

# Social learning

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## Structural Analysis and Visualization of Networks



NATIONAL RESEARCH  
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- 1 Social learning
- 2 Reaching consensus
  - DeGroot model
- 3 Social influence networks

- Social learning is changing ones behavior or beliefs based on direct observation of others (imitation, aggregation, adoption)
- Local interactions (network)
- Information is dispersed through the network
- No centralized mechanism for information aggregation

"Reaching a consensus", Morris DeGroot 1974

Consensus - mutual agreement on a subject among group of people

- Group of people with opinions on the subject
- Can a common belief be reached?
- How long would it take?
- How each individual belief contribute to consensus?
- Which individuals have the most influence over final beliefs?

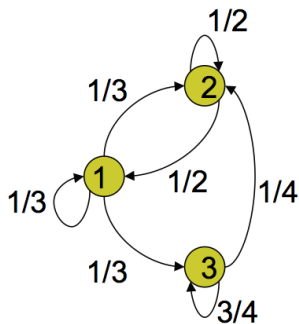
- Opinion  $p_i(t) \in [0..1]$ ,
- Initial opinion on the subject  $p_i(0)$
- $T_{ij}$  is weight on the opinion of others,  $i \rightarrow j$ , how much  $i$  "listens to opinion" of  $j$ ,  $\sum_j T_{ij} = 1$
- Opinion update

$$p_i(t+1) = \sum_j T_{ij} p_j(t)$$

- Could a consensus be reached, i.e. all opinions converge to the same value?

$$\lim_{t \rightarrow \infty} p_i(t) = p^\infty$$

# Example 1



$$T = \begin{pmatrix} 1/3 & 1/3 & 1/3 \\ 1/2 & 1/2 & 0 \\ 0 & 1/4 & 3/4 \end{pmatrix}$$

# Example 1

- Initial beliefs

$$p(0) = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}$$

- Updating

$$p(1) = Tp(0) = \begin{pmatrix} 1/3 & 1/3 & 1/3 \\ 1/2 & 1/2 & 0 \\ 0 & 1/4 & 3/4 \end{pmatrix} \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} = \begin{pmatrix} 1/3 \\ 1/2 \\ 0 \end{pmatrix}$$

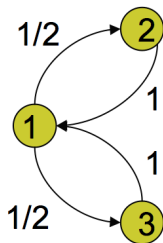
$$p(2) = Tp(1) = \begin{pmatrix} 1/3 & 1/3 & 1/3 \\ 1/2 & 1/2 & 0 \\ 0 & 1/4 & 3/4 \end{pmatrix} \begin{pmatrix} 1/3 \\ 1/2 \\ 0 \end{pmatrix} = \begin{pmatrix} 5/18 \\ 5/12 \\ 1/8 \end{pmatrix}$$

- Consensus

$$p(20) = Tp(19) = \begin{pmatrix} 3/11 \\ 3/11 \\ 3/11 \end{pmatrix}$$

$$p(21) = Tp(20) = p(20)$$

## Example 2



$$T = \begin{pmatrix} 0 & 1/2 & 1/2 \\ 1 & 0 & 0 \\ 1 & 0 & 0 \end{pmatrix}$$



## Example 2

- Initial beliefs

$$p(0) = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}$$

- Updating

$$p(1) = Tp(0) = \begin{pmatrix} 0 & 1/2 & 1/2 \\ 1 & 0 & 0 \\ 1 & 0 & 0 \end{pmatrix} \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} = \begin{pmatrix} 0 \\ 1 \\ 1 \end{pmatrix}$$

$$p(2) = Tp(1) = \begin{pmatrix} 0 & 1/2 & 1/2 \\ 1 & 0 & 0 \\ 1 & 0 & 0 \end{pmatrix} \begin{pmatrix} 0 \\ 1 \\ 1 \end{pmatrix} = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}$$

- Consensus

$$\nexists \lim_{t \rightarrow \infty} T^t p(0)$$

# Perron - Frobenius Theorem

Perron & Frobenius, 1912, linear algebra.

