

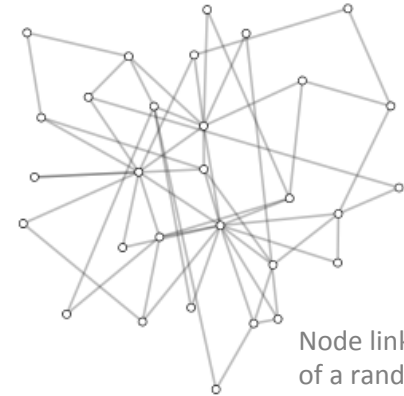
# Anisotropic Radial Layout for Visualizing Centrality and Structure in Graphs

Mukund Raj and Ross T. Whitaker

25<sup>th</sup> International Symposium on Graph Drawing and Network Visualization  
Boston, September 2017

# Node-link Diagram

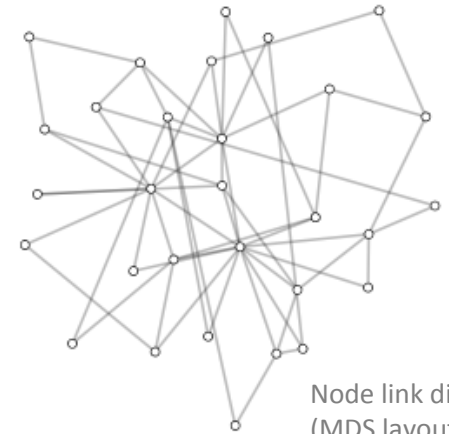
- Goals
  - Improve aesthetics
  - Reduce visual clutter
  - ***Convey features of interest***
    - ***Structural features based on internode distances***
    - ***Importance of nodes***



Node link diagram  
of a random graph

# Visualizing Structure

- Preserve internode distances in the drawing
- Method:
  - Dimensionality reduction techniques
    - **Multidimensional Scaling (MDS)**
    - t-SNE (stochastic neighbor embedding)



Node link diagram  
(MDS layout)

# Multidimensional Scaling (MDS)

- Conveys similarity between objects
- Minimizes energy function (stress)

$$\sigma(X) = \sum_{u,v} w_{uv} (d_{uv} - \|\bar{x}_u - \bar{x}_v\|_2)^2$$

$$X \in \mathbb{R}^{n \times 2}$$

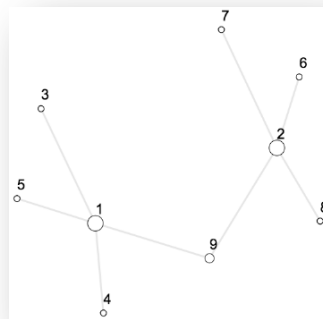
$d_{uv}$  := ideal distance between nodes  $u$  and  $v$

$w_{uv}$  := weight on edge between  $u$  and  $v$

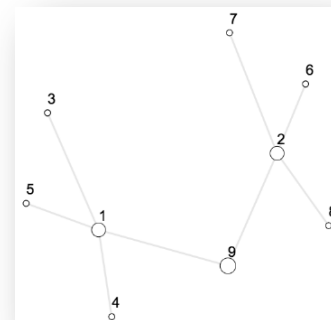
$\bar{x}_u \in \mathbb{R}^2$  := position of  $x$

# Importance of Nodes

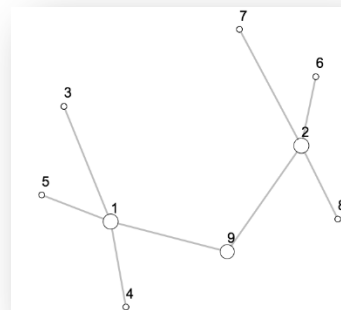
- Using graph structure
  - Node centrality
    - *Degree centrality*: number of neighbors
    - *Closeness centrality*: reciprocal of sum of distances
    - *Betweenness centrality*: number of shortest paths passing through the node
    - Many more ...



