

## Outline



## Background: <br> Electrical matrix equations

Geometry of a graph

## Adjacency matrix A



$$
A_{N \times N}=\left[\begin{array}{cccccc}
0 & 1 & 1 & 0 & 0 & 1 \\
1 & \mathrm{Q} & 1 & 0 & 1 & 1 \\
1 & 1 & 0 & 1 & 0 & 0 \\
0 & 0 & 1 & 0 & 1 & 0 \\
0 & 1 & 0 & 1 & 0 & 1 \\
1 & 1 & 0 & 0 & 1 & 0
\end{array}\right]
$$

For an undirected graph: $A=A^{\top}$ is symmetric
Number of neighbors of node $i$ is the degree: $\quad d_{i}=\sum_{k=1}^{N} a_{i k}$
if there is a link between node i and j , then $\mathrm{a}_{\mathrm{ij}}=1$
else $\mathrm{a}_{\mathrm{ij}}=0$ TUDelft

## Incidence matrix B



Col sum B is zero: $\quad u^{T} B=0$
where the all-one vector $u=(1,1, \ldots, 1)$

## Laplacian matrix Q



$$
Q_{N \times N}=\left[\begin{array}{cccccc}
3 & -1 & -1 & 0 & 0 & -1 \\
-1 & 4 & -1 & 0 & -1 & -1 \\
-1 & -1 & 3 & -1 & 0 & 0 \\
0 & 0 & -1 & 2 & -1 & 0 \\
0 & -1 & 0 & -1 & 3 & -1 \\
-1 & -1 & 0 & 0 & -1 & 3
\end{array}\right]
$$

$Q=B B^{T}=\Delta-A \quad$ Since $B B^{T}$ is symmetric, so are
$\Delta=\operatorname{diag}\left(\begin{array}{llll}d_{1} & d_{2} & \ldots & d_{N}\end{array}\right)$ $A$ and $Q$. Although $B$ specifies directions, $A$ and $Q$ lost this info here.

Basic property: $Q u=0$
$u$ is an eigenvector of $Q$ Belonging to eigenvalue $\mu=0$
$Q u=B B^{T} u=0 \quad$ because $0=u^{T} B=B^{T} u$ THDelft


## Function of network

- Usually, the function of a network is related to the transport of items over its underlying graph
- In man-made infrastructures: two major types of transport
- Item is a flow (e.g. electrical current, water, gas,...)
- Item is a packet (e.g. IP packet, car, container, postal letter,...)
- Flow equations (physical laws) determine transport (Maxwell equations (Kirchhoff \& Ohm), hydrodynamics, NavierStokes equation (turbulent, laminar flow equations, etc.)
- Protocols determine transport of packets (IP protocols and IETF standards, car traffic rules, etc.)


## Linear dynamics on networks

Linear dynamic process: "proportional to" ( $\sim$ ) graph of network

## Examples:



- water (or gas) flow ~ pressure
- displacement (in spring) ~ force
- heat flow ~ temperature
- electrical current $\sim$ voltage

| $\mathrm{X}=$ | Q | C |
| :--- | :--- | :--- |
| injected |  |  |
| nodal |  |  |
| current |  |  |
| vector |  |  |$\quad$| weighted |
| :--- |
| Laplacian |
| of the |
| graph |$\quad$| nodal |
| :--- |
| potential |
| vector |

## Pseudoinverse of the Laplacian (review)

The inverse of the current-voltage relation $x=Q v$
is the voltage-current relation $\quad v=Q^{\dagger} \boldsymbol{X}$
subject to $u^{T} x=0$ and $u^{T} v=0$
The spectral decomposition

$$
\tilde{Q}=\sum_{k=1}^{N-1} \tilde{\mu}_{k} z_{k} z_{k}^{T}
$$

allows us to compute the pseudoinverse (or Moore-Penrose inverse)

$$
Q^{\dagger}=\sum_{k=1}^{N-1} \frac{1}{\tilde{\mu}_{k}} z_{k} z_{k}^{T}
$$

The effective resistance $N \times N$ matrix is $\widetilde{\Omega}=u \zeta^{T}+\zeta u^{T}-2 Q^{\dagger}$, where the $N \times 1$ vector $\zeta=\left(Q_{11}^{\dagger}, Q_{22}^{\dagger}, \cdots, Q_{N N}^{\dagger}\right)$
An interesting graph metric is the effective graph resistance

$$
R_{G}=N u^{T} \zeta=N \operatorname{trace}\left(Q^{\dagger}\right)=N \sum_{k=1}^{N-1} \frac{1}{\mu_{k}}
$$



