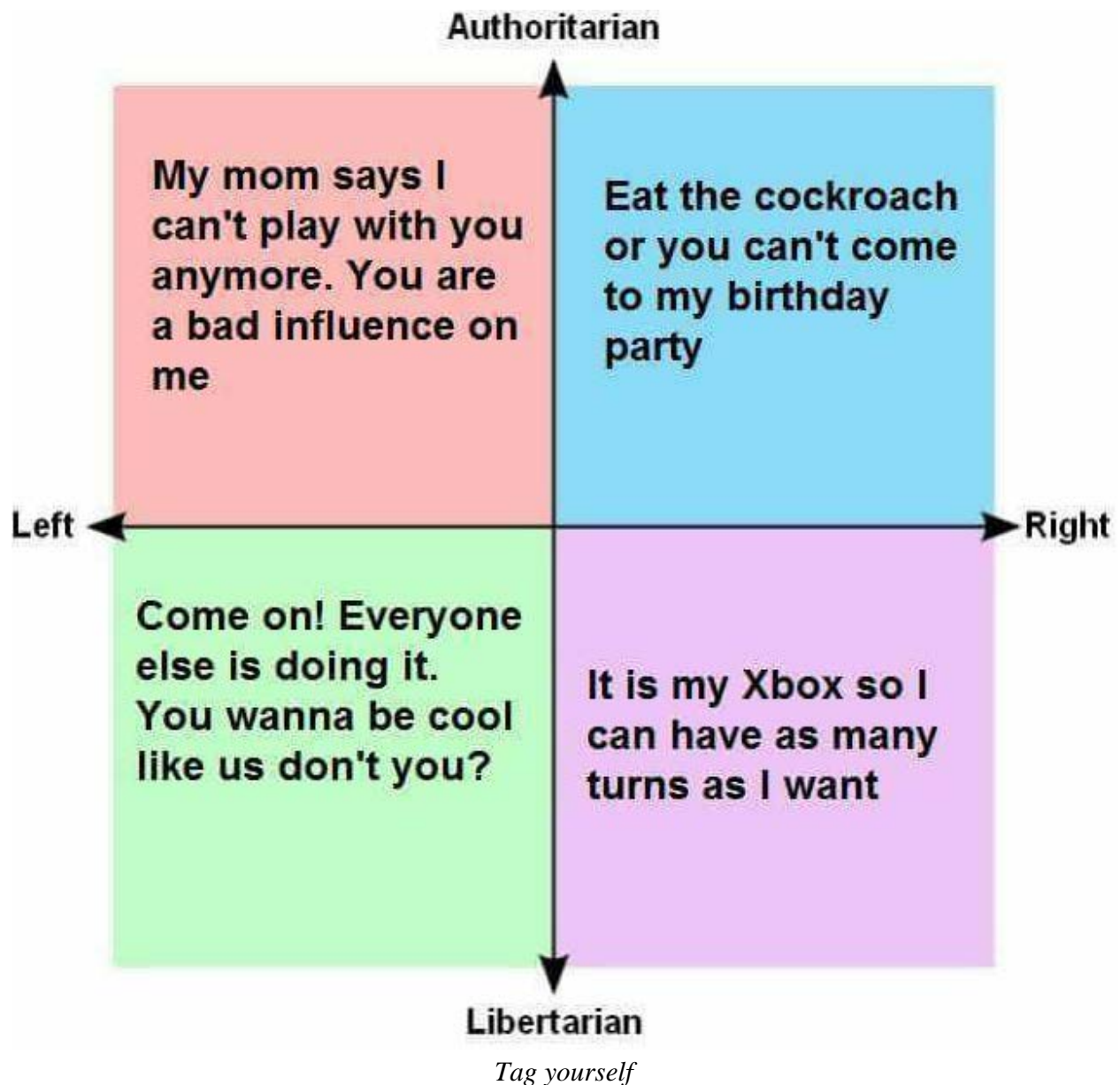
- **Latin prefix, Greek Suffix**
- **Idiophily?**

2. Latent Space

- **Can we state something more formal?**
- **Combinations & intensities of attributes assign the actors to coordinates in a *latent space***
- **More attributes in common, or attributes with similar intensities, mean the actors are closer together in this space.**
- **Tobler's Law: Everything is related to everything else, but near things are more related than distant things.**
- **(Q. Are there *any* networks that do *not* correspond to a latent space?)**