Degree centrality

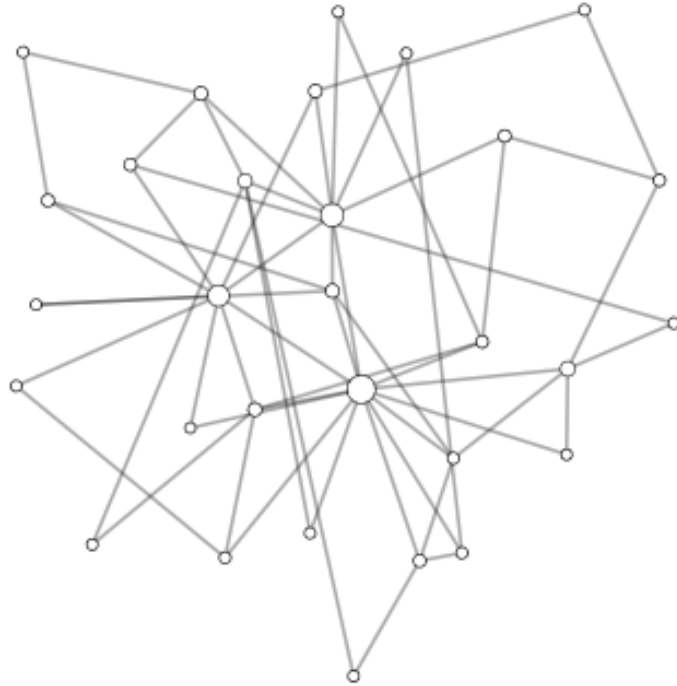


Closeness centrality



Betweenness centrality

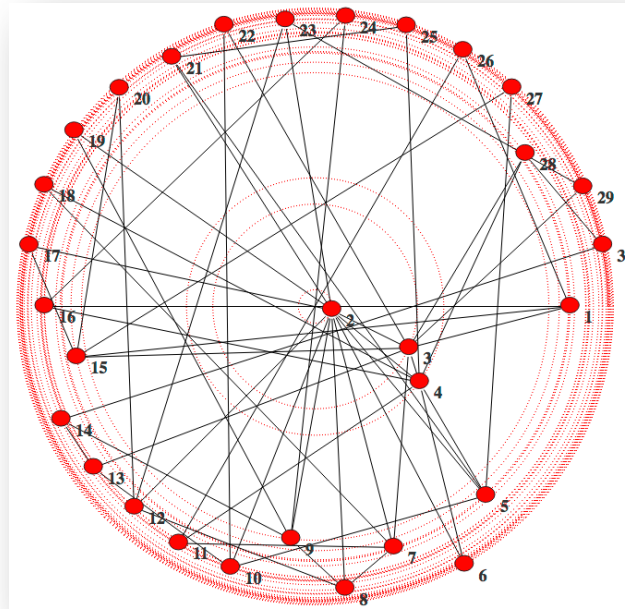
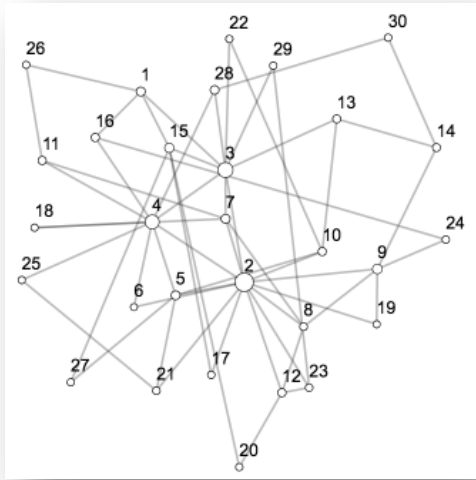
# Visualizing Centrality



- Node size encodes centrality
- Conflict between position and size channels

# Visualizing Centrality

- Radial layout [Yee 2001]

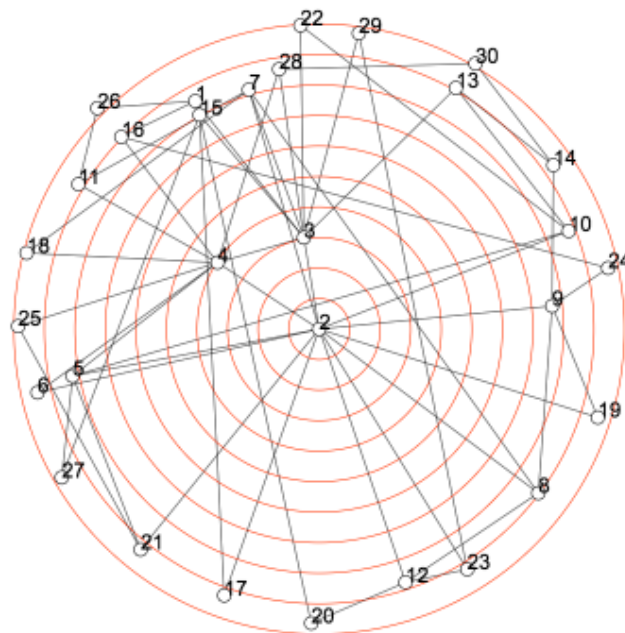
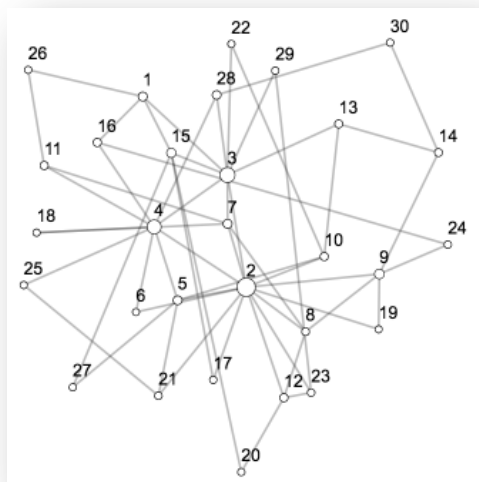


Distance from center encodes betweenness centrality.

\* Figure generated using SocNetV.

# Visualizing Centrality and Structure

- (Approximate) Distance preserving radial layout [Brandes 2009, Baingana 2014]

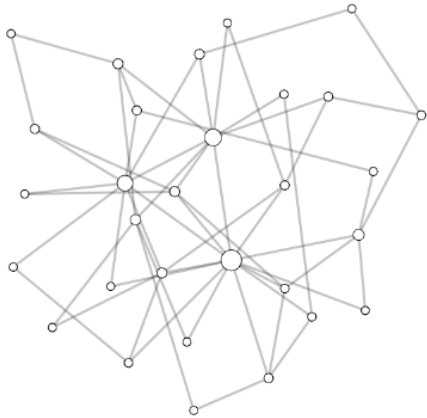


Minimize:  $\underbrace{\sigma(X)}_{\text{MDS objective}} + \underbrace{\rho(X, \bar{c})}_{\text{Penalty}}$

