

Nonlinear Laplacian for Digraphs and its Applications to Network Analysis

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Question: Can we develop spectral graph theory for digraphs?

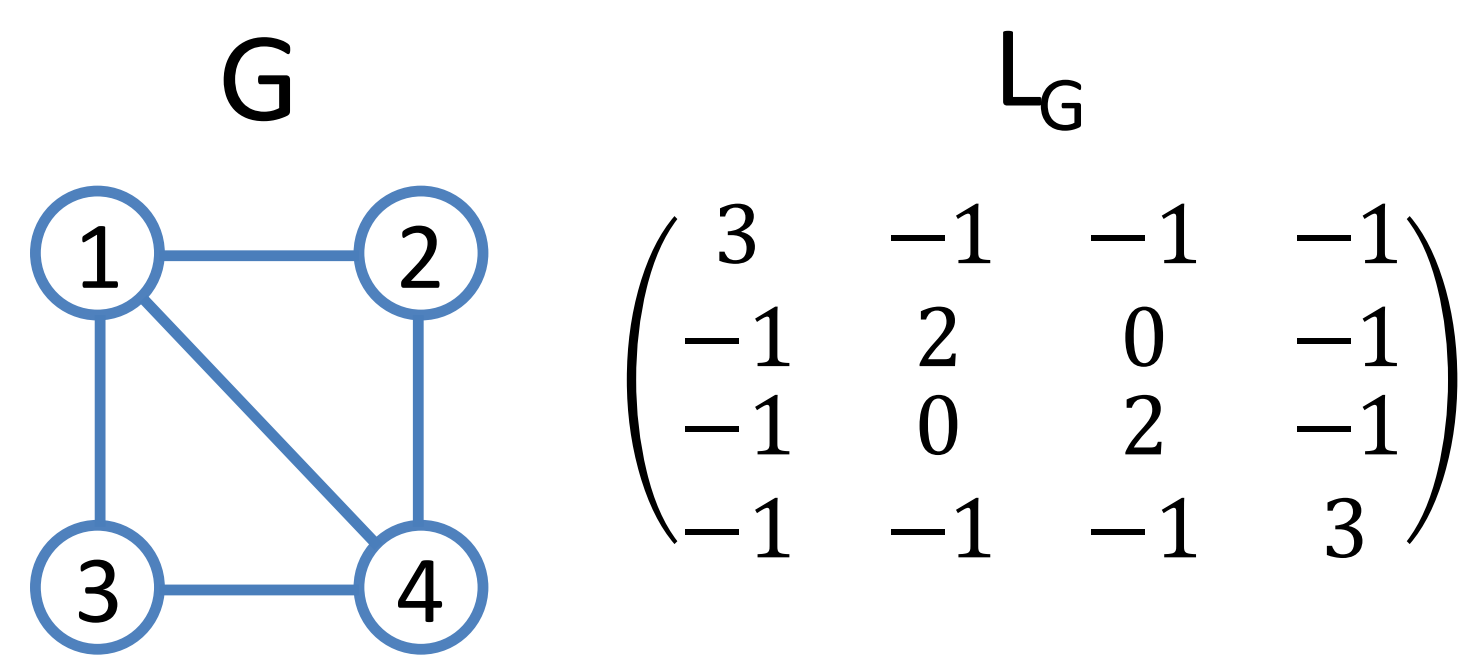
- Spectral graph theory analyzes graph properties via eigenpairs of associated matrices (in particular, Laplacian).
- Well established for undirected graphs.
- Extensions for digraphs are largely unexplored although many real-world networks are directed.

Definition

The Laplacian for an undirected graph G

Adjacency matrix A_G
Degree Matrix D_G

Laplacian $L_G := D_G - A_G$
Normalized Laplacian:
 $\mathcal{L}_G := D_G^{-1/2} L_G D_G^{-1/2}$



The Laplacian for a digraph G (proposed)

Laplacian $L_G: \mathbb{R}^V \rightarrow \mathbb{R}^V$ transforms $\mathbf{x} \in \mathbb{R}^V$ as follows:

- Construct an undirected graph H as follows:
- For each arc $u \rightarrow v$
 - If $x(u) \geq x(v)$, add an undirected edge $\{u, v\}$.
 - Otherwise, add self-loops to u and v .
- Output $L_H \mathbf{x}$.

