



PAKDD2020
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NODE CONDUCTANCE: A Scalable Node Centrality Measure on Big Networks

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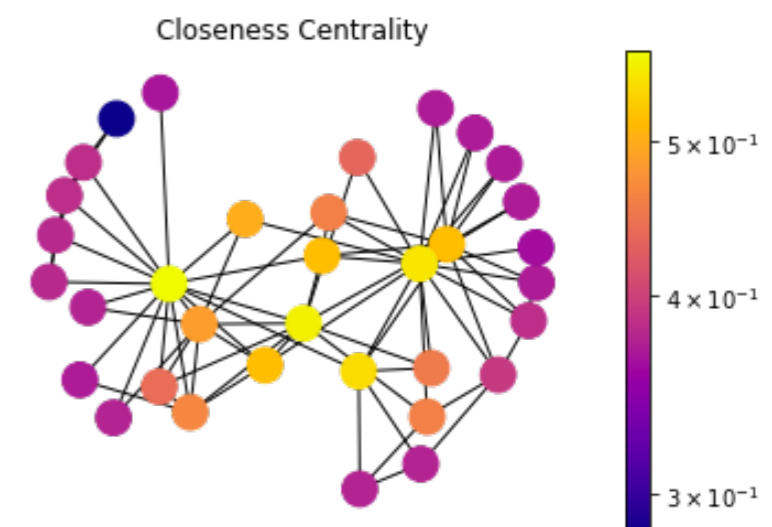
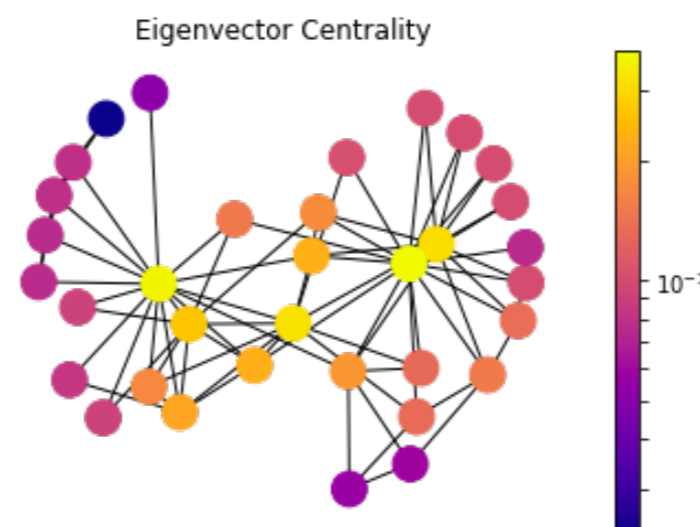
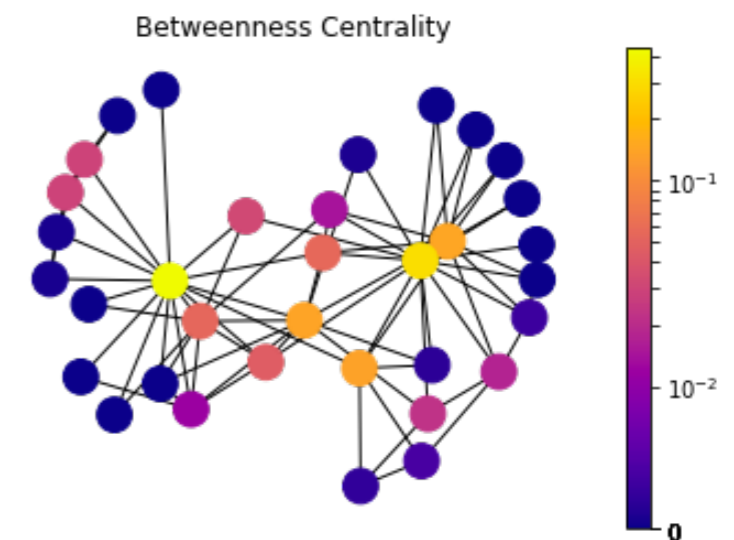
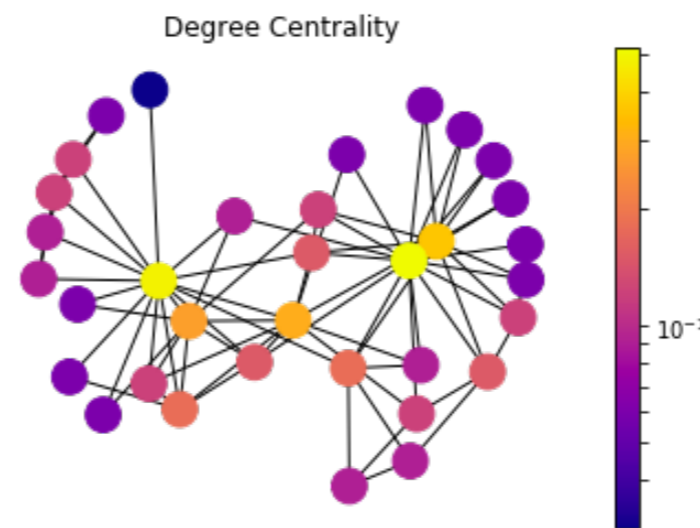
Node Centrality

