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networks

Random walks on

Random walks and diffusion on networks

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Outline

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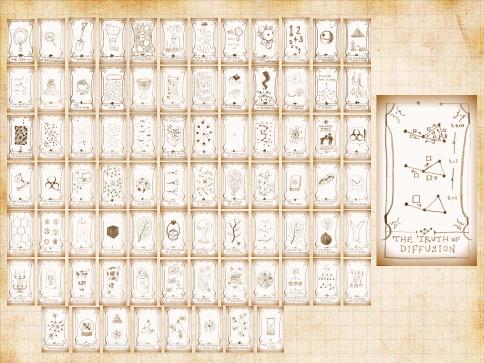
Random walks on networks

Random walks on networks





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Random walks on networks—basics:

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- Imagine a single random walker moving around on a network.
- At t = 0, start walker at node j and take time to be discrete.
- Q: What's the long term probability distribution for where the walker will be?
- Solution Define $p_i(t)$ as the probability that at time step t, our walker is at node i.
- \Im We want to characterize the evolution of $\vec{p}(t)$.
- Solution First task: connect $\vec{p}(t+1)$ to $\vec{p}(t)$.
- 🚳 Let's call our walker Barry.
- Unfortunately for Barry, he lives on a high dimensional graph and is far from home.
- line still: Barry is texting.





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Where is Barry?

Consider simple undirected, ergodic (strongly connected) networks.

As usual, represent network by adjacency matrix A where

> $a_{ij} = 1$ if *i* has an edge leading to *j*, $a_{ij} = 0$ otherwise.

Barry is at node *j* at time *t* with probability *p_j(t)*.
In the next time step, he randomly lurches toward one of *j*'s neighbors.
Barry arrives at node *i* from node *j* with

probability $\frac{1}{k_i}$ if an edge connects *j* to *i*.

🚳 Equation-wise:

$$p_i(t+1) = \sum_{j=1}^n \frac{1}{k_j} a_{ji} p_j(t).$$

where k_j is j's degree. Note: $k_i = \sum_{j=1}^n a_{ij}$.





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Inebriation and diffusion:

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Excellent observation: The same equation applies for stuff moving around a network, such that at each time step all material at node *i* is sent to its neighbors.

 $x_i(t)$ = amount of stuff at node *i* at time *t*.

$$x_i(t+1) = \sum_{j=1}^n \frac{1}{k_j} a_{ji} x_j(t).$$

Andom walking is equivalent to diffusion C.





Where is Barry?

S Linear algebra-based excitement: $p_i(t+1) = \sum_{j=1}^n a_{ji} \frac{1}{k_j} p_j(t)$ is more usefully viewed as

 $\vec{p}(t+1) = A^{\mathsf{T}} K^{-1} \vec{p}(t)$

where $[K_{ij}] = [\delta_{ij}k_i]$ has node degrees on the main diagonal and zeros everywhere else.

So... we need to find the dominant eigenvalue of $A^{\mathsf{T}}K^{-1}$.

- Expect this eigenvalue will be 1 (doesn't make sense for total probability to change).
- The corresponding eigenvector will be the limiting probability distribution (or invariant measure).
- Extra concerns: multiplicity of eigenvalue = 1, and network connectedness.





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Where is Barry?

🛞 By inspection, we see that

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 $\vec{p}(\infty) = \frac{1}{\sum_{i=1}^{n} k_i} \vec{k}$ satisfies $\vec{p}(\infty) = A^{\mathsf{T}} K^{-1} \vec{p}(\infty)$ with eigenvalue 1. \bigotimes We will find Barry at node *i* with probability proportional to its degree k_i . Beautiful implication: probability of finding Barry travelling along any edge is uniform. Diffusion in real space smooths things out. On networks, uniformity occurs on edges. lacktrian So in fact, diffusion in real space is about the edges too but we just don't see that.





Other pieces:

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Goodness: $A^{\mathsf{T}}K^{-1}$ is similar to a real symmetric matrix if $A = A^{\mathsf{T}}$.

Solution Consider the transformation $M = K^{-1/2}$:

$$K^{-1/2} A^{\mathsf{T}} K^{-1} K^{1/2} = K^{-1/2} A^{\mathsf{T}} K^{-1/2}.$$

Since $A^{\mathsf{T}} = A$, we have

 $(K^{-1/2}AK^{-1/2})^{\mathsf{T}} = K^{-1/2}AK^{-1/2}.$

Upshot: A^TK⁻¹ = AK⁻¹ has real eigenvalues and a complete set of orthogonal eigenvectors.
Can also show that maximum eigenvalue magnitude is indeed 1.





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