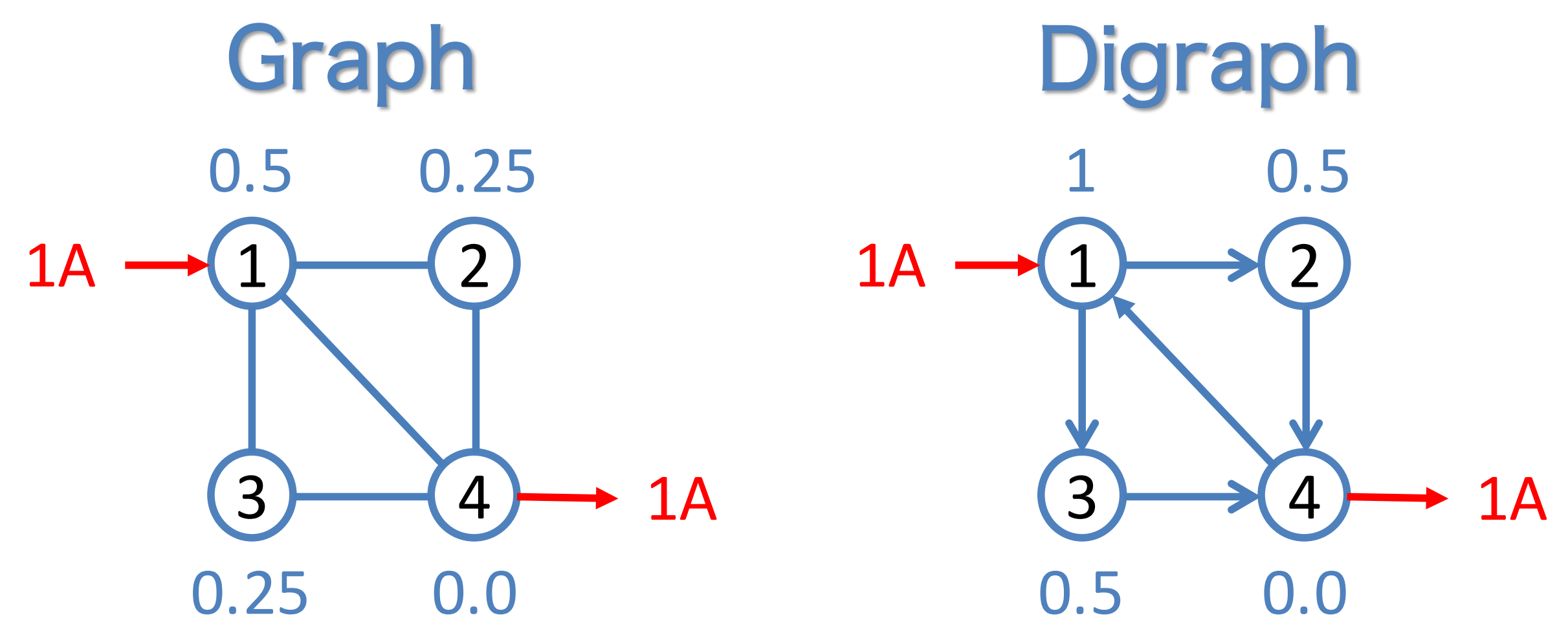
Normalized Laplacian: $\mathcal{L}_G: \mathbf{x} \mapsto D_G^{-1/2} L_G (D_G^{-1/2} \mathbf{x})$

Interpretation via electrical circuits

Regard G as an electrical circuit.

- Graph: edge = resistance of 1Ω
- Digraph: arc = diode of 1Ω (current flows only one way)

For each $u \in V$, flow a current of $\mathbf{b}(u)$ amperes to u .
The voltages $\mathbf{x} \in \mathbb{R}^V$ of vertices is given by $L_G(\mathbf{x}) = \mathbf{b}$.