For stochastic matrices  $\sum_j T_{ij} = 1$ :

## Theorem

Let  $\mathbf{T}$  be a square 1) non-negative  $T_{ij} \geq 0$ , 2) irreducible, 3) aperiodic stochastic matrix. Then there exist

$$\lim_{t \rightarrow \infty} T_{ij}^t = \pi_j$$

where

$$\pi_j = \sum_i \pi_i T_{ij}$$

$\pi = (\pi_1, \pi_2, \dots, \pi_n)$  - is the left eigenvalue of  $\mathbf{T}$ , corresponding to  $\lambda_1 = 1$   
and  $\sum_i \pi_i = 1$

Irreducible matrix - strongly connected graph

Aperiodic matrix - the greatest common divisor of the length of the cycles in the associated graph is  $\text{gcd} = 1$

- Limiting belief

$$p(t) = Tp(t-1) = T^2p(t-2) = T^t p(0)$$

$$p^\infty = \lim_{t \rightarrow \infty} p(t) = \lim_{t \rightarrow \infty} T^t p(0)$$

$$\lim_{t \rightarrow \infty} T^t = \lim_{t \rightarrow \infty} \begin{pmatrix} T_{11} & \dots & T_{1n} \\ \dots & \dots & \dots \\ T_{n1} & \dots & T_{nn} \end{pmatrix}^t = \begin{pmatrix} \pi_1 & \pi_2 & \dots & \pi_n \\ \pi_1 & \pi_2 & \dots & \pi_n \\ \pi_1 & \pi_2 & \dots & \pi_n \end{pmatrix}$$

$$\mathbf{p}^\infty = \begin{pmatrix} \pi_1 & \pi_2 & \dots & \pi_n \\ \pi_1 & \pi_2 & \dots & \pi_n \\ \pi_1 & \pi_2 & \dots & \pi_n \end{pmatrix} \begin{pmatrix} p_1(0) \\ \dots \\ p_n(0) \end{pmatrix} = \begin{pmatrix} p^\infty \\ \dots \\ p^\infty \end{pmatrix}$$

- Left eigenvector

$$\pi \mathbf{T} = \pi \lambda$$

- Example 1

$$T^{20} = \begin{pmatrix} 1/3 & 1/3 & 1/3 \\ 1/2 & 1/2 & 0 \\ 0 & 1/4 & 3/4 \end{pmatrix}^{20} = \begin{pmatrix} 0.27 & 0.36 & 0.36 \\ 0.27 & 0.36 & 0.36 \\ 0.27 & 0.36 & 0.36 \end{pmatrix}$$

$$p^\infty = \begin{pmatrix} 0.27 & 0.36 & 0.36 \\ 0.27 & 0.36 & 0.36 \\ 0.27 & 0.36 & 0.36 \end{pmatrix} \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} = \begin{pmatrix} 0.27 \\ 0.27 \\ 0.27 \end{pmatrix}$$

- Left eigenvector

$$\pi \mathbf{T} = \pi \lambda$$

$$\lambda = \{1, 0.5, 0.083\}, \quad \pi_1 = (0.27, 0.36, 0.36)$$

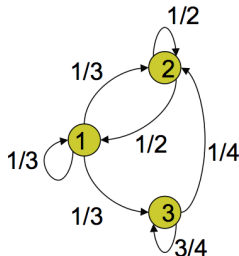
# Social influence

- Limiting belief

$$p^\infty = \lim_{t \rightarrow \infty} T^t p(0) = \Pi p(0) = \sum_i \pi_i p_i(0)$$

- Influence vector

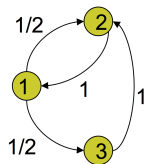
$$\pi = (\pi_1, \pi_2, \dots, \pi_n)$$



$$\pi = (0.27, 0.36, 0.36)$$

# Social influence

Equal weights on edges,  $A_{ij}$  - adjacency matrix,  $d_i$  - node out-degree



- Opinion updates:

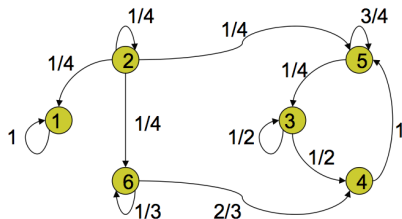
$$p_i = \sum_j T_{ij} p_j = \sum_j \frac{A_{ij}}{d_i} p_j$$
$$p = (D^{-1}A)p$$