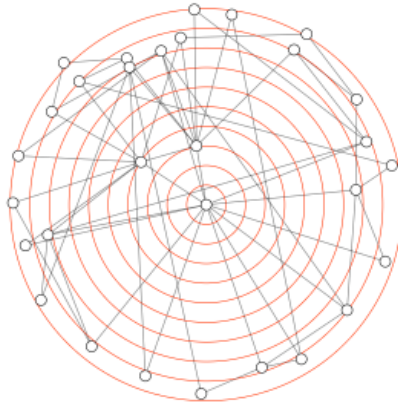


# Objectives

- Convey centrality
- Avoid conflicting perceptual cues
- Reduce structural (distance) distortion



MDS layout preserves structure



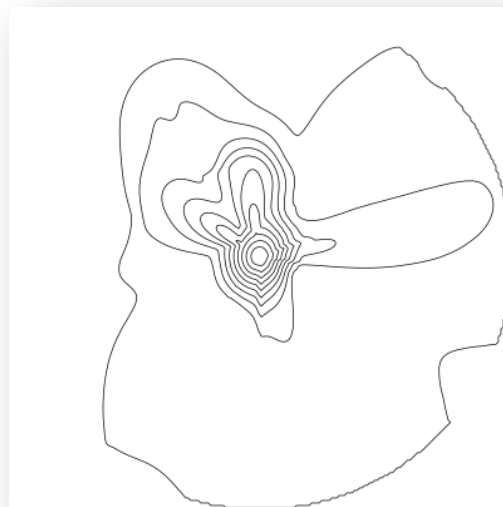
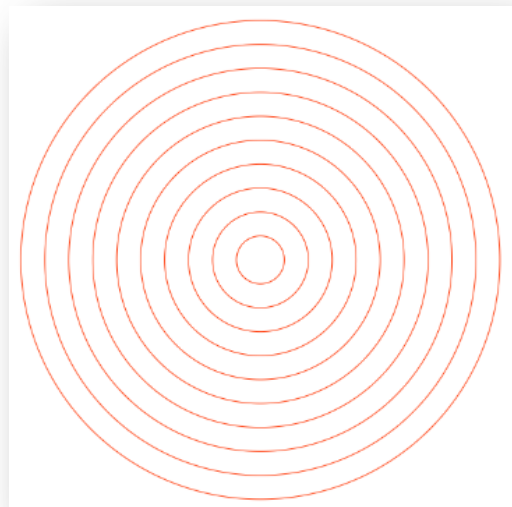
Radial layout highlights node centrality



Highlight centrality and minimize distance distortion

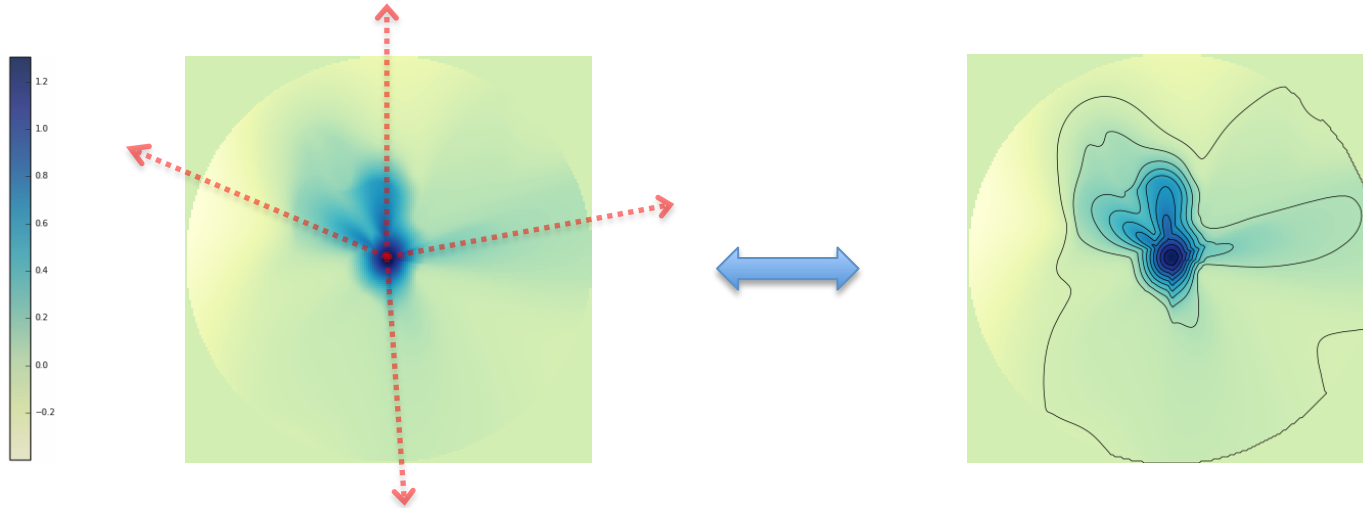
# Anisotropic Radial Layouts

- Relax circular constraint
- Use star shaped curves



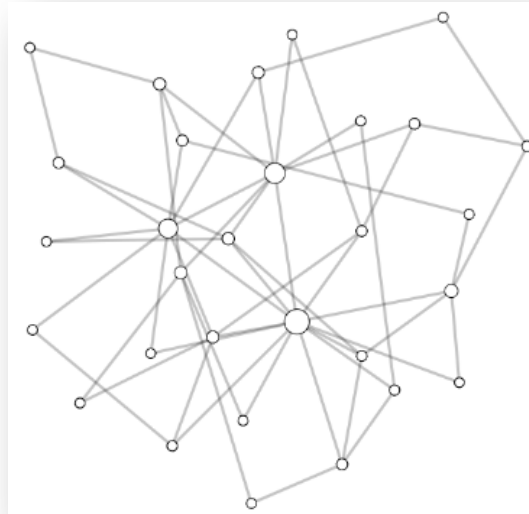
# Radial Monotonicity

- Strictly decreasing along all center outward directions
- Guarantees star shaped contours