Inverses: $x=Q v \Leftrightarrow v=Q^{\dagger} \boldsymbol{x} \quad$ with voltage reference $u^{T} v=0$
$Q^{\dagger}$ : pseudoinverse of the weighted Laplacian obeying $Q Q^{\dagger}=Q^{\dagger} Q=I-\frac{1}{N} J$
$J=u u^{T}$ : all-one matrix
$u$ : all-one vector
Unit current injected in node $i$
nodal potential of $i$ $v_{i}=Q_{i i}^{\dagger}$

The best spreader is node $k$ with $\boldsymbol{Q}_{\boldsymbol{k} \boldsymbol{k}}^{\dagger} \leq \boldsymbol{Q}_{\boldsymbol{i} \boldsymbol{i}}^{\dagger}$ for $1 \leq i \leq N$

## Outline



## Background: Electrical matrix equations

## Geometry of a graph

## THDelft

## Three representations of a graph

Topology domain

$A_{N \times N}=\left[\begin{array}{cccccc}0 & 1 & 1 & 0 & 0 & 1 \\ 1 & 0 & 1 & 0 & 1 & 1 \\ 1 & 1 & 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 & 0 & 1 \\ 1 & 1 & 0 & 0 & 1 & 0\end{array}\right]$

Devriendt, K. and P. Van Mieghem, 2018, "The Simplex Geometry of Graphs Delft University of Technology, report20180717. (http://arxiv.org/abs/1807.06475).

Spectral domain
$A=A^{T}=X \Lambda X^{T}$
$X_{N \times N}$ : orthogonal eigenvector matrix
$\Lambda_{N \times N}$ : diagonal eigenvalue matrix

Geometric domain


Undirected graph on $N$ nodes
= simplex in Euclidean ( N -1)-dimensional space
is book comprises, in addition to auxiliary material, the research on which I have worked for over 50 years."
Cambridge
Cambridge

## What is a simplex?

Roughly: a simplex is generalization of a triangle to $N$ dimensions

Potential: Euclidean geometry is the oldest, mathematical theory

## Spectral decomposition weighted Laplacian

Spectral decomposition: $\quad Q=Z M Z^{T}$
where $M=\operatorname{diag}\left(\mu_{1}, \mu_{2}, \cdots, \mu_{N-1}, 0\right)$, because $Q u=0$
and the eigenvector matrix $Z$ obeys $Z^{\top} Z=Z Z^{\top}=I$ with structure
$\begin{aligned} \text { node } \underset{Z}{\Longrightarrow}\end{aligned}\left[\begin{array}{cccc}\left(z_{1}\right)_{1} & \left(z_{2}\right)_{1} & \cdots & \left(z_{N}\right)_{1} \\ \left(z_{1}\right)_{2} & \left(z_{2}\right)_{2} & \cdots & \left(z_{N}\right)_{2} \\ \vdots & \vdots & \ddots & \vdots \\ \left(z_{1}\right)_{N} & \left(z_{2}\right)_{N} & \cdots & \left(z_{N}\right)_{N}\end{array}\right]=\left[\begin{array}{cccc}\left(z_{1}\right)_{1} & \left(z_{2}\right)_{1} & \cdots & 1 / \sqrt{N} \\ \left(z_{1}\right)_{2} & \left(z_{2}\right)_{2} & \cdots & 1 / \sqrt{N} \\ \vdots & \vdots & \ddots & \vdots \\ \left(z_{1}\right)_{N} & \left(z_{2}\right)_{N} & \cdots & 1 / \sqrt{N}\end{array}\right]$

## Spectral decomposition weighted Laplacian (2)

Only for a positive semi-definite matrix, it holds that

$$
Q=Z M Z^{T}=(Z \sqrt{M})(Z \sqrt{M})^{T}
$$

The matrix $S=(Z \sqrt{M})^{T}$ obeys $Q=S^{T} S$ and has rank $N-1$ (row $N=0$ due to $\mu_{N}=0$ )

$$
S=\left[\begin{array}{cccc}
\sqrt{\mu_{1}}\left(z_{1}\right)_{1} & \sqrt{\mu_{1}}\left(z_{1}\right)_{2} & & \cdots \\
\sqrt{\mu_{1}}\left(z_{1}\right)_{N} \\
\sqrt{\mu_{2}}\left(z_{2}\right)_{1} & \sqrt{\mu_{2}}\left(z_{2}\right)_{2} & & \cdots \\
\vdots & \sqrt{\mu_{2}}\left(z_{2}\right)_{N} \\
\vdots & \vdots & \ddots & \vdots \\
\sqrt{\mu_{N-1}}\left(z_{N-1}\right)_{1} & \sqrt{\mu_{N-1}}\left(z_{N-1}\right)_{2} & \ldots & \sqrt{\mu_{N-1}}\left(z_{N-1}\right)_{N} \\
0 & 0 & & 0
\end{array}\right]
$$