- Random walk (PageRank)

$$r_i = \sum_j P_{ji} r_j = \sum_j \frac{A_{ji}}{d_j} r_j$$
$$r = r(D^{-1}A)$$

# Closed sets

- A set of nodes  $C$  is called a *closed set* if there is no direct link from the node in  $C$  to the node outside  $C$  (there is no  $i \in C$  and  $j \notin C$  such that  $T_{ij} > 0$ )



- $T$  is convergent if and only if every set of nodes that is strongly connected and closed is aperiodic
- Every strongly connected closed and aperiodic set will reach own consensus

- Time-varying updates:

$$p(t) = \left[ (1 - \lambda(t))I + \lambda(t)\hat{T} \right] p(t - 1)$$

DeMarzo, Vayanos, Zwibel, 2003

- Similar beliefs:

$$T_{ij}(p(t)) = \begin{cases} \frac{1}{n_i(p(t))}, & |p_i(t) - p_j(t)| < d, n_i = \#\{k : |p_i - p_k| < d\} \\ 0, & \text{otherwise} \end{cases}$$

Krause

- Close beliefs:

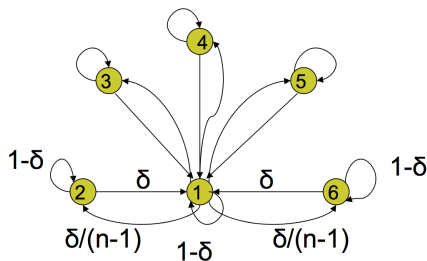
$$T_{ij}(p(t)) = \frac{e^{-\gamma_{ij}|p_i(t) - p_j(t)|}}{\sum_k e^{-\gamma_{ik}|p_i(t) - p_k(t)|}}$$

- Under some conditions consensus could be reached in all models



# Wisdom of the crowd

- Wisdom of crowd - taking into account collective opinion of a group of individuals for collective decision
- Claim: collective decision is better than decision by any individual due to independent judgements and aggregation process that removes random noise and averages out decisions.
- Will consensus converge to "correct" values?
- Opinion leader



Noah Friedkin 1991, ...,1999

- interpersonal influence for collective decisions
- modifying attitudes and opinions by interaction
- influence process in a group of  $N$  actors:

$$\mathbf{y}(t) = \mathbf{D}\mathbf{T}\mathbf{y}(t - 1) + (\mathbf{I} - \mathbf{D})\mathbf{y}(0)$$

$\mathbf{y}(t)$  - vector of actors' opinion at time  $t$

$\mathbf{y}(0)$  - vector of actors' initial opinion

$\mathbf{T}$  -  $N \times N$  matrix of interpersonal influence

$\mathbf{D} = \text{diag}(d_{11}, \dots, d_{NN})$  - matrix of actors' susceptibilities to interpersonal influence

- Assuming the process reaches an equilibrium,  $\lim_{t \rightarrow \infty} \mathbf{y}(t) = \mathbf{y}^\infty$

$$\mathbf{y}^\infty = \mathbf{D}\mathbf{T}\mathbf{y}^\infty + (\mathbf{I} - \mathbf{D})\mathbf{y}(0)$$

- Solution:

$$\mathbf{y}^\infty = (\mathbf{I} - \mathbf{D}\mathbf{T})^{-1}(\mathbf{I} - \mathbf{T})\mathbf{y}(0)$$

- Limiting beliefs (no consensus):

$$\mathbf{y}^\infty = \begin{pmatrix} y_1^\infty \\ y_2^\infty \\ \dots \\ y_n^\infty \end{pmatrix}$$

- Reaching a Consensus, M. DeGroot, J. Amer. Stat. Assoc., Vol 69, N 345, pp 118-121, 1974
- A Necessary and Sufficient Condition for Reaching a Consensus Using DeGroot's Method, R. Berger, J. Amer. Stat. Assoc., Vol 76, N 374, pp 415-418, 1981
- Naive Learning in Social Networks and the Wisdom of Crowds, B. Golub and M. Jackson, Amer. Econ. J. Microeconomics. 2010
- Social Influence Networks and Opinion Change, Friedkin, Noah E. and Eugene C. Johnsen. Advances in Group Processes 16:1-29, 1999