- **Local Centrality**

- ego-network
- less informative

- **Global Centrality**

- ideal routes
- unrealistic
- infeasible to compute

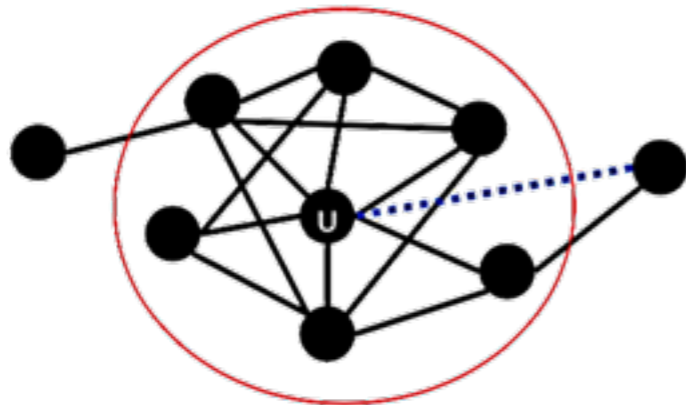


Refer to <https://aksakalli.github.io/2017/07/17/network-centrality-measures-and-their-visualization.html>

Node Conductance

- **Conductance**

- measures how hard it is to leave **a set of** nodes



$$\phi(S) = \frac{\text{cut}(S)}{\min(\text{vol}(S), \text{vol}(\bar{S}))}$$

- **Node Conductance**

- measures how hard it is to leave **a certain** node
- the sum of the probability that i is revisited at s -th step, where s is the integer between 1 and ∞ .

$$\text{NC}_{\infty}(i) \equiv \sum_{s=1}^{\infty} P(i|i, s).$$

Node Conductance

$$\text{NC}_\infty(i) \equiv \sum_{s=1}^{\infty} P(i|i, s).$$

Definition

adjacency matrix

degree vector

diagonal matrix of degree

Notation

\mathbf{A}

$\mathbf{d} = \mathbf{A}\mathbf{1}$

$\mathbf{D} = \text{diag}(\mathbf{d})$

- For a walk starting at node i , the probability that we find it at j after exactly s steps is given by

$$P(j|i, s) = [(\mathbf{D}^{-1} \mathbf{A})^s]_{ij}.$$

- NC_r denotes the sum of the probability that the node is revisited at the step s , s is between 1 and r

$$\text{NC}_r(i) = \sum_{s=1}^r P(i|i, s) = \mathbf{P}_{ii}^{(r)}, \quad \mathbf{P}^{(r)} = \sum_{s=1}^r (\mathbf{D}^{-1} \mathbf{A})^s,$$

Node Conductance

- Supposed that r approaches infinity,

$$\begin{aligned}\mathbf{P}^{(\infty)} &= \sum_{s=1}^{\infty} (\mathbf{D}^{-1} \mathbf{A})^s = \sum_{s=0}^{\infty} (\mathbf{D}^{-1} \mathbf{A})^s - \mathbf{I} \\ &= (\mathbf{I} - \mathbf{D}^{-1} \mathbf{A})^{-1} - \mathbf{I} = (\mathbf{D} - \mathbf{A})^{-1} \mathbf{D} - \mathbf{I}.\end{aligned}$$