$Q=\sum_{k=1}^{N-1} \mu_{k} Z_{k} Z_{k}^{T}$
THDelft

## Geometrical representation of a graph

$$
S=\left[\begin{array}{cccc}
\sqrt{\mu_{1}}\left(z_{1}\right)_{1} & \sqrt{\mu_{1}}\left(z_{1}\right)_{2} & \cdots & \sqrt{\mu_{1}}\left(z_{1}\right)_{N} \\
\sqrt{\mu_{2}}\left(z_{2}\right)_{1} & \sqrt{\mu_{2}}\left(z_{2}\right)_{2} & \cdots & \sqrt{\mu_{2}}\left(z_{2}\right)_{N} \\
\vdots & \vdots & \ddots & \vdots \\
\sqrt{\mu_{N-1}}\left(z_{N-1}\right)_{1} & \sqrt{\mu_{N-1}}\left(z_{N-1}\right)_{2} & \ldots & \sqrt{\mu_{N-1}}\left(z_{N-1}\right)_{N} \\
0 & 0 & & 0
\end{array}\right]
$$

The $i$-th column vector $s_{i}=\left(\sqrt{\mu_{1}}\left(z_{1}\right)_{i}, \sqrt{\mu_{2}}\left(z_{2}\right)_{i}, \cdots, \sqrt{\mu_{N}}\left(z_{N}\right)_{i}=0\right)$ represents a point $p_{i}$ in ( $N-1$ )-dim space (because $S$ has rank $N-1$ )


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Faces of a simplex
Each connected, undirected graph on $N$ nodes corresponds to 1 specific simplex in $N-1$ dimensions (Fiedler)
$V$ is a set of vertices of the simplex in $\mathbb{R}^{N-1}$, corresponding to a set of nodes in the graph $G$

1


A face $F_{V}=\left\{p \in \mathbb{R}^{N-1} \mid p=S x_{V}\right.$ with $\left(x_{V}\right)_{i} \geq 0$ and $\left.u^{T} x_{V}=1\right\}$
The vector $x_{V} \in \mathbb{R}^{N}$ is a barycentric coordinate with $\begin{cases}\left(x_{V}\right)_{i} \in \mathbb{R} & \text { if } i \in V \\ \left(x_{V}\right)_{i}=0 & \text { if } i \notin V\end{cases}$

## Centroids

$$
c_{V}=S \frac{u_{V}}{|V|} \text { is the centroid of face } F_{V} \text { with }\left(u_{V}\right)_{i}=1_{i \in V}
$$


a centroid of a face is a vector

centroid of simplex is origin
$u_{V}=u-u_{V} \rightarrow|V| c_{V}=S\left(u-u_{V}\right)=-(N-V) c_{V}$ TUWDelft

## Geometric representation of a graph

$\left\|s_{i}\right\|_{2}^{2}=d_{i} \quad\left\|s_{i}-s_{j}\right\|_{2}^{2}=\left(s_{i}-s_{j}\right)^{T}\left(s_{i}-s_{j}\right)=s_{i}^{T} s_{i}+s_{j}^{T} s_{j}-2 s_{i}^{T} s_{j}$

$$
\begin{aligned}
& =Q_{i i}+Q_{j j}-2 Q_{i j} \\
& =d_{i}+d_{j}+2 a_{i j} \text { for } i \neq j, \text { else zero }
\end{aligned}
$$



The matrix with off-diagonal elements $d_{i}+d_{j}+2 a_{i j}$ is a distance matrix (if the graph $G$ is connected)

The geometric graph representation is not unique (node relabeling changes $Z$ )
$s_{i}^{T} s_{j}=\sum_{k=1}^{N-1} \sqrt{\mu_{k}}\left(z_{k}\right)_{i} \sqrt{\mu_{k}}\left(z_{k}\right)_{j}=\sum_{k=1}^{N-1} \mu_{k}\left(z_{k} z_{k}^{T}\right)_{i j}=Q_{i j}$
$Q=\sum_{k=1}^{N-1} \mu_{k} z_{k} z_{k}^{T}$ and $Q=S^{T} S$
THDelft

## Geometry of a graph (dual representation)

Spectral decomposition: $Q^{\dagger}=Z M^{\dagger} Z^{T}=\left(Z \sqrt{M^{\dagger}}\right)\left(Z \sqrt{M^{\dagger}}\right)^{T}$
The matrix $S^{\dagger}=\left(Z \sqrt{M^{\dagger}}\right)^{T}$ has rank $N-1$ and $Q^{\dagger}=\left(S^{\dagger}\right)^{T} S^{\dagger}$
The $i$-th column vector $s_{i}^{\dagger}$ obeys