# Step1: Initialization

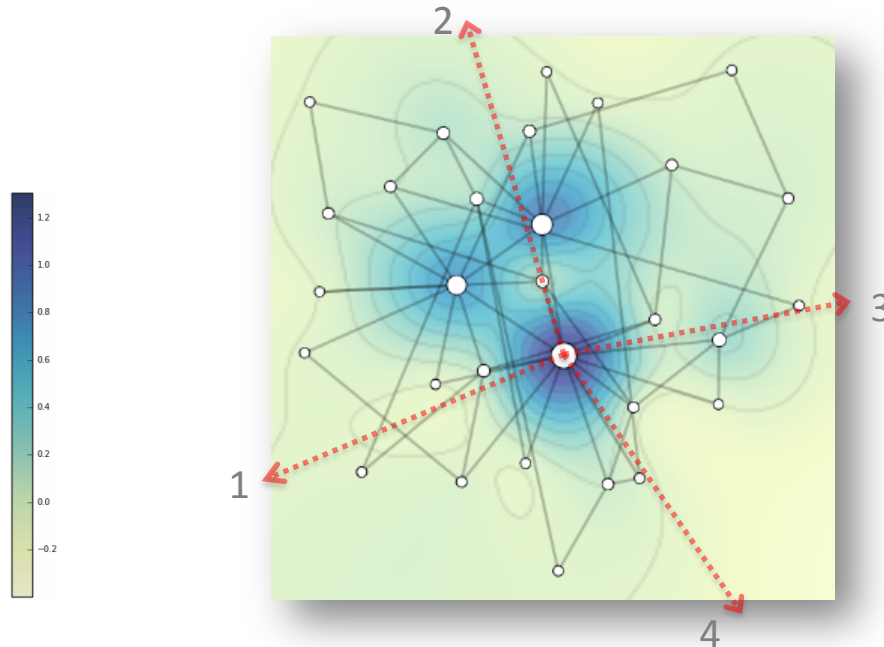
- Initialize positions using MDS.



Node sizes encode centrality.

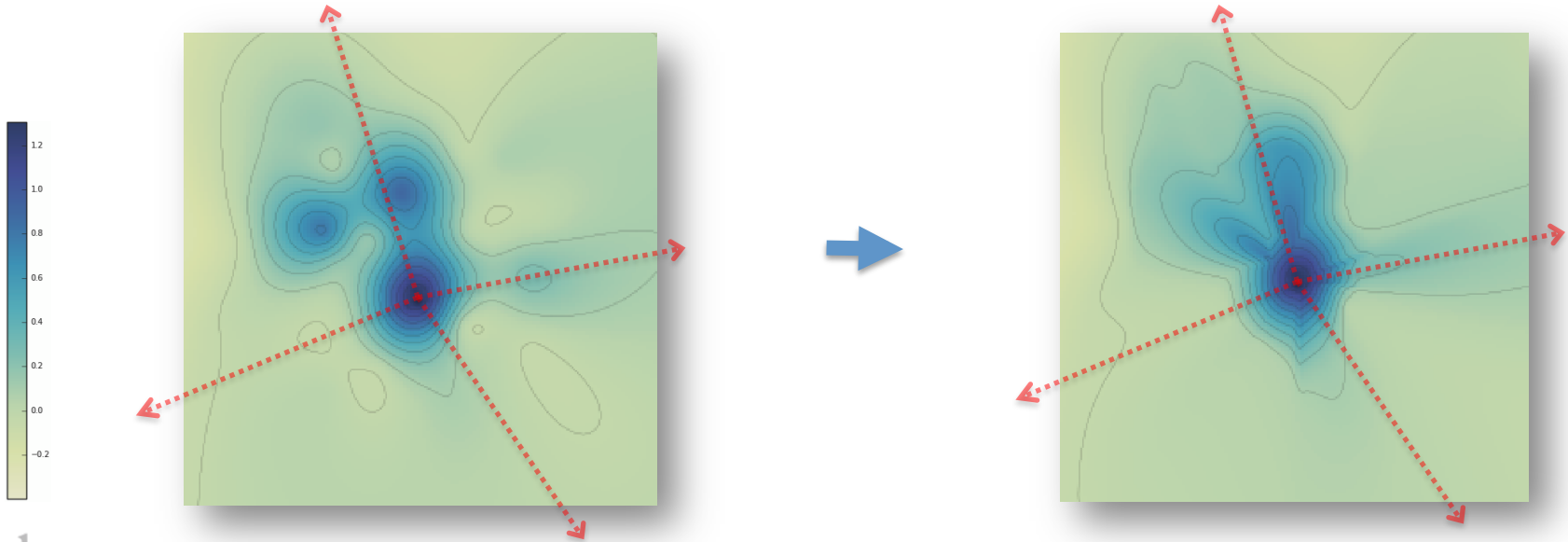
# Step2: Interpolation

- Smooth interpolation of the centrality (e.g. using thin plate spline)