Laplacian Matrix

- Pseudo-inverse of Laplacian Matrix:

$$g(\lambda_k) = \begin{cases} \frac{1}{\lambda_k}, & \text{if } \lambda_k \neq 0 \\ 0, & \text{if } \lambda_k = 0 \end{cases}, \quad \mathbf{L}_{ii}^\dagger = \sum_{k=1}^{N-1} g(\lambda_k) u_{ik}^2,$$

- Node Conductance only concerns about the diagonal:

$$\text{NC}_\infty(i) \propto \mathbf{L}_{ii}^\dagger \cdot d_i,$$

Related Work

$$SC(i) = \sum_{s=1}^{\infty} \frac{(A^s)_{ii}}{s!}$$

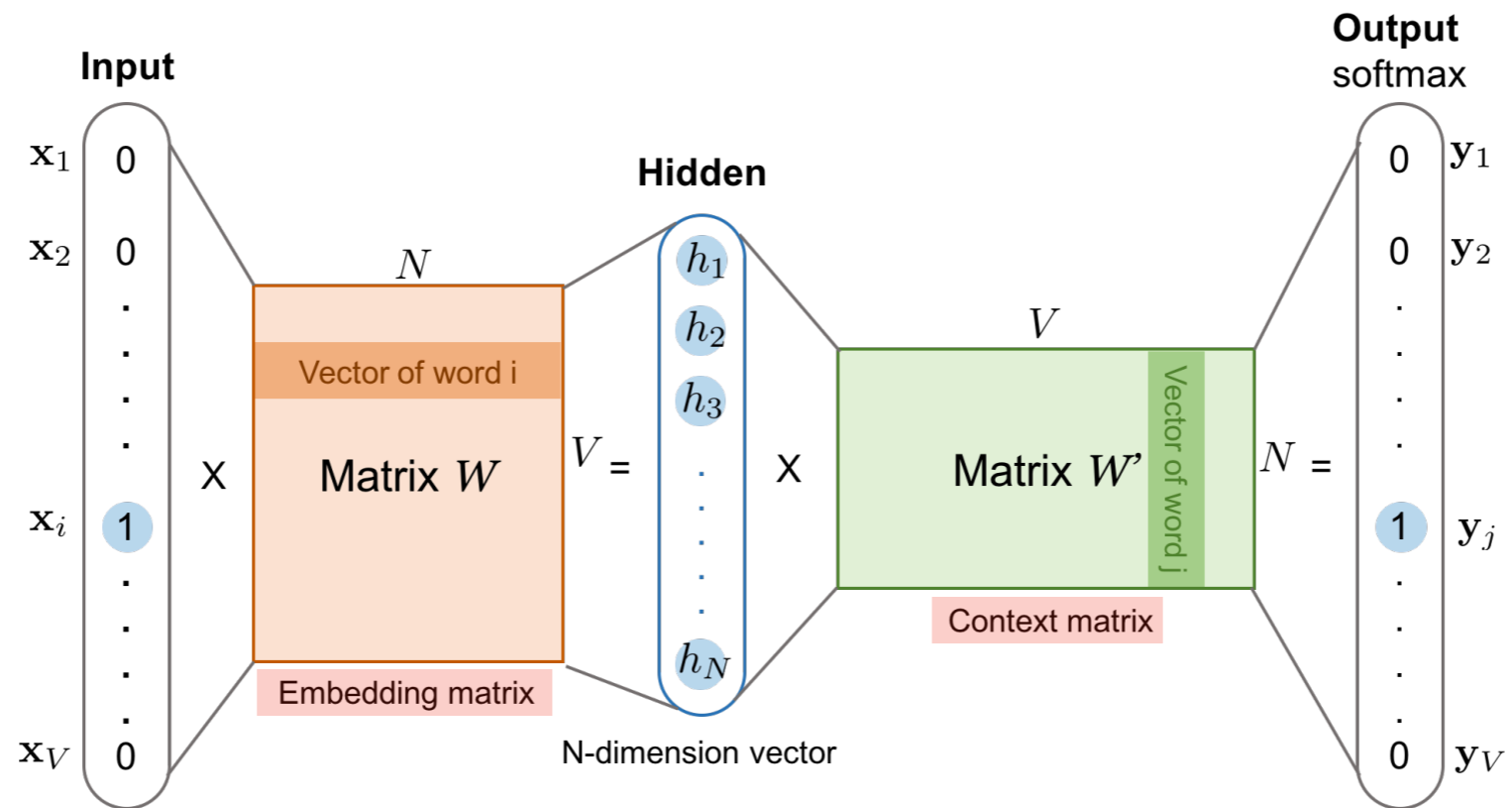
- Subgraph Centrality
 - the “sum” of closed walks of different lengths in the network starting and ending at vertex i .
 - adding a scaling factor to the denominator in order to make the SC value converge
 - less interpretive