Wikipedia: The political compass



3. Homophily in Latent Space

The graph at right is described by an $n \times n$ adjacency matrix \mathbf{A} , where $n = 10$, whose entries

$$A_{ij} = \begin{cases} f(d_{ij}) & \text{if } f(d_{ij}) > m \\ 0 & \text{otherwise} \end{cases} \quad \text{where } f(d_{ij}) = \frac{1}{1 + e^{0.5(5 - d_{ij})}}, \quad \frac{1}{1 + e^{0.5(5 - d_{ij})}}, \quad 1 - \frac{d_{ij}^2}{2}, \quad 0$$

```

var X = get("X"); inc("Xi"); set("x", X[get("Xi")]); update("blurb_dij"); distanceFunction = distances[get("Xi")];
figure[1].load([0, matriculate(figure[2].data, get("m"))]); var X = get("X"); dec("Xi"); var Xi = get("Xi"); set("x",
X[get("Xi")]); update("blurb_dij"); distanceFunction = distances[get("Xi")]; figure[1].load([0, matriculate(figure[2].data,
get("m"))]); ,

```

\$blurb\$ [" d_{ij} \text{ is the } \href{http://en.wikipedia.org/wiki/Euclidean_distance}{Euclidean} \text{ distance between } i \text{ and } j", " d_{ij} \text{ is the } \href{http://en.wikipedia.org/wiki/Manhattan_distance}{Manhattan} \text{ distance between } i \text{ and } j", " d_{ij} \text{ is the } \href{http://en.wikipedia.org/wiki/Hamming_distance}{Hamming} \text{ distance between } i \text{ and } j", " $x_i y_i$ \text{ and } x_j y_j",] get("blurbs")[get("Xi")] set("blurb", get("blurbs")[get("Xi")]); , and $m = 0.35$ inc("m"); figure[1].load([0, matriculate(figure[2].data, get("m"))]); dec("m"); figure[1].load([0, matriculate(figure[2].data, get("m"))]); is an arbitrary cutoff distance. If the edge weights are interpreted as probabilities of connections, then the matrix describes a distribution of possible graphs reflecting the same underlying social space.

Click and drag the points to change the graph.

4. Latent Space Dynamics

- **We know attributes can change.**
- **This means the actors' positions in latent space change.**
- **There are *many* possible ways this could happen.**
- **In social contexts, we expect nontrivial dynamics.**