## Volume of simplex and inverse simplex of a graph

Volume of the simplex

$$
V_{G}=\frac{N}{(N-1)!} \sqrt{\xi}
$$

where the number of (weighted) spanning trees $\xi$ is $\xi=\frac{1}{N} \prod_{k=1}^{N-1} \mu_{k}$
Volume of the inverse simplex
$V_{G}^{+}=\frac{1}{(N-1)!\sqrt{\xi}}$
Hence: $\quad \frac{V_{G}}{V_{G}^{+}}=N \xi=\prod_{k=1}^{N-1} \mu_{k}$

## Steiner ellipsoid of simplex


projection $s_{1}^{T} \epsilon_{2}=\mu_{2}\left(z_{2}\right)_{1}$
semi-axis: $\left\|\epsilon_{2}\right\|=\sqrt{\frac{N}{N-1} \mu_{2}}$
volume:
$V_{\varepsilon_{S}}=\frac{\pi^{N / 2}}{\Gamma(N / 2+1)} \frac{N^{N / 2}}{(N-1)^{N / 2}} \sqrt{\prod_{k=1}^{N} \mu_{k}}$
Hence,

$$
V_{\varepsilon_{S}}^{2}=\frac{(N \pi)^{N}}{(\Gamma(N / 2+1))^{2}(N-1)^{N}} \prod_{k=1}^{N} \mu_{k}
$$

## altitude(s) in a simplex


$\left\|a_{\{2\}}\right\|^{2}=\frac{1}{Q_{22}^{\dagger}}$
Fiedler

$\left\|a_{2}^{+}\right\|^{2}=\frac{1}{d_{2}}$

The altitude from a vertex $s_{i}^{+}$to the complementary face $F_{\{l\}}^{+}$in the inverse simplex (dual graph representation) has a length equal to the inverse degree of node i
recall that $Q_{i i}^{\dagger}=v_{i}$ (nodal potential, best spreader)
TUD ${ }^{T}$ fft


Metrics $\sqrt{\omega_{i j}}$ and $\omega_{i j}$
$\left\|s_{i}^{\dagger}-s_{j}^{\dagger}\right\|_{2}=\sqrt{\omega_{i j}}$
the Euclidean distance between vertices of inverse simplex $S^{\dagger}$

vertices of $\mathrm{S}^{\dagger}$ are an embedding of nodes of the graph G
according to the metric $\sqrt{\omega_{\mathrm{ij}}}$ (a.o. obeying the triangle inequality)

Also $\left\|s_{i}^{\dagger}-s_{j}^{\dagger}\right\|_{2}^{2}=\omega_{i j}$ is a metric
Inverse simplex $\mathrm{S}^{\dagger}$ of the graph G with positive link weights is hyperacute

## Generalization metrics

$Q^{\dagger}$ is the Gramm matrix of a hyperacute simplex $S^{+} \rightarrow$ determines a metric
$m_{i j}^{(f)}=\left(e_{i}-e_{j}\right)^{T}(f(Q))^{\dagger}\left(e_{i}-e_{j}\right)$ is a metric when $f(Q)$ is a Laplace matrix
"Statistical physics" metrics on a graph
$m_{i j}^{(s t a t(g))}=\left(e_{i}-e_{j}\right)^{T}\left(e^{\left(Q+p Q_{K}\right) t}-I+g Q_{K}\right)^{\dagger}\left(e_{i}-e_{j}\right)=\sum_{k=1}^{N-1} \frac{\left.\left(\left(z_{k}\right)-\left(z_{k}\right)\right)_{j}\right)^{2}}{e^{\left(\mu_{k}+p N\right) t}+g}$
with $t=\frac{1}{k_{B} T}$ and $p N=-E_{f}$ (chemical potential or Fermi energy)
$\mathrm{g}=-1 \rightarrow$ Bose-Einstein
$\mathrm{g}=0 \rightarrow$ Maxwell-Boltzmann
$\mathrm{g}=1 \rightarrow$ Fermi-Dirac

## Summary

- Linearity between process and graph naturally leads to the weighted Laplacian $Q$ and its pseudoinverse $Q^{\dagger}$
- Spectral decomposition of the weighted Laplacian $Q$ and its pseudoinverse $Q^{\dagger}$ provides an $N-1$ dimensional simplex representation of each graph,
- allowing computations in the $N-1$ dim. Euclidean space (in which a distance/norm is defined)
- geometry for (undirected) graphs
- Open: "Which network problems are best solved in the simplex representation?"

| Books |  |
| :---: | :---: |
| Performance Analysis of Complex Networks and Systems Piet Van Mieghem <br> Graph Spectra for Complex Networks Piet Van Mieghem | Data Communications Networking |
| Articles: http://www.nas.ewi.tudelft.nl | 30 |
|  | TUDelft |