Related Work

$$\text{PR} = \mathbf{D}(\mathbf{D} - \alpha\mathbf{A})^{-1}\mathbf{1}$$

- PageRank
 - the stationary distribution of the random walk
 - the probability that a random walk, with infinite steps, starts from **any node** and hits the node under consideration.

DeepWalk Vectors



- **Syntagmatic:** If two nodes have a strong connection in the network, the value of $\mathbf{W}_i \cdot \mathbf{C}_j$ is large.
- **Paradigmatic:** If two nodes have similar neighbors, the value of $\mathbf{W}_i \cdot \mathbf{W}_j$ is high.

DeepWalk Vectors & Node Conductance

- DeepWalk loss function

$$\mathcal{L} = \sum_{i \in \mathcal{V}_W} \sum_{j \in \mathcal{V}_C} \#(i, j)_r (\log \sigma(\mathbf{w}_i \cdot \mathbf{c}_j)) \\ + \sum_{i \in \mathcal{V}_W} \#(i)_r \left(k \cdot \sum_{\text{neg} \in \mathcal{V}_C} P(\text{neg}) \log \sigma(-\mathbf{w}_i \cdot \mathbf{c}_{\text{neg}}) \right).$$

- Comparing the derivative to zero $\text{NC}_r(i)$

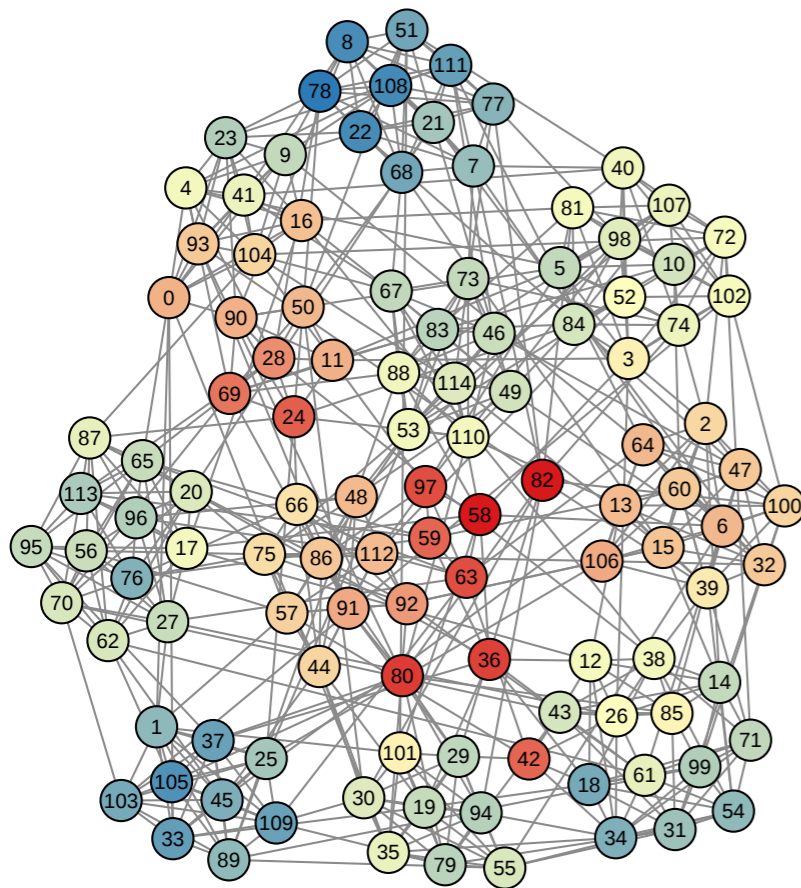
$$\mathbf{w}_i \cdot \mathbf{c}_j = \log \left(\frac{\#(i, j)_r}{\#(i)_r P(j)} \right) - \log k,$$