Properties

(λ, \mathbf{v}) is an **eigenpair** of \mathcal{L}_G if $\mathcal{L}_G(\mathbf{v}) = \lambda \mathbf{v}$
Trivial eigenpair $(\lambda_1, \mathbf{v}_1)$ with $\lambda_1 = 0$.

For any subspace U of positive dimension, $\Pi_U \mathcal{L}_G$ has an eigenpair. (Π_U = Projection matrix to U)

\Rightarrow Another eigenvalue of \mathcal{L}_G exists by choosing $U = \mathbf{v}_1^\perp$.
Let λ_2 = the second smallest eigenvalue.

Algorithm

Computing λ_2 is (likely to be) NP-hard.

Suppose we start the diffusion process

$$d\mathbf{x} = -\Pi_U \mathcal{L}_G(\mathbf{x}) dt$$

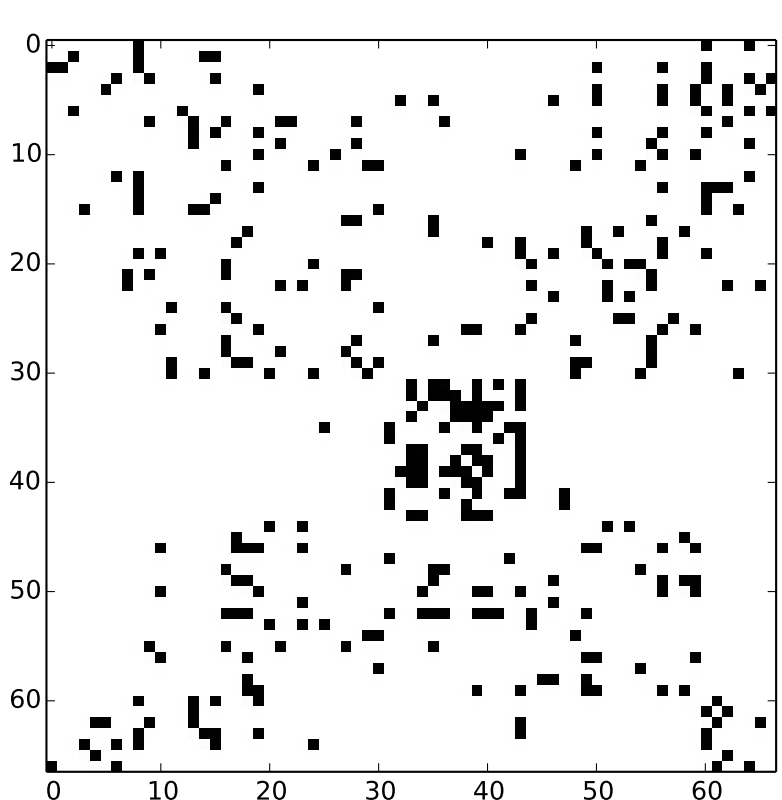
from a vector in the subspace $U = \mathbf{v}_1^\perp$.

- \mathbf{x} converges to an eigenvector orthogonal to \mathbf{v}_1 .
- Rayleigh quotient never increases during the process