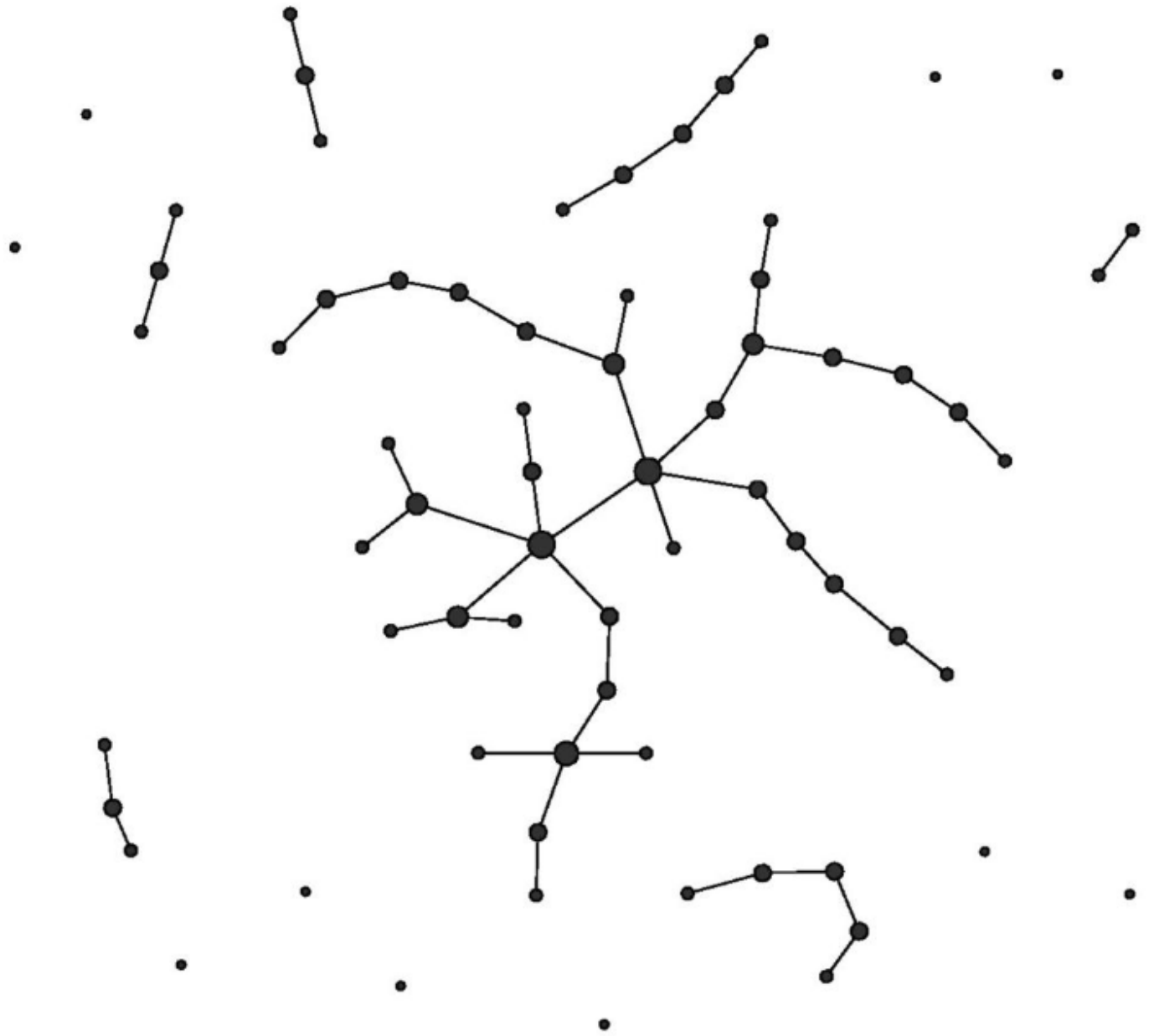
- **People influence each other and are influenced by social forces.**

Paranormal Beliefs (2016-2018)	2016	2017	2018	Change 2016-2018
Ancient, advanced civilizations, such as Atlantis, once existed	39.6	55.0	56.9%	17.3%
Aliens have visited Earth in our ancient past	27.0	35.0	41.4%	14.4%
Places can be haunted by spirits	46.6	52.3	57.7%	11.1%
Aliens have come to Earth in modern times	24.7	26.2	35.1%	10.4%
Bigfoot is a real creature	13.5	16.2	20.7%	7.2%
Some people can move objects with their minds	19.1	25.0	26.2%	7.1%
Fortune tellers and psychics can foresee the future	14.1	19.4	17.2%	3.1%

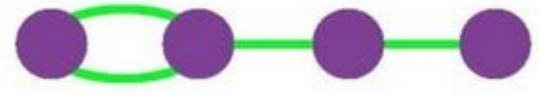
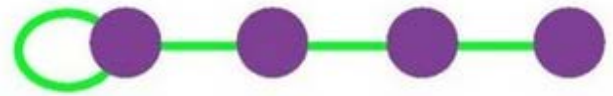
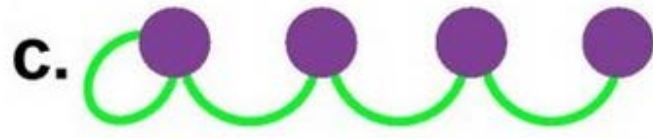
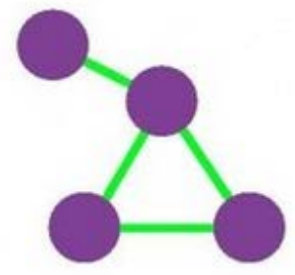
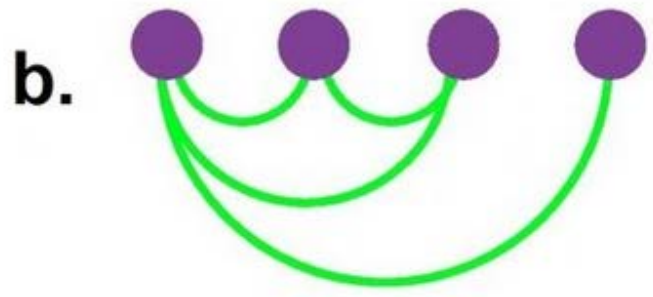
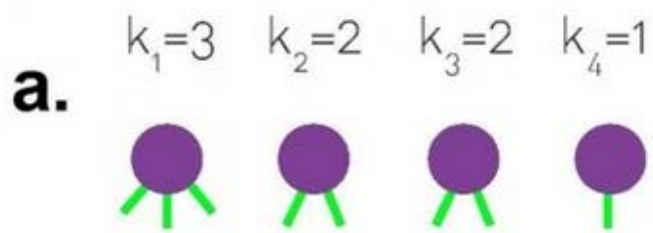
Chapman University Survey of American Fears