- Estimating probability by the actual number of observations

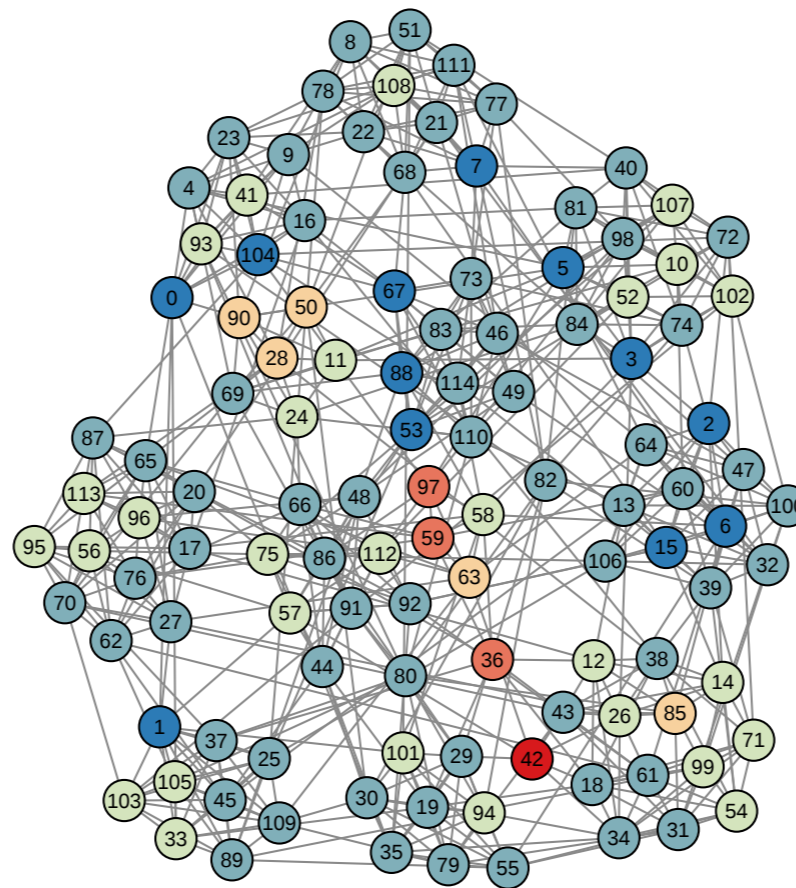
$$\text{NC}_r(i) = \exp(\mathbf{w}_i \cdot \mathbf{c}_i) \cdot k \cdot P(i) \propto \exp(\mathbf{w}_i \cdot \mathbf{c}_i) \cdot \text{deg}(i).$$