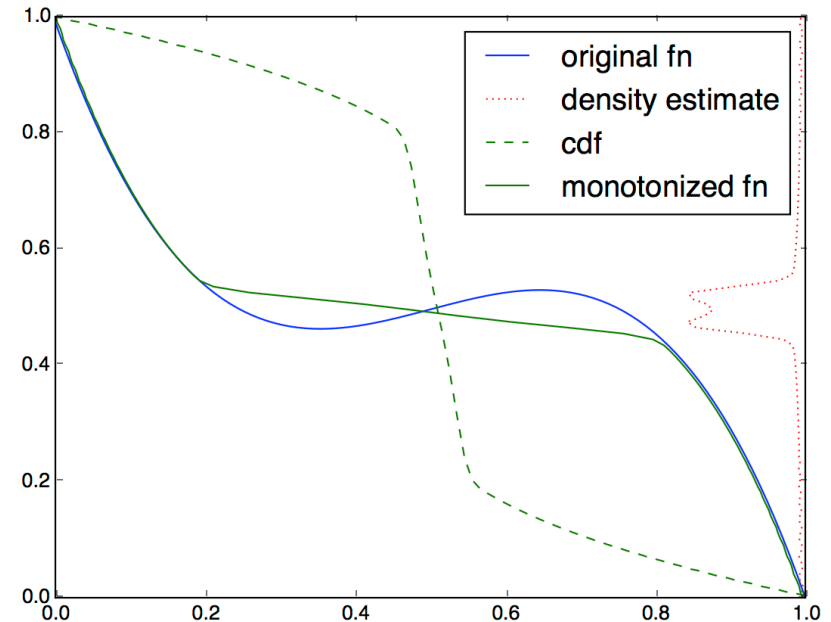
# Step 3: Monotonization

- Monotonizing along radial axis



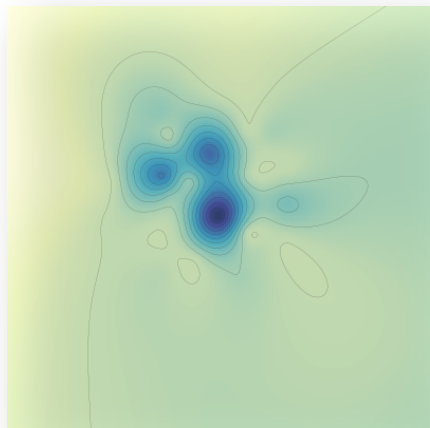
# Step3: Monotonization

- Monotonizing a 1D function [Dette et al 2006]
- Steps
  1. Construct density estimate
  2. Compute cdf
  3. Invert cdf
- Output: smooth monotonic approximation

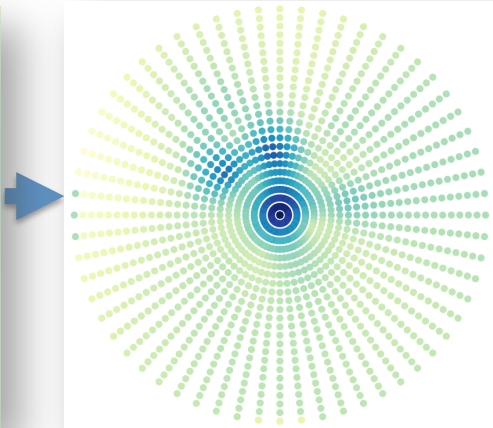


# Step3: Monotonization

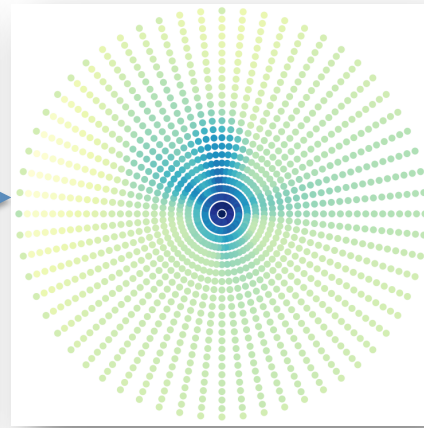
- Radially monotonizing a 2D field



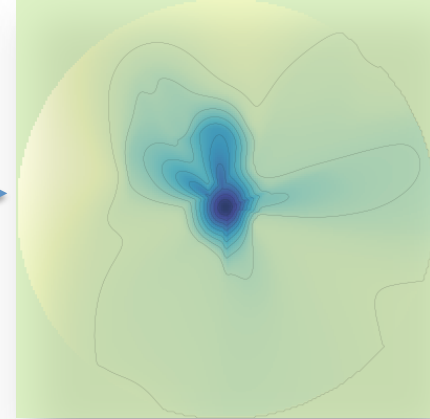
Interpolation field on a Cartesian grid



Resample on a regular **polar** grid



Monotonize along each angle



Interpolated and resampled **Monotonic field 'M'**

- If input field is smooth, independent 1D monotonization results in a smooth 2D field [Dette et al 2006]



