

Fabio Pasqualetti



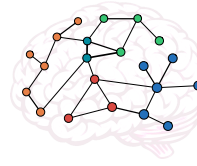
Department of Mechanical Engineering  
University of California, Riverside

Analysis and Control of Complex Networks  
Philadelphia, PA, July 9 2019

## In this talk

- 1 Controllability bounds for complex networks
  - Many complex networks are difficult to control
  - Some (anisotropic) networks are easy to control
- 2 Design networks with prescribed controllability profiles
  - Networks with diagonal controllability Gramian
- 3 Controllability tradeoffs in complex networks
  - Controllability can be improved only at the expenses of robustness

## Complex networks



Brain network



Power network



Social network

Analyze, predict, and control **dynamics** over **large networks**

- Asymptotic properties, scaling laws, tradeoffs
- Controllability, resilience, fragility, synchronization, spectrum ...  
... (natural/social/technological systems)
- Algebraic graph theory, systems theory, sparse matrix analysis

## Problem formulation

### System description

$$x(t+1) = Ax(t) + Bu(t) \quad (\text{or } \dot{x}(t) = Ax(t) + Bu(t))$$

- The matrix  $A$  is sparse (interaction graph)
- $B = [e_{i_1} \ e_{i_2} \ \cdots \ e_{i_m}]$  where  $e_i = \underbrace{[0 \ 0 \ \cdots \ 0 \ 1 \ 0 \ \cdots \ 0]^T}_{i\text{-th component equals 1}}$

### Controllability definitions

Controllability Gramian:  $\mathcal{W}_T = \sum_{t=0}^{T-1} A^t B B^T A^t$

**Small**  $\lambda_{\min}(\mathcal{W}_T) \Leftrightarrow$  **Small controllability degree**

**Large**  $\lambda_{\min}(\mathcal{W}_T) \Leftrightarrow$  **Large controllability degree**

## Few nodes cannot control symmetric networks

### Upper bound on controllability degree

$$\lambda_{\min}(\mathcal{W}_T) \leq \frac{\mu^{2\lceil n(\mu)/m \rceil - 2}}{1 - \mu^2}$$

- $0 < \mu < 1$
- $n(\mu) = |\{\lambda : \lambda \in \text{spec}(A), |\lambda| \leq \mu\}|$

Typically,  $n(\mu)$  grows linearly with the network cardinality:

- 1 For fixed number of control nodes, the controllability degree **decreases exponentially** with the network cardinality
- 2 For fixed controllability degree, the number of control nodes **must grow linearly** with the network cardinality
- 3 Counterintuitive result w.r.t. structural controllability

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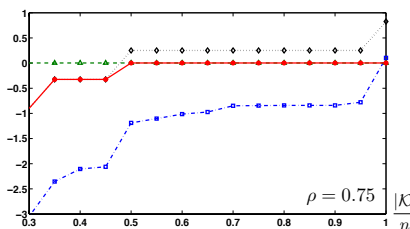
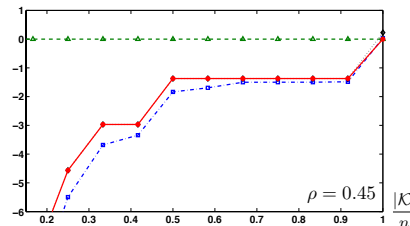
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## Network matrix Schur stable $\Rightarrow n(\mu) = n$

### Upper bound with $\mu = \lambda_{\max}$

$$\lambda_{\min}(\mathcal{W}_T) \leq \frac{\lambda_{\max}^{2\lceil n/m \rceil - 2}}{1 - \lambda_{\max}^2(A)}$$

$$A = \frac{\rho}{3} \begin{bmatrix} 1 & 1 & 0 & \cdots & 1 \\ 1 & 1 & 1 & \cdots & 0 \\ \vdots & \ddots & \ddots & \ddots & \vdots \\ 0 & \cdots & 1 & 1 & 1 \\ 1 & \cdots & 0 & 1 & 1 \end{bmatrix}$$



The smaller the spectral radius, the tighter the bound

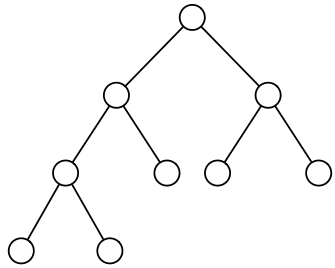
## Can few nodes control asymmetric networks?

