

# Resistive Networks

CS292 F  
May 3, 2021  
Lecture 9



# LANCZOS

$$\boxed{Q_k^T} \boxed{A} \boxed{Q_k} = \begin{array}{|c|} \hline T_k \\ \hline \end{array} \quad Q_k^T Q_k = I_k$$

Then  $V_k^T T_k V_k = \Theta$  evals of  $T_k$ .

$$\Theta = \text{Diag}(\theta_1, \theta_2, \theta_3, \dots, \theta_k)$$

$k$  Ritz values approximate  
 $k$  of the  $n$  evals of  $A$ .

The extreme Ritz values converge  
quickly to the extreme  
evals of  $A$ .

Disadvantage: How do you get  
the middle evals?

As  $k$  increases.

# Spectral transforms for interior eigenvalues.

Say we want an eigenvalue near  $\sigma$ .

Matrix  $A$ , want  $Aw = \lambda w$ ,  $\lambda$  near  $\sigma$ .

Look at  $A - \sigma I$ . This is symmetric.  
(Not positive definite)

If  $Aw = \lambda w$  then  $(A - \sigma I)w = (\lambda - \sigma)w$ .

And  $A^{-1}w = \frac{1}{\lambda}w$  and  $(A - \sigma I)^{-1}w = \frac{1}{\lambda - \sigma}w$

and  $\frac{1}{\lambda - \sigma}$  is an  
"external" eigenvalue of  $A$   
"extreme"

APPLY LANCZOS TO  $(A - \sigma I)^{-1}$ .

NEED TO MULTIPLY  $v = (A - \sigma I)^{-1}g_k$

$\Rightarrow$  solve  $(A - \sigma I)v = g_k$  for  $v$ .

STILL SPARSE, BUT NOT POSITIVE DEF.

$$A_w = \tau_w$$

$$Ax = b \quad \left( \begin{smallmatrix} \text{given} \\ b \end{smallmatrix} \right)$$

## SOLVING SYSTEMS OF EQUATIONS

$$Lx = b$$

↑ (weighted) Laplacian.  
Symmetric  
positive semi-definite  
not completely general.

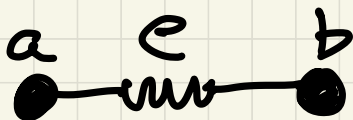
## APPLICATIONS:

PDE's; Electrical circuits;  
data analysis; graph algorithms  
(21<sup>st</sup> century)

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There are surprisingly efficient  
algorithms for  $Lx = b$ , Laplacian.  
(ACTIVE RESEARCH)

# NETWORKS OF RESISTORS



resistance  $r(a, b)$   
or  $r(e)$   
(in ohms)

current from  $a$  to  $b$  (in amperes)

$$i(a, b) = -i(b, a)$$

potential at  $a$  or at  $b$  (voltage)

$v(a) - v(b)$  potential  
difference.

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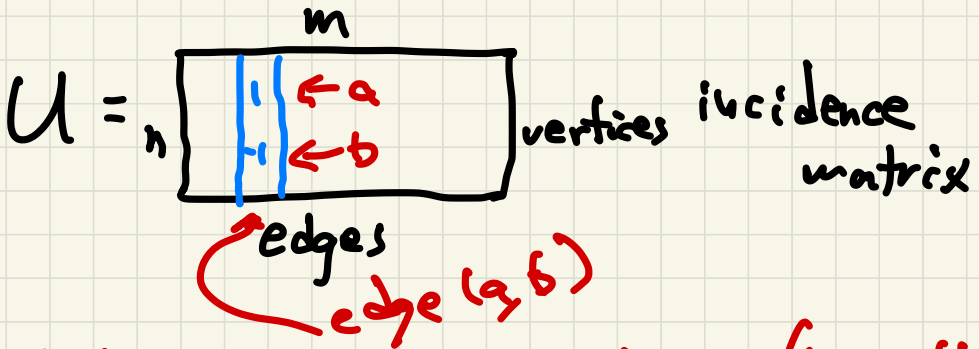
Ohm's law:  $V = IR$   $v(a) - v(b) = i(a, b) r(a, b)$

Kirchoff's  
current law:  $\forall a \in V : \sum_{(a, b) \in E} i(a, b) = 0$

Kirchoff's  
voltage law: sum of  $v(a) - v(b)$  around  
any cycle is 0. Automatic.

Resistive network  $\equiv$  weighted graph.

DEF



Know that  $UU^T = L$  is a Laplacian (unweighted)  
 (now  $U$  is the transpose of Spielman's)

$$R = \begin{matrix} m \\ \text{edges} \end{matrix} \begin{matrix} n \\ \text{vertices} \end{matrix} = \text{Diag}(r(e))_{e \in E} \quad v = \begin{pmatrix} v(1) \\ \vdots \\ v(n) \end{pmatrix} \begin{matrix} \text{vertex} \\ \text{potential} \\ (n) \end{matrix}$$

$$i = \begin{pmatrix} i(1) \\ \vdots \\ i(m) \end{pmatrix} = \begin{matrix} \text{currents} \\ \text{of edges} \end{matrix} (m)$$

$$\text{OHM: } Ri = U^T V \Rightarrow i = R^{-1} U^T V$$

$$\text{KIRCHOFF: } \ddot{U} \ddot{i} = Q \quad U i = i_{\text{ext}}$$

Consider an externally applied current at each vertex (or at some vertices):  $i_{\text{ext}}^{(v)}$

so  $i_{\text{ext}}$  is an  $n$ -vector with  $\mathbf{1}^T i_{\text{ext}} = 0$

$$Ri = U^T v \Rightarrow i = R^{-1} U^T v$$

$$U_i = i_{\text{ext}}$$

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$$U R^{-1} U^T v = i_{\text{ext}}$$

weighted Laplacian  $L = U R^{-1} U^T$

$R^{-1} = \text{Diag}\left(\frac{1}{r(e)}\right)_{e \in E}$  Define  $c(e) = \frac{1}{r(e)}$   
 inverse resistance (mhos)

So  $L v = i_{\text{ext}}$

DEF A vector  $v$  of vertex labels is harmonic at vertex  $a$  of a <sup>weighted</sup> graph

$$\equiv v(a) = \frac{\sum_{ab \in E} c(a,b) v(b)}{\sum_{ab \in E} c(a,b)}$$

RECALL we defined  $d(a) = \sum_{ab \in E} c(a,b)$

If  $Lv = i_{\text{ext}}$  then

$v$  is harmonic at every vertex with no external current.

$Lv = i_{\text{ext}}$ . we said  $\mathbf{1}^T i_{\text{ext}} = 0$ .

$L(v+1) = Lv + L1 = Lv$  so wlog

we can take  $v$  to have sum 0 as well,

$\mathbf{1}^T v = 0$ . (replace  $v$  by  $v - \frac{1}{n} \mathbf{1}^T v$ )



$$\mathbf{1}^T \mathbf{v} = 0 \quad \mathbf{1}^T \mathbf{i}_{\text{ext}} = 0$$

$L\mathbf{v} = \mathbf{i}_{\text{ext}}$  : can think of

$$L : \mathbb{R}^V / \mathbb{1} \rightarrow \mathbb{R}^V / \mathbb{1}$$

as an  $n-1$  dimensional linear transformation, which is nonsingular if the graph is connected.

AND IN FACT THE SOLUTION TO  $L\mathbf{v} = \mathbf{i}_{\text{ext}}$  IS:

$$\mathbf{v} = L^+ \mathbf{i}_{\text{ext}}$$

$\nwarrow$  pseudoinverse of  $L$ .