Compute approximate Node Conductance by DeepWalk vectors

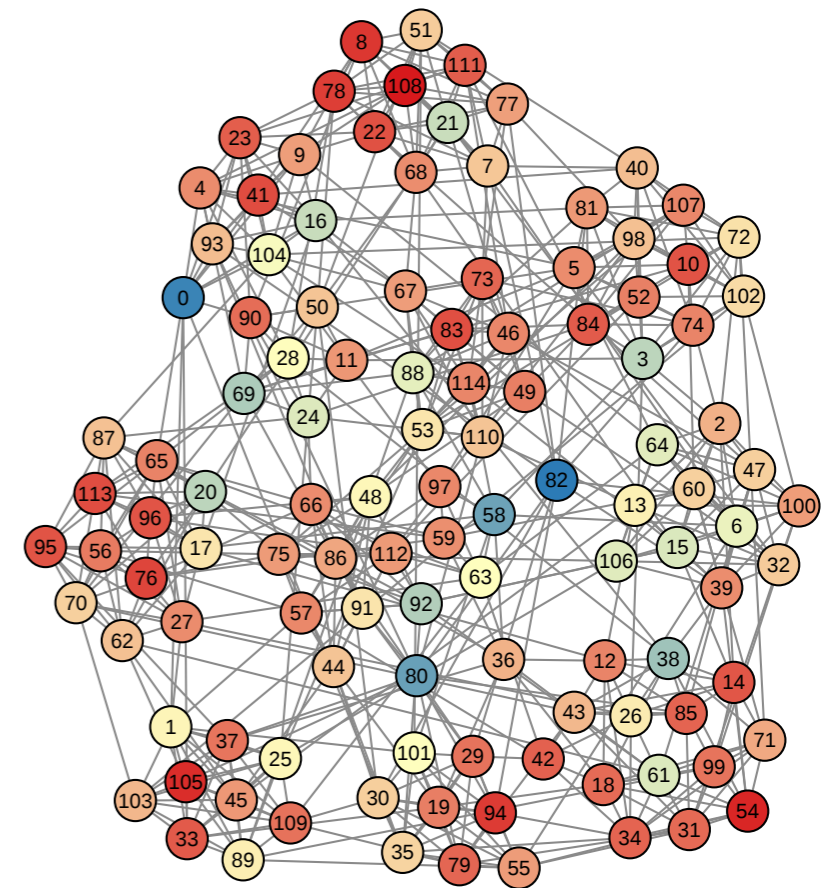
Visualization



Node Conductance



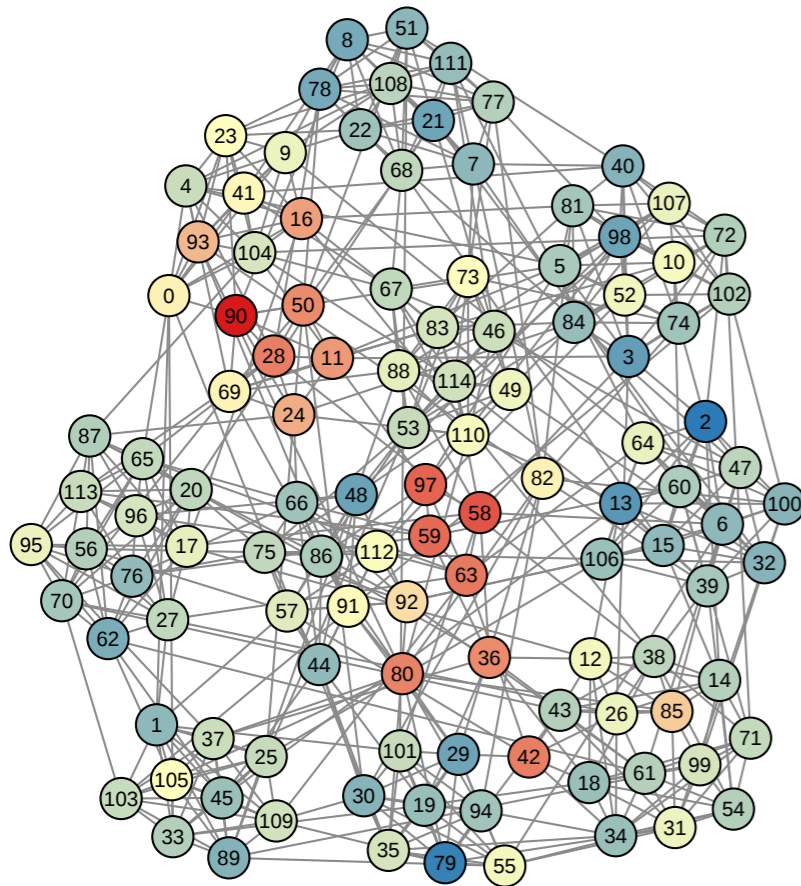
Degree



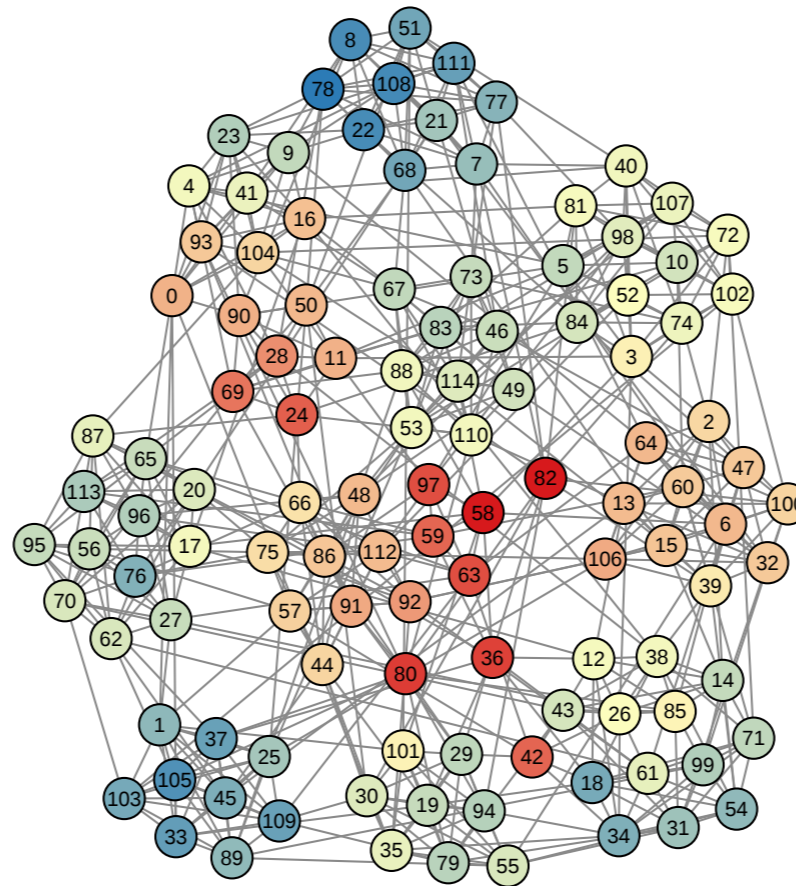
