

# Random walks and diffusion on networks

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# Outline

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# Random walks on networks—basics:

- ▶ 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?
- ▶ Define  $p_i(t)$  as the probability that at time step  $t$ , our walker is at node  $i$ .
- ▶ We want to characterize the evolution of  $\vec{p}(t)$ .
- ▶ 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.
- ▶ Worse still: Barry is **hopelessly drunk**.

# Where is Barry?

- ▶ Consider simple undirected networks with an edges either present or absent.
- ▶ Represent network by a symmetric **adjacency matrix**  $A$  where

$$a_{ij} = 1 \text{ if } i \text{ and } j \text{ are connected,}$$

$$a_{ij} = 0 \text{ otherwise.}$$

- ▶ Barry is at node  $i$  at time  $t$  with probability  $p_i(t)$ .
- ▶ In the next time step he **randomly lurches** toward one of  $i$ 's neighbors.
- ▶ Equation-wise:

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

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

# Where is Barry?

- ▶ Linear algebra-based excitement:

$p_j(t+1) = \sum_{i=1}^n \frac{1}{k_i} a_{ji} p_i(t)$  is more usefully viewed as

$$\vec{p}(t+1) = AK^{-1}\vec{p}(t)$$

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  $AK^{-1}$ .
- ▶ 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.

# Where is Barry?

- ▶ By inspection, we see that

$$\vec{p}(\infty) = \frac{1}{\sum_{i=1}^n k_i} \vec{k}$$

satisfies  $\vec{p}(\infty) = AK^{-1}\vec{p}(\infty)$  with eigenvalue 1.

- ▶ We will find Barry at node  $i$  with probability proportional to its degree  $k_i$ .
- ▶ Nice implication: probability of finding Barry travelling along any edge is **uniform**.
- ▶ Diffusion in real space smooths things out.
- ▶ On networks, uniformity occurs on edges.
- ▶ So in fact, diffusion in real space is **about the edges too** but we just don't see that.

## Other pieces:

- ▶ Good news:  $AK^{-1}$  is similar to a real symmetric matrix.
- ▶ Consider the transformation  $M = K^{-1/2}$ :

$$K^{-1/2}AK^{-1}K^{1/2} = K^{-1/2}AK^{-1/2}.$$

- ▶ Since  $A^T = A$ , we have

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

- ▶ Upshot:  $AK^{-1}$  has real eigenvalues and a complete set of orthogonal eigenvectors.
- ▶ Can also show that maximum eigenvalue magnitude is indeed 1.
- ▶ Other goodies: next time round.

# References I

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