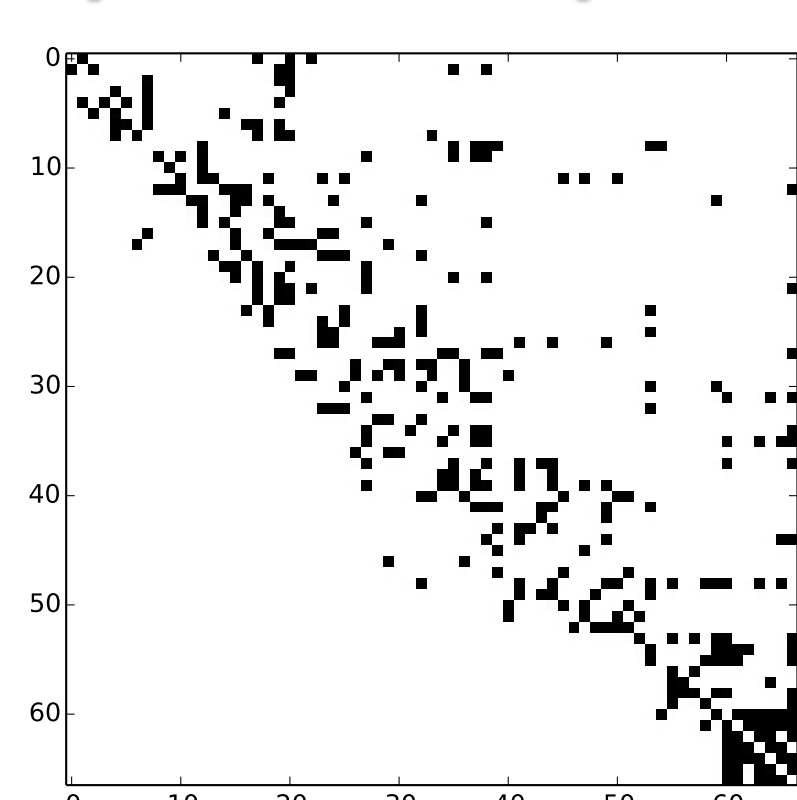
Visualization

Friendship network at a high school in Illinois
($u \rightarrow v$: u regards v as a friend)
Reorder vertices according to the eigenvector computed by the diffusion process

Chung's Laplacian



Proposed Laplacian



λ_2 is the minimum of $\sum_{u \rightarrow v} (\mathbf{x}(u) - \mathbf{x}(v))^2 \pi_u / d_u^+$ subject to $\mathbf{x} \neq 0, \mathbf{x} \perp \mathbf{v}_1$.

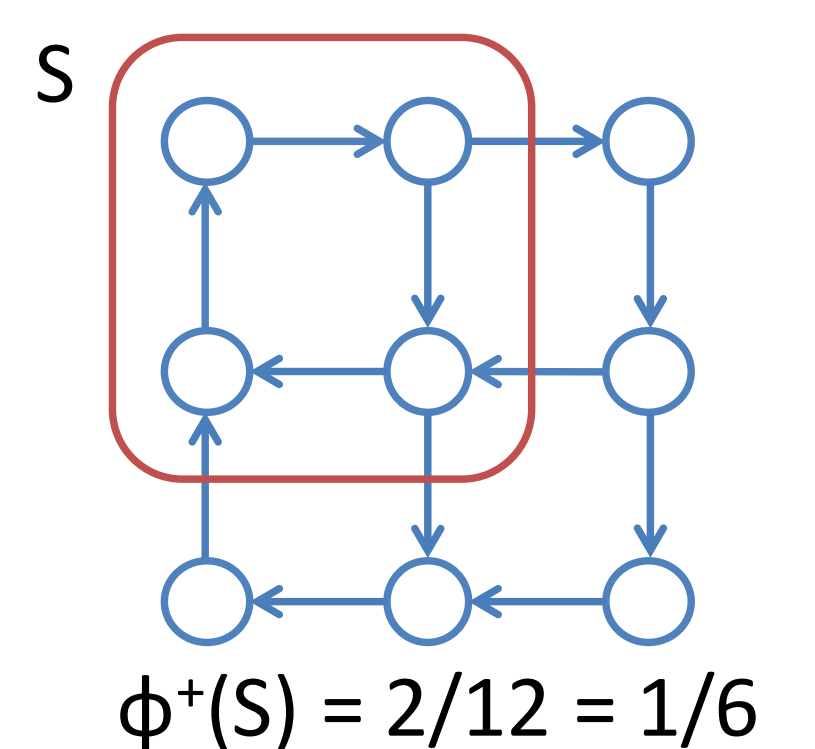
$$\sum_{u \rightarrow v} (\max(\mathbf{x}(u) - \mathbf{x}(v), 0))^2$$

Community Detection

(Directed) **conductance** $\phi^+(S)$ of S : $\frac{\text{cut}^+(S)}{\min(\text{vol}(S), \text{vol}(V-S))}$

$\text{vol}(S)$: total degree of $u \in S$.

$\text{cut}^+(S)$: # of arcs from S to $V-S$



Cheeger's inequality for digraphs:

$$\lambda_2 \leq \min_S \phi^+(S) \leq 2\sqrt{\lambda_2}$$

Conductance of the set of the first k vertices after reordering.