Betweenness

Node Conductance gives **low** value to nodes with **low degree** and **high betweenness** centrality.

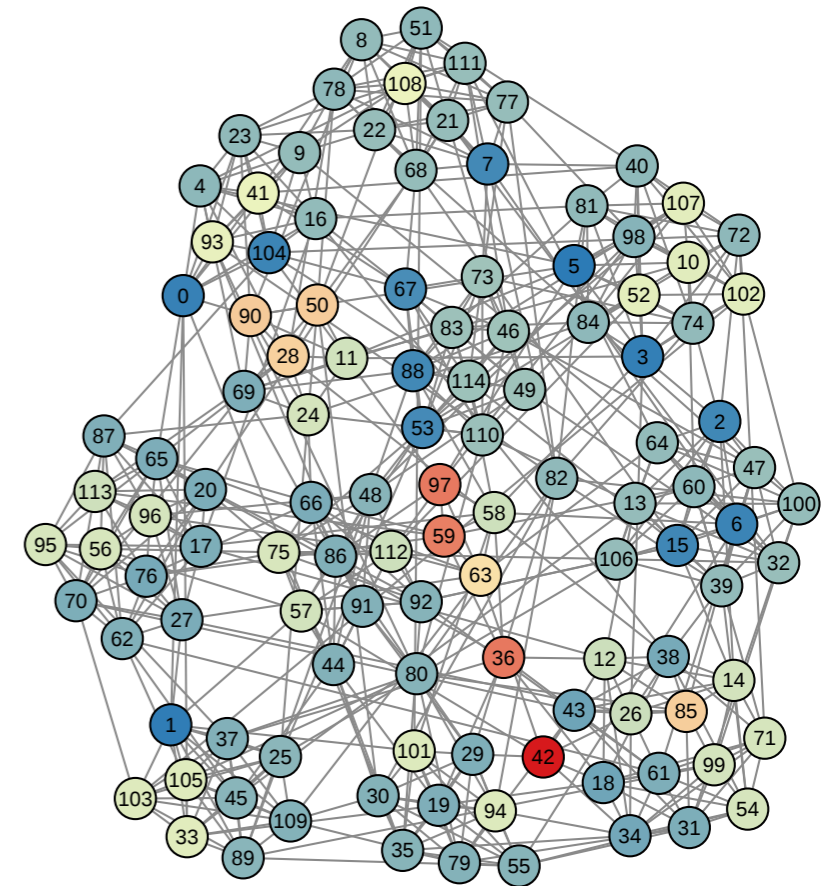
Visualization



Node Conductance (DW)