### Upper bound on controllability degree

$$\lambda_{\min}(\mathcal{W}_{K,T}) \leq \text{cond}^2(V) \frac{\mu^{2\lceil n(\mu)/m \rceil - 2}}{1 - \mu^2}$$

- $V$  is an eigenvector matrix of  $A$
- $\text{cond}(V) = \sigma_{\max}(V)/\sigma_{\min}(V)$  is the condition number of  $V$

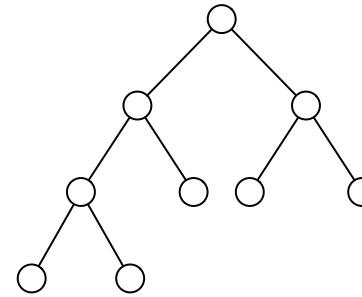
- 1 If  $\text{cond}(V)$  remains bounded with the network dimension ...  
... then the network remains difficult to control
- 2 If  $\text{cond}(V)$  grows with the network dimension ...  
... then the network may be easy to control



- $A = [a_{ij}]$
- $0 < a_{\min} \leq a_{ij} \leq a_{\max} < \infty$
- $V =$  eigenvector matrix
- How does  $\text{cond}(V)$  scale with  $n$ ?

Facts:

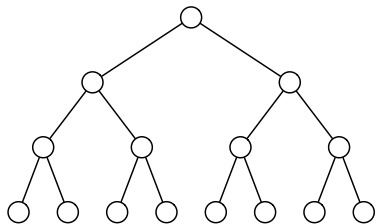
- $A$  acyclic  $\Rightarrow$  there exists  $D = \text{diag}(d_{ij})$  s.t.  $\tilde{A} = DAD^{-1}$  symmetric
- $\tilde{A}$  symmetric  $\Rightarrow$  there exists  $\tilde{A}W = W\Lambda$  with  $\text{cond}(W) = 1$
- $\tilde{A}W = W\Lambda \Rightarrow A(D^{-1}W) = (D^{-1}W)\Lambda \Rightarrow V = D^{-1}W$
- $\text{cond}(V) = \|V\| \|V^{-1}\| \leq \|D\| \|D^{-1}\| \|W\| \|W^{-1}\| \leq \text{cond}(D) \underbrace{\text{cond}(W)}_{=1}$



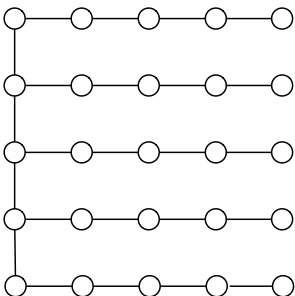
- $0 < a_{\min} \leq a_{ij} \leq a_{\max} < \infty$
- $D = \text{diag}(d_{ij})$  s.t.  $DAD^{-1}$  symmetric
- $DAD^{-1}W = W\Lambda$ ,  $\text{cond}(W) = 1$
- $\lambda_{\min}(\mathcal{W}_{\mathcal{K}, \tau}) \leq \text{cond}^2(D) \frac{\mu^{2\lceil n(\mu)/m \rceil - 2}}{1 - \mu^2}$
- $\text{cond}(D)$  depends on network diameter
- $\text{cond}(D) \leq \left( \frac{a_{\max}}{a_{\min}} \right)^{\text{diameter}}$

### Controllability of acyclic networks

Acyclic networks with sublinear diameter ( $o(n)$ ) are difficult to control  
(by constant number of control nodes)



Diameter  $\in O(\log n)$



Diameter  $\in O(\sqrt{n})$

### Networks that are difficult to control

- Symmetric networks
- Acyclic networks with  $o(n)$  diameter
- Networks with constant condition number

Are there networks that are easy to control?