# Step 4: Optimization objective

- Combine MDS with centrality
- Add penalty for deviation of node centrality from associated field value

$$\gamma(X) = \underbrace{\sigma(X)}_{\text{MDS stress}} + w_\rho \underbrace{\left( M_{X, \bar{c}}(X) - \bar{c} \right)}_{\text{Penalty}}$$

$$X \in \mathbb{R}^{n \times 2}$$

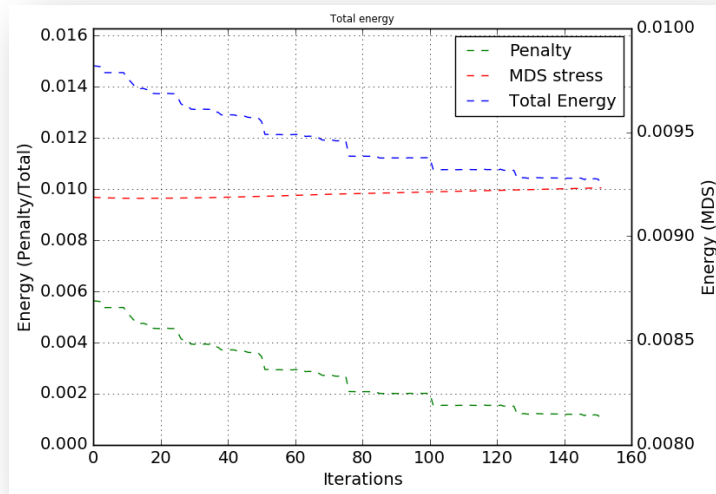
$$\bar{c} \in \mathbb{R}^n$$

$$M_{X, \bar{c}}(\cdot) := \text{Monotonic field}$$

# Step 5: Optimization

- Using gradient descent

$$\nabla\gamma(X) = \nabla\sigma(X) + w_\rho \times \underbrace{2(M_{X,\bar{c}}(X) - \bar{c}) \odot \nabla M_{X,\bar{c}}(X)}_{\text{Gradient of penalty}}$$



Gradient of penalty

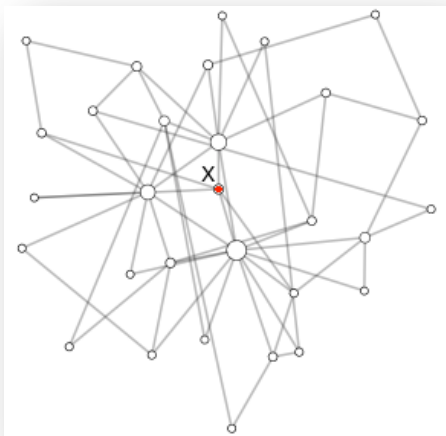
$$X \in \mathbb{R}^{n \times 2}$$

$$\bar{c} \in \mathbb{R}^n$$

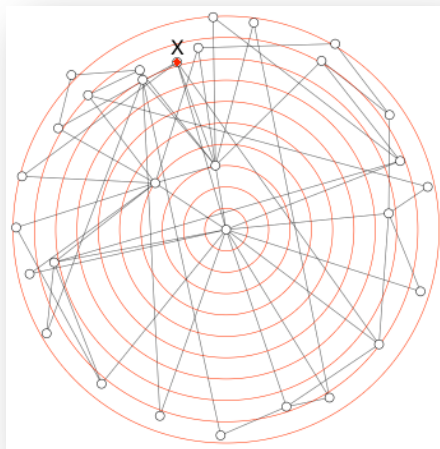
$M_{X,\bar{c}}(\cdot) :=$  Monotonic field

# Results

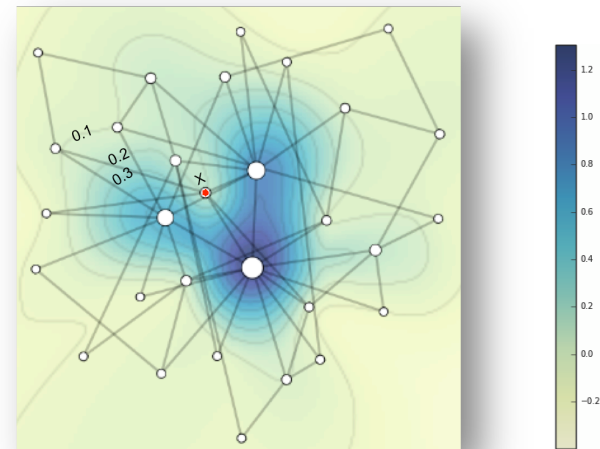
- Random graph generated using the Barabasi-Albert model.



