

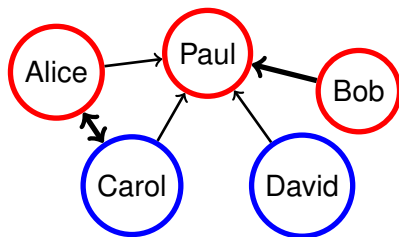
# From Networks to Numerical Linear Algebra

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## Example: Opinions in Networks



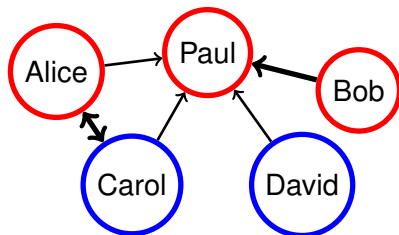
# Basic model

Extension of DeGroot model due to Friedkin and Johnsen:

- ▶ Directed graph with nodes  $1, \dots, n$ , weights  $w_{ij}$
- ▶ Each node has two quantities:
  - ▶ Fixed *internal opinion*  $s_i \in \mathbb{R}$
  - ▶ Variable *expressed opinion*  $z_i \in \mathbb{R}$
- ▶ Update equation:

$$z_i^{\text{new}} \leftarrow \frac{s_i + \sum_{j \in N(i)} w_{ij} z_j^{\text{old}}}{1 + \sum_{j \in N(i)} w_{ij}}$$

## Example: Opinions in Networks



$$W = \begin{bmatrix} 0 & 0 & w_{AC} & 0 & 0 & w_{AP} \\ 0 & 0 & 0 & 0 & 0 & w_{BP} \\ w_{CA} & 0 & 0 & 0 & 0 & w_{CP} \\ 0 & 0 & 0 & 0 & 0 & w_{DP} \\ 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

# Matrix reformulation

Scalar form:

$$z_i^{\text{new}} \leftarrow \frac{s_i + \sum_{j \in N(i)} w_{ij} z_j^{\text{old}}}{1 + \sum_{j \in N(i)} w_{ij}}$$

Matrix form:

$$(D + I)z^{\text{new}} = s + Wz^{\text{old}}$$

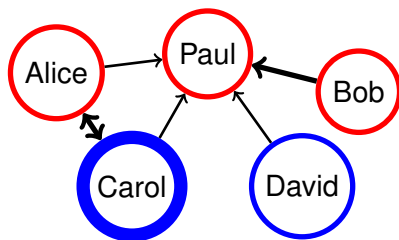
This is Jacobi iteration! Converges to solution of

$$(L + I)x = s$$

where  $L = D - W$  is the (directed) graph Laplacian:

$$L_{ij} = \begin{cases} -w_{ij}, & i \neq j \\ \sum_{k \in N(i)} w_{ik}, & i = j \end{cases}$$

## Another perspective



Carol is “pulled” to:

- ▶ To be true to her personal beliefs ( $z_C - s_C$  small)
- ▶ To agree with Alice ( $z_C - z_A$  small)
- ▶ To agree with Paul ( $z_C - z_B$  small)

She is unhappy to the extent that she cannot reconcile these.

# Equilibrium state and game theory

Define local cost function:

$$c_i = \frac{1}{2} \left( (z_i - s_i)^2 + \sum_{j \in N(i)} w_{ij} (z_i - z_j)^2 \right)$$

Node  $i$  chooses opinion to optimize  $z_i$ :

$$\frac{\partial c_i}{\partial z_i} = (z_i - s_i) + \sum_{j \in N(i)} w_{ij} (z_i - z_j) = 0$$

Nash equilibrium satisfies  $(L + I)x = s$ .

# Nash equilibrium vs social optimum

Define the *social cost*

$$c(z) = \sum_i c_i(z) = \frac{1}{2} \left( z^T (A + I) z - 2z^T s + s^T s \right).$$

where  $A = L + L^T$

- ▶ Nash equilibrium: Node  $i$  chooses  $x_i$  to minimize  $c_i$ .
- ▶ Social optimum: Choose  $y$  globally to minimize  $c(y)$

Equations for social optimum:  $(A + I)y = s$ .



# Price of anarchy

The *price of anarchy* is

$$\text{PoA}(s) = \frac{c(y)}{c(x)} = \frac{s^T B s}{s^T C s}$$

where

$$B = (A + I)^{-1} - I + (A + I)^{-1} A (A + I)^{-1}$$

$$C = ((L + I)^{-1} - I)^T ((L + I)^{-1} - I) + (L + I)^{-T} A (L + I)^{-1}$$

Undirected case:  $L$  symmetric,  $B = p(L)$ ,  $C = q(L)$ , and

$$\max_{s \neq 0} \frac{s^T B s}{s^T C s} = \max_{\lambda \in \Lambda(L)} \frac{p(\lambda)}{q(\lambda)} \leq \max_{t \geq 0} \frac{p(t)}{q(t)} = \frac{9}{8}.$$

Directed graphs: still an eigenvalue problem.

# Mean opinion and influence

Let  $e$  be the vector of all ones. Mean opinion at Nash is:

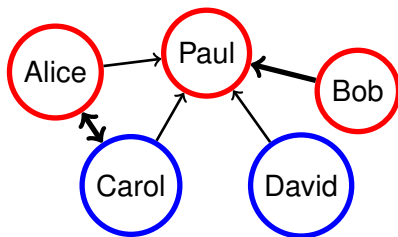
$$\bar{x} = \frac{1}{n} e^T x = \frac{1}{n} e^T (L + I)^{-1} s = f^T s$$

The *influence vector*

$$f = (L + I)^{-T} e$$

tells how much each node influences the mean opinion.

## Influence in the model network



Bold lines are weight 2, regular are weight 1:

$$f = \left[ \frac{1}{2} \quad \frac{1}{3} \quad \frac{1}{2} \quad \frac{1}{2} \quad \frac{19}{6} \right]^T$$

# Distribution of influence

The *uniform influence* case  $f = e/n$  occurs when

$$L^T e = 0,$$

i.e. graph is *Eulerian* (in-degree weight = out-degree weight).  
Uniform influence always true for socially optimal opinion!

In general, max influence is

$$\max_i f_i = \|(L + I)^{-1}\|_1$$

Note:  $\|(L + I)^{-1}\|_\infty = 1$ . What about other norms?

## Variance of opinion

Assume  $s$  is normalized to mean zero. Variance in intrinsic opinion:

$$\text{Var}[s] = \frac{1}{n} s^T s$$

What about the variance in the expressed opinion at Nash?

$$\frac{\text{Var}[x]}{\text{Var}[s]} = \frac{s^T (L + I)^{-1} (I - ee^T/n) (L + I)^{-1} s}{s^T s}$$

So if  $s$  is not identically zero, then

$$\frac{\sigma_x}{\sigma_s} \leq \|(I - ee^T/n)(L + I)^{-1}\|_2 \leq \sqrt{\max_i f_i}$$

Variance can only increase if influence is non-uniform.

# More entertainment

Can add edges to improve social cost (by  $\leq$  PoA):

1. Figuring out where to add a little weight is easy
2. Figuring out best edge additions is NP-hard

Changing from quadratic cost is interesting, but harder.

David S. Bindel, Sigal Oren, and Jon Kleinberg. “How Bad is Forming Your Own Opinion?,” FOCS 2011.