## 1 Controllability bounds for complex networks

- Many complex networks are difficult to control
- Some (anisotropic) networks are easy to control

## 2 Design networks with prescribed controllability profiles

- Networks with diagonal controllability Gramian

## 3 Controllability tradeoffs in complex networks

- Controllability can be improved only at the expenses of robustness

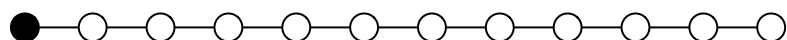


$$A = \begin{bmatrix} 0 & 0 & 0 & \dots & 0 \\ 1 & 0 & 0 & \dots & 0 \\ 0 & 1 & \ddots & \ddots & \vdots \\ \vdots & \ddots & \ddots & 0 & 0 \\ 0 & \dots & \dots & 1 & 0 \end{bmatrix}$$

- The network is stable with identity controllability matrix
- The controllability degree is independent of the network cardinality

## Controllability of Toeplitz line networks

#1



$$A = \begin{bmatrix} a & b & 0 & \dots & 0 \\ c & a & b & \dots & 0 \\ 0 & c & \ddots & \ddots & \vdots \\ \vdots & \ddots & \ddots & a & b \\ 0 & \dots & \dots & c & a \end{bmatrix}$$

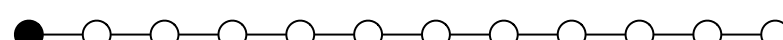
### Controllability of line networks

A toeplitz line network is **easy** to control if one of the following holds:

- $\frac{a(b+c)}{4bc} \leq 1$  and  $1 < (b-c)^2(1 - \frac{a^2}{4bc})$
- $\frac{a(b+c)}{4bc} > 1$  and  $1 \leq c + b - a$

## Controllability of Toeplitz line networks

#2



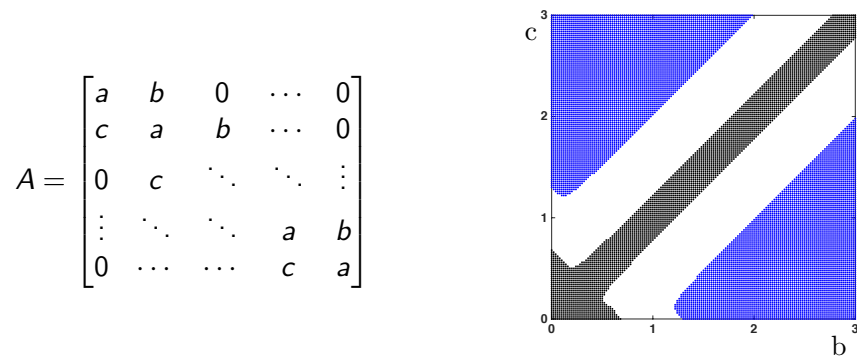
### Controllability of line networks

A toeplitz line network is **difficult** to control if one of the following holds:

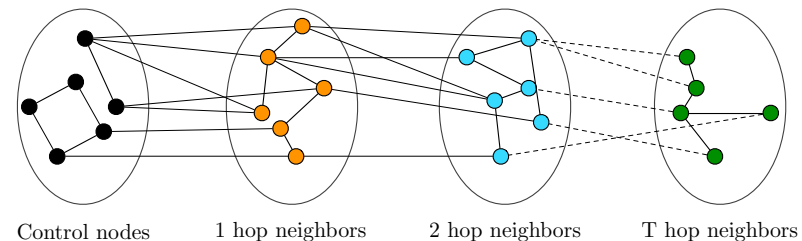
- $a + b + c < 1$
- $ab < c < \frac{b}{a}$  and  $a \geq \sqrt{bc}$
- $ba^{n_a} < c < \frac{b}{a^{n_a}}$  and  $a < \sqrt{bc}$  with  $n_a = \frac{2}{\pi} \arccos\left(\frac{-a}{\sqrt{bc}}\right) - 1$

- Easy to control  $\Rightarrow$  Energy is bounded independent of cardinality
- Difficult to control  $\Rightarrow$  Energy increases exponentially with cardinality  
(with constant number of control nodes)



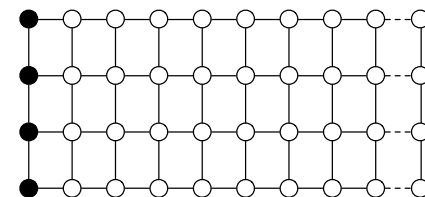


- 1 Given  $a$ , if  $c$  is **sufficiently larger than  $b$** , then line networks are controlled with finite energy, independent of the network dimension
- 2 **Spatial instability** is exploited to have high controllability degree  
(large non-normality degree)
- 3 The width of the uncontrollable region depends on the parameter  $a$



Easy to control line-like networks:

- forward matrices of full rank
- $k$ -neighbors  $\leq (k-1)$ -neighbors
- condition on matrices:  
 $\|C_i^\dagger\|_2^{\max} (1 + \|D_i\|_2^{\max} + \|B_i\|_2^{\max}) < 1$



## Then...

### Networks that are difficult to control

- Symmetric networks
- Acyclic networks with  $o(n)$  diameter
- Networks with constant condition number

### Networks that are easy to control

- Sufficiently anisotropic networks with triangular controllability matrix
  - Toeplitz line networks
  - block tridiagonal networks
- Sufficiently anisotropic cycle networks

Are there *interesting* networks that are easy to control?