Node Conductance

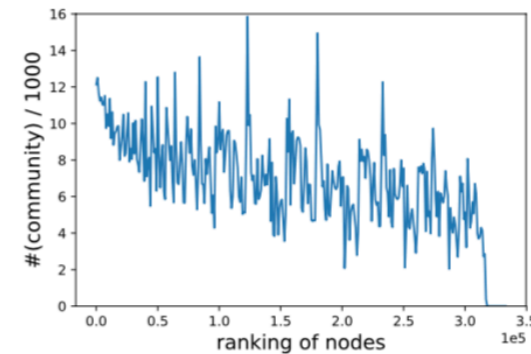


PageRank

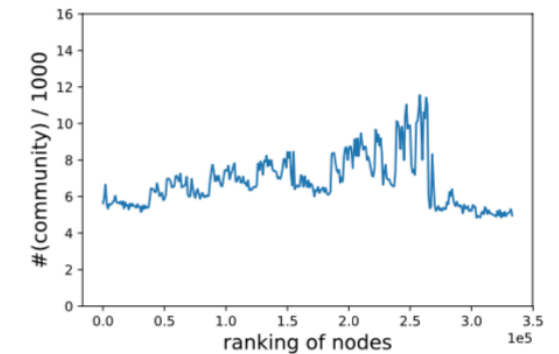
Node Conductance looks quite different with **PageRank**. **Node Conductance** and its **approximation** are similar.

Finding Nodes Spanning Several Communities

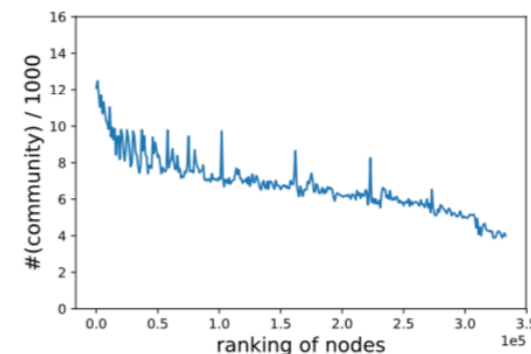
- Plot the communities numbers of nodes (y-axis) in the order of each centrality measure (x-axis).
- Node Conductance provides the smoothest curve comparing with the other five metrics.



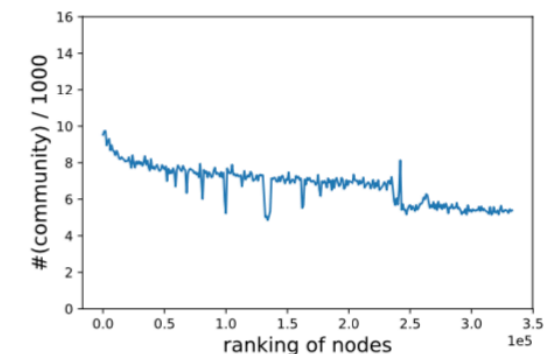
(a) Degree Centrality



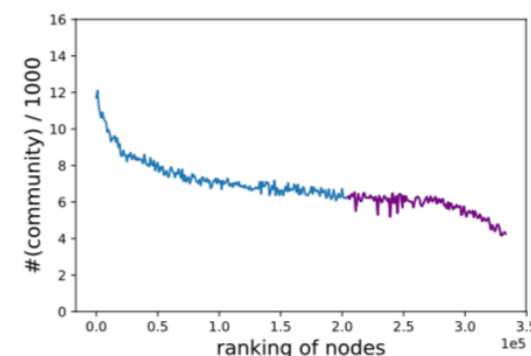
(b) Clustering Coefficient



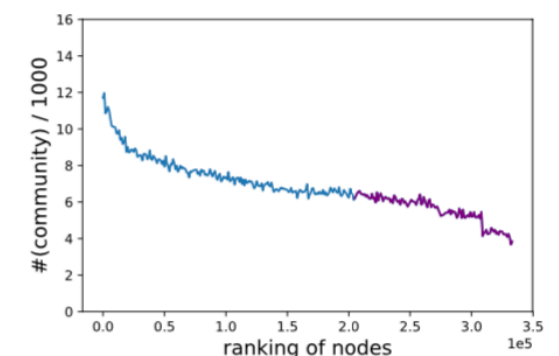
(c) Eigenvector Centrality



(d) Approximate Betweenness



(e) PageRank



(f) Node Conductance

Conclusion

- Intuition
 - the probability of revisiting the target node in a random walk
- Approximation
 - by the dot product of the input and output vectors
- Mining influential nodes



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THANKS!

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