MDS layout



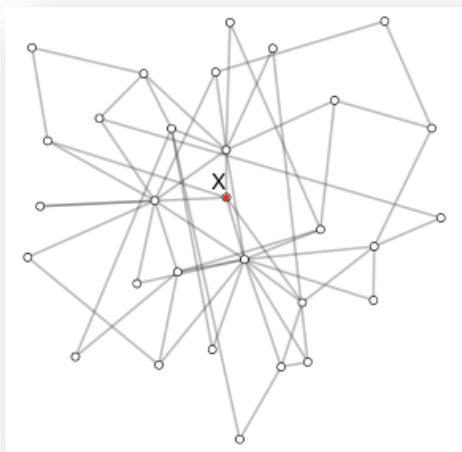
Radial layout



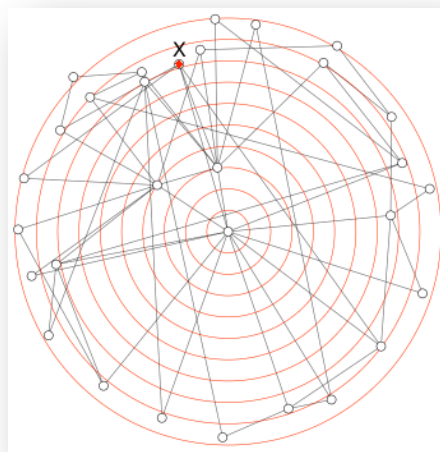
Anisotropic radial layout

# Results

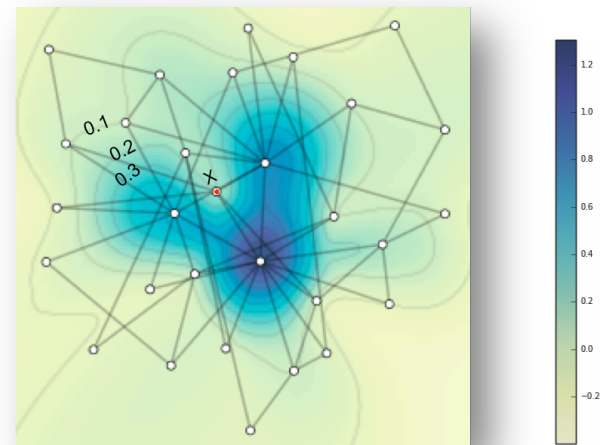
- Random graph generated using the Barabasi-Albert model.



MDS layout



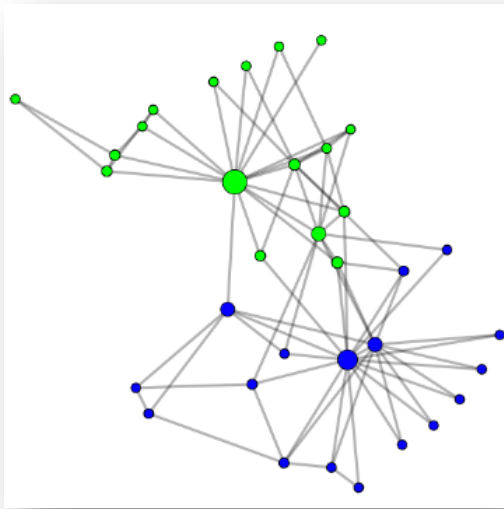
Radial layout



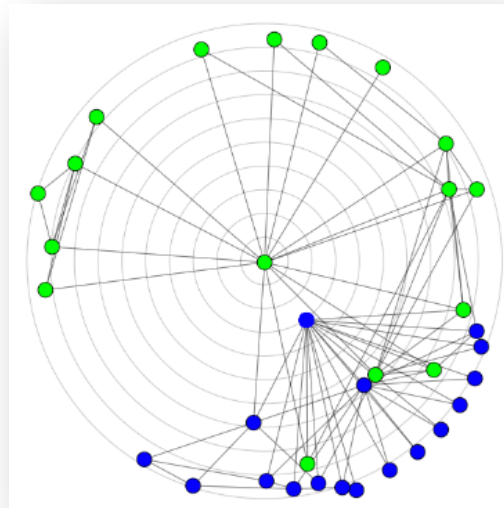
Anisotropic radial layout

# Results

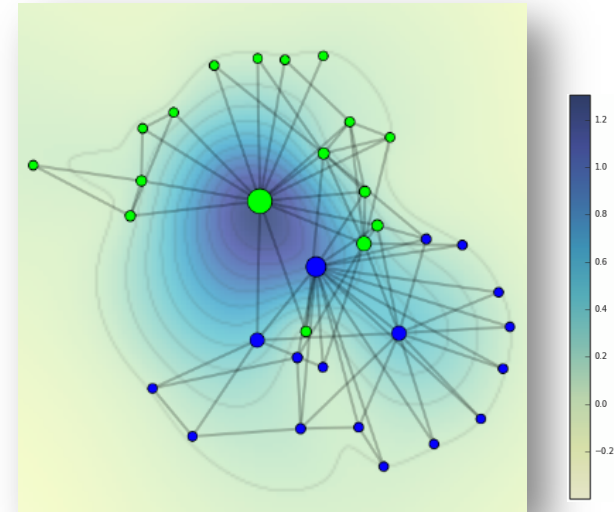
- Zachary's karate club network [Zachary 1977]