## Outline

- 1 Controllability bounds for complex networks
  - Many complex networks are difficult to control
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- 2 Design networks with prescribed controllability profiles
  - Networks with diagonal controllability Gramian
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## Networks with prescribed controllability Gramian #1

Can we design networks with prescribed Gramian (i.e., contr. degree)?

- Focus on the case where  $\mathcal{W}$  is **diagonal** (restrictive yet insightful)
- Continuous-time network dynamics

### Sign-skew-symmetric network

The network  $\mathcal{G} = (\mathcal{V}, \mathcal{E})$  with input nodes  $\mathcal{S}$  is **sign-skew-symmetric** if

- $a_{ij}a_{ji} < 0$  if  $(i, j) \in \mathcal{E}$ , and  $a_{ij} = 0$  otherwise;
- $a_{ii} < 0$  if  $i \in \mathcal{S}$ , and  $a_{ii} = 0$  otherwise.

### Control impact along a path

For a sign-skew-symmetric network  $\mathcal{G} = (\mathcal{V}, \mathcal{E})$  with input nodes  $\mathcal{S}$ ,

- the **control impact** along the path  $(i_1, i_2, \dots, i_p)$ , with  $i_i \in \mathcal{S}$  is

$$\beta_{i_1, \dots, i_p} = \frac{1}{|a_{i_1 i_1}|} \left| \frac{a_{i_2 i_1}}{a_{i_1 i_2}} \right| \left| \frac{a_{i_3 i_2}}{a_{i_2 i_3}} \right| \dots \left| \frac{a_{i_p i_{p-1}}}{a_{i_{p-1} i_p}} \right|$$

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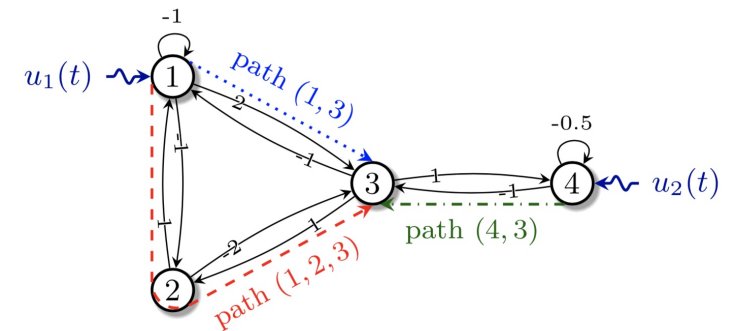
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## Networks with diagonal controllability Gramian #2

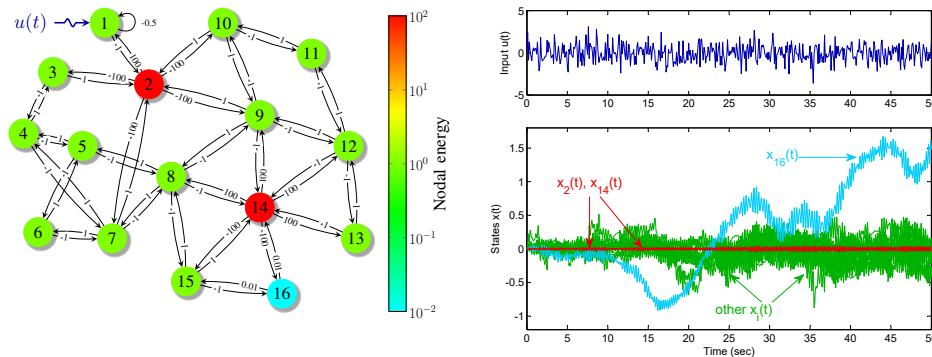


### Uniformly input-connected network

The network  $\mathcal{G} = (\mathcal{V}, \mathcal{E})$  with input nodes  $\mathcal{S}$  is **uniformly input-connected** if

- it is sign-skew-symmetric, and
- for every node  $i$ , all control impacts to  $i$  are equal to  $\beta_i \in \mathbb{R}_{>0}$ .