5. Null Models

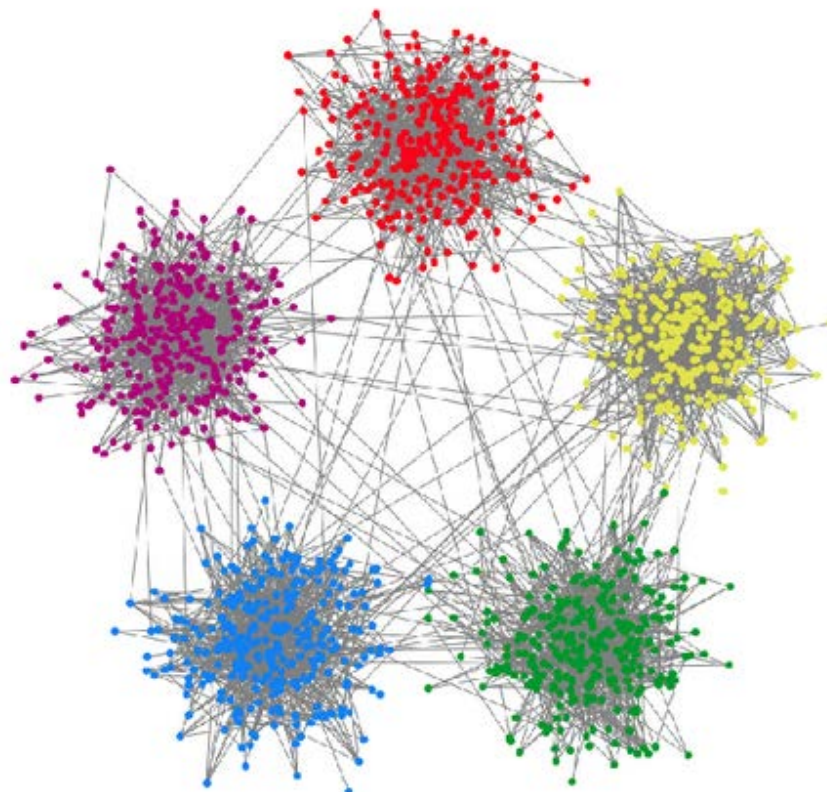
- **Network science has some big null models.**
- **Erdos-Renyi — random graphs**
- **Configuration models — random with fixed degree sequence**
- **Stochastic block models — random with some community structure**



Wikipedia: Erdos-Renyi model



Wikipedia: Configuration model



Emmanuel Abbe: Stochastic block model

- **Good null models show what can be explained by chance.**
- **Random dynamics are useful for testing non-random alternatives.**
- **A good null model for latent space network dynamics will have random dynamics.**