MDS layout



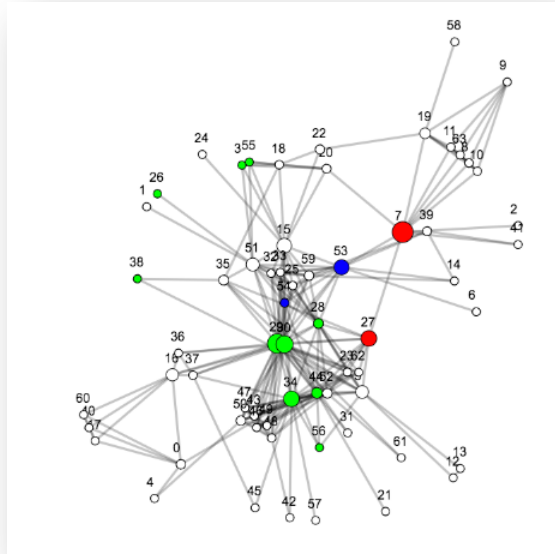
Radial layout



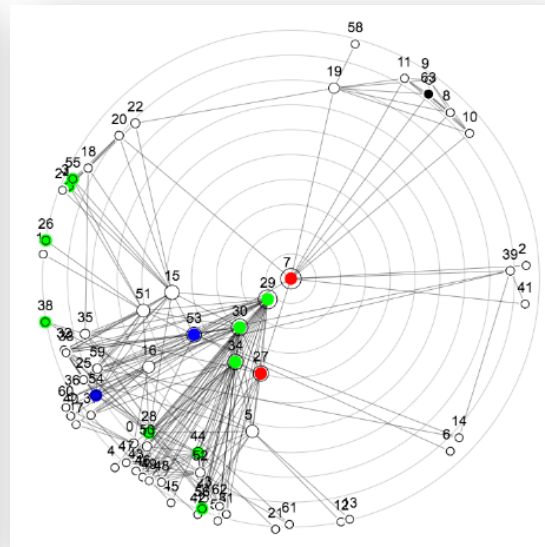
Anisotropic radial layout

# Results

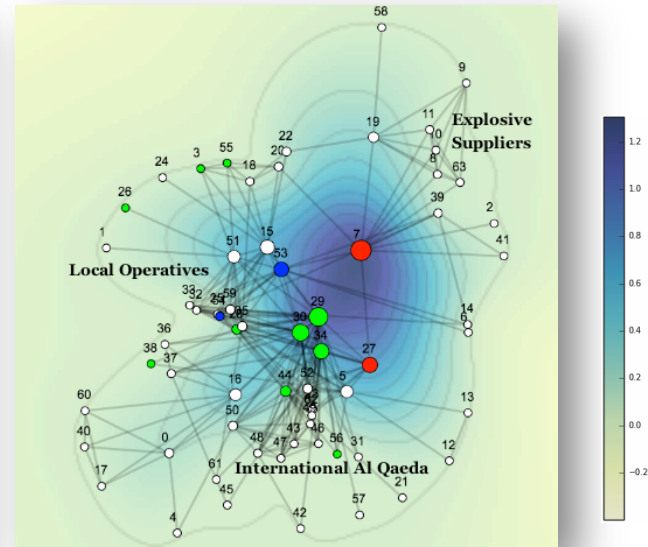
- Terrorist network from the Madrid train bombing incident in 2004 [Rodriguez 2005]



MDS layout



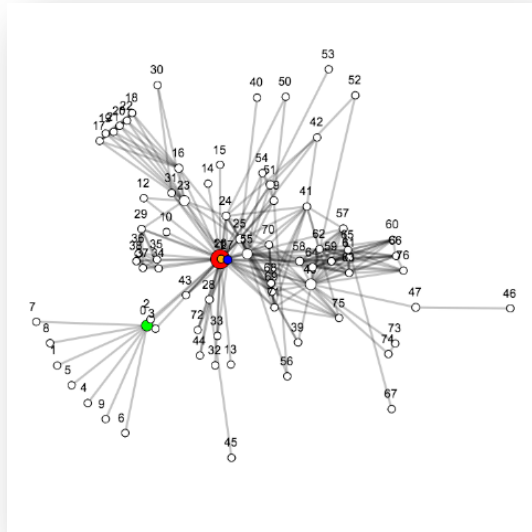
Radial layout



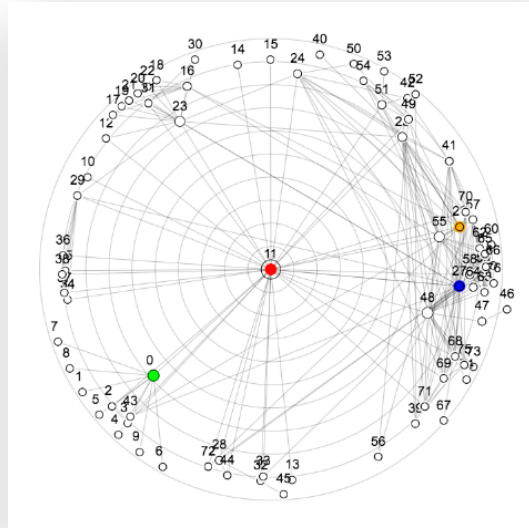
Anisotropic radial layout

# Results

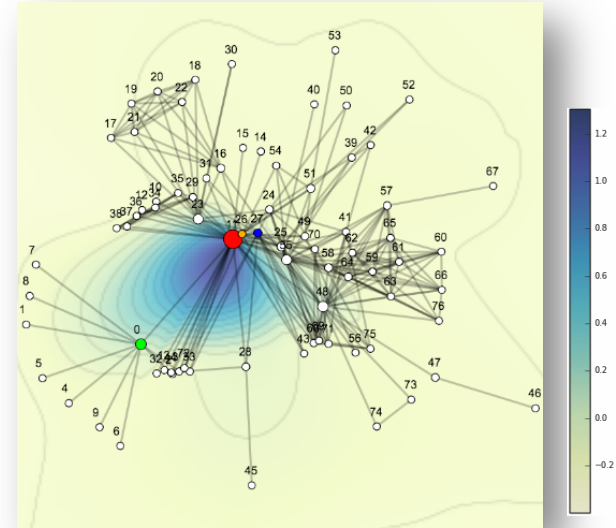
- Les Miserables character associations



MDS layout



Radial layout



Anisotropic radial layout

# Conclusions

- Strategy for preserving centrality and structure
- Layout Algorithm
- Suitable for real world networks
- Future work
  - Automatic parameter estimation
  - Better optimization method
  - Experiment with large graph (higher node count)



# Thanks

- This work was supported by National Science Foundation (NSF) grant IIS-1212806.