Stable + uniformly input-connected  $\iff \mathcal{W} = \text{diag}(\beta_1, \dots, \beta_n)$



Select weights to assign control impacts  $\Rightarrow$  Control energy by design

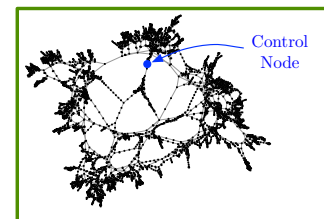
- 1 Large class of continuous-time networks is easy to control
  - That is, control energy may not increase with network cardinality
- 2 Energy to control a node may be independent of graphical distance
  - That is, distant nodes may require little energy to change their state
- 3 We can create networks with desired nodal energies
  - That is, we can secure selected nodes from exogenous disturbances

Is there a price to pay for easy controllability?

## Outline

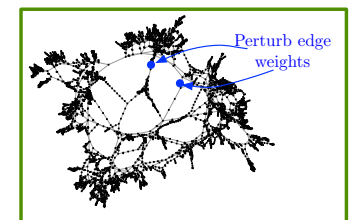
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## Two (independent?) properties of dynamical networks



For a dynamic network, **controllability** is the ability to drive the state to arbitrary configurations through external controls

For a dynamic network, **fragility** refers to the ability to maintain a stable behavior in the face of small edge perturbations



## A definition of network fragility

### Network dynamics

$$\dot{x}/x^+ = Ax + Bu$$

### Fragility definition

Stability radius:  $r(A) = \min\{\|\Delta\| : A + \Delta \text{ is not Hurwitz stable}\}$

**Small**  $r(A)$   $\Leftrightarrow$  **Large fragility degree**

**Large**  $r(A)$   $\Leftrightarrow$  **Small fragility degree**

- Small  $r(A) \Rightarrow$  easy to destabilize network
- Perturbation  $\Delta$  is complex and non-structured
- Equivalently,  $r(A) = \min_{\omega \in \mathbb{R}} \sigma_{\min}(i\omega I - A) = \frac{1}{\max_{\omega \in \mathbb{R}} \|(i\omega I - A)^{-1}\|}$

## Why are natural and technological networks fragile? #1

Empirical evidence and theories suggest that large networks are fragile ...  
... (ecological [Nat.'72], power [Chaos'07], neural [NeuralComp.'14])

### Controllability vs fragility in networks

$$\lambda_{\min}(\mathcal{W}) \leq \bar{\lambda}(\mathcal{W}) \leq \frac{n_c}{n} \left( 1 + \frac{4\|A - A^T\|}{3\pi} \frac{1}{r(A)} \right) \frac{1}{r(A)}$$

- $n$  = cardinality of the network
- $n_c$  = number of control nodes
- $\bar{\lambda}(\mathcal{W})$  = average eigenvalue of Gramian  $\mathcal{W}_{\infty}$

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## Why are natural and technological networks fragile? #2

### Controllability vs fragility in networks

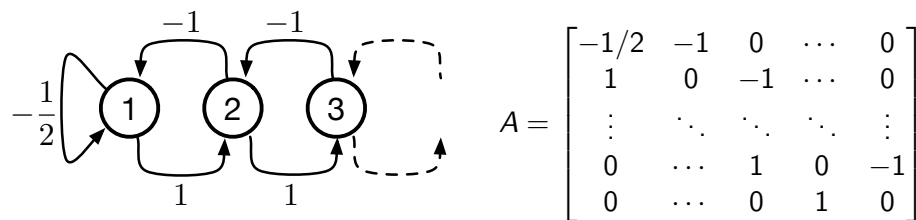
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- 1 few control nodes  $n_c \Rightarrow$  large control energy
- 2 large stability radius  $r(A) \Rightarrow$  small Gramian eigenvalues ...  
... (robust networks cannot be easy to control)
- 3 if cardinality  $n$  increases  $\Rightarrow$  controllability and/or robustness decrease  
... (assuming terms in bracket are sublinear in  $n$ )