- **This will make it easier to spot networks that don't.**

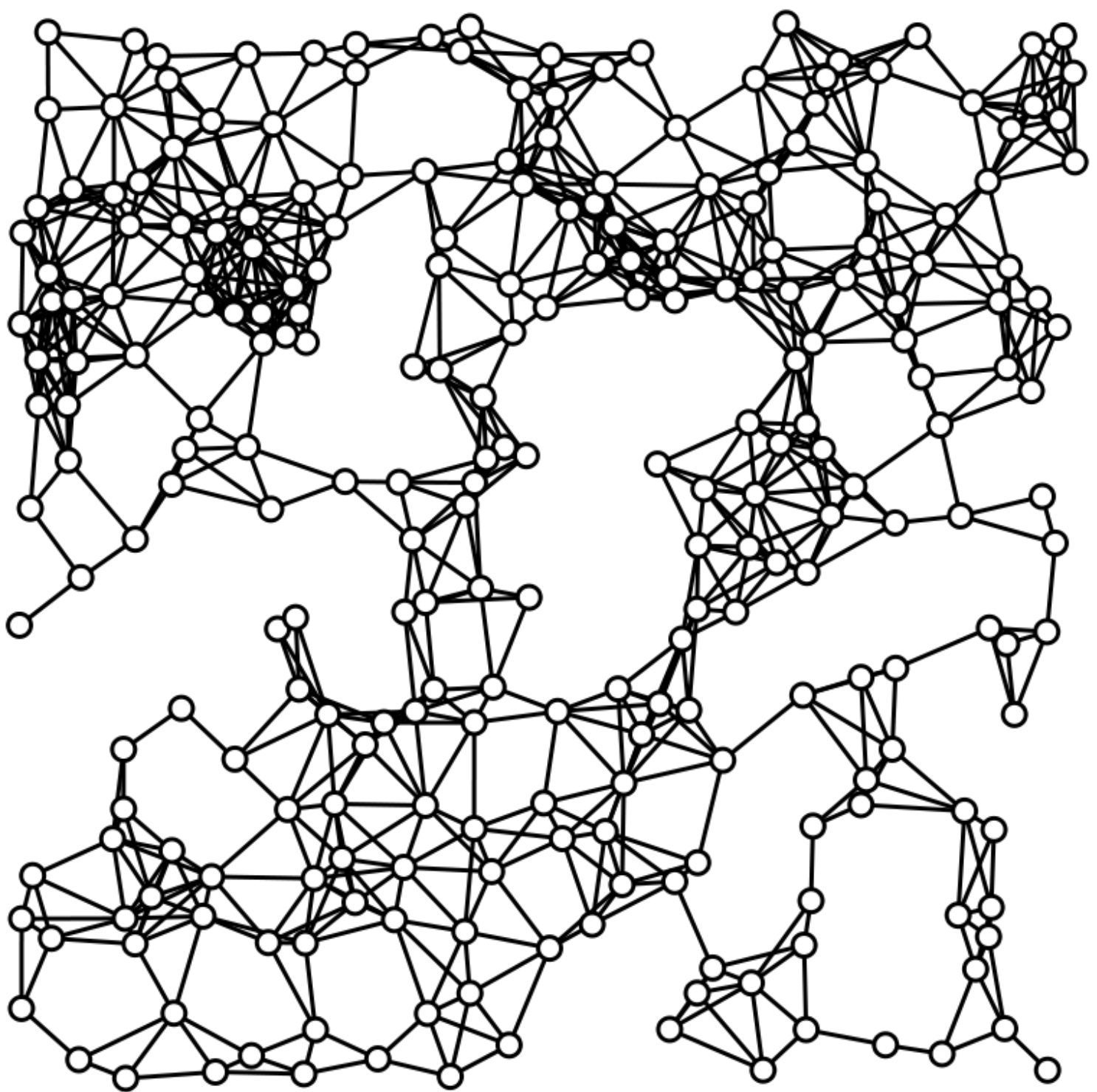
6. Temporal Networks

- **We typically understand static networks to be ‘snapshots’ of dynamic networks.**
- **Real networks are *temporal*.**
- **Static networks can be instantaneous snapshots or aggregate over a window of time.**
- **What is a null model for temporal networks?**
- **Random, but time dependent**

Building Blocks

6. Random Geometric Graphs

- **Graphs**
- **Geometric**
- **Random**



Wikipedia: Random geometric graph

- **Nodes are placed using some *point process***
- **(say, uniformly at random)**
- **Nodes within some radius of one another are connected.**
- **Studied in the context of wireless networks.**

7. Brownian Motion

- **Move a normally distributed distance in either direction.**
- **Like a particle in a fluid.**
- **Shows up in physics, statistics, quantitative finance, etc.**
- **Preserves the intensity of point processes.**

8. Mobile Geometric Graphs

- **Random geometric graphs with Brownian motion**
- **Brownian motion adds time dependence while preserving randomness**
- **Very little studied**
- **Existing work focuses on the mathematical statistics of their communication properties.**

Contd