### For symmetric networks:

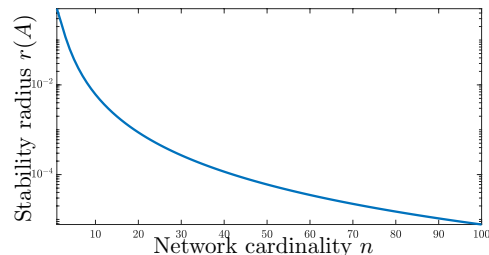
$$\lambda_{\min}(\mathcal{W}) \leq \bar{\lambda}(\mathcal{W}) \leq \frac{n_c}{n} \frac{1}{r(A)}$$

## An example: easily controllable but fragile network

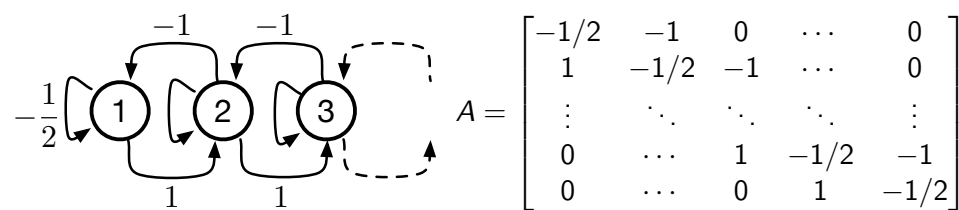


- control node =  $\{1\}$
- $\mathcal{W} = I \Rightarrow$  easy to control

As  $n$  increases, contr. degree is fixed, and fragility increases

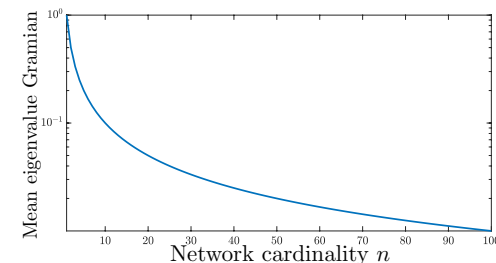


## An example: robust but poorly controllable network



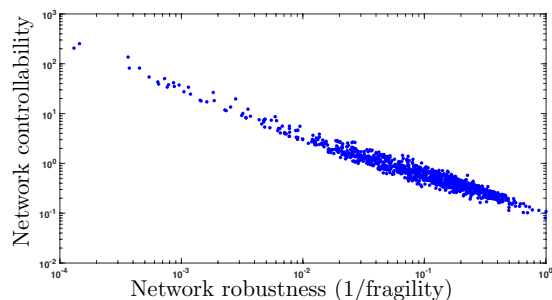
- control node =  $\{1\}$
- $A$  is normal;  $r(A) = 0.5$

As  $n$  increases, fragility degree is fixed, and contr. decreases



## A fundamental tradeoff between controllability and fragility

$$\text{controllability} \leq (\text{network parameters}) \cdot \text{fragility}$$



- Controllability implies fragility in complex networks
- Fundamental tradeoffs, valid for many (linear) network models
- Predator-prey dynamics, neural systems, traffic networks ...

## Summary

### In this talk

- 1 Metrics for network controllability and bounds
- 2 Design of networks that are easy to control
- 3 Controllability vs fragility tradeoff

### Research directions

- 1 Conjecture: networks with  $o(n)$  diameter are difficult to control
- 2 Criteria for networks that are easy to control
- 3 Model-based vs data-driven control of complex networks (poster)



S. Zhao and F. Pasqualetti, "Networks with Diagonal Controllability Gramians: Analysis, Graphical Conditions, and Design Algorithms," *Automatica*, 2019.



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G. Bianchin, F. Pasqualetti, S. Zampieri, "The Role of Diameter in the Controllability of Complex Networks," *IEEE Conf. on Decision and Control*, 2015.



F. Pasqualetti, C. Favaretto, S. Zhao, S. Zampieri, "Fragility and Controllability Tradeoff in Complex Networks," *American Control Conference*, 2018.

# Acknowledgments

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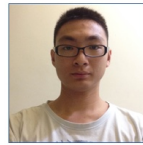
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UNIVERSITY OF CALIFORNIA  
RESEARCH INITIATIVES



## Controllability, Fragility, and Tradeoffs in Complex Networks

Fabio Pasqualetti



Department of Mechanical Engineering  
University of California, Riverside

Analysis and Control of Complex Networks  
Philadelphia, PA, July 9 2019