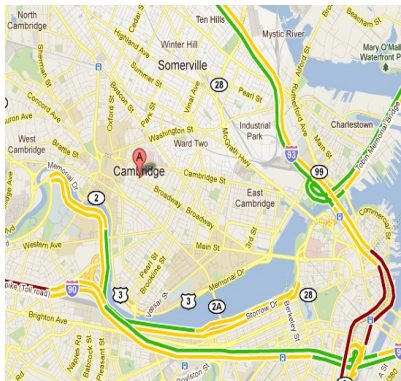


Resilient distributed control of dynamical flow networks

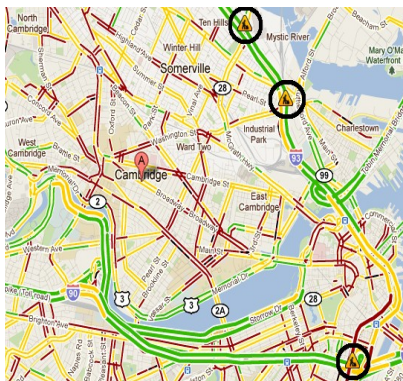
Giacomo Como
Department of Automatic Control
Lund University

LCCC Workshop on Dynamics and Control in Networks
October 17, 2014

Large-scale infrastructure networks



typical Monday at 18:30

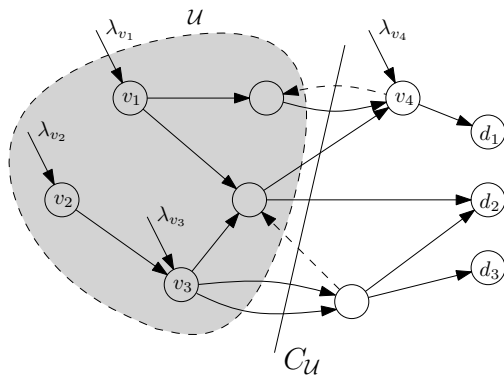


Monday, July 11, at 18:30

- ▶ good in business as usual, prone to disruptions
- ▶ cascade effects

⇒ network vulnerability ≫ \sum component vulnerabilities

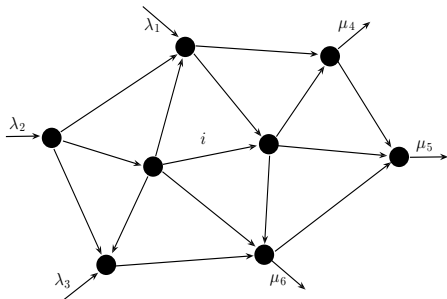
Max-flow min-cut theorem ('56)



$$\exists \text{ feasible equilibrium flow} \iff \min_{\mathcal{U}} \{C_{\mathcal{U}} - \lambda_{\mathcal{U}}\} > 0$$

► static, centralized, global information

Optimal network flow

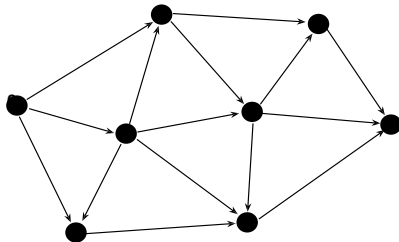


$$\min \sum_{i,j} c_{ij}(f_{ij})$$

$$\lambda_i + \sum_j f_{ji} - \sum_j f_{ij} - \mu_i = 0$$

► static, convex, rich duality theory

Wardrop equilibrium ('56)



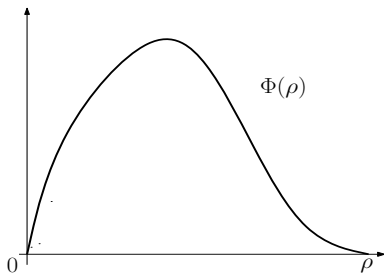
- ▶ 'the journey times in all routes actually used are equal and less than those which would be experienced on any unused route'

$f^W \Leftrightarrow$ Nash equilibrium of congestion game

- ▶ user optimum vs social optimum: price of anarchy
- ▶ using tolls on links allows one to align f^W with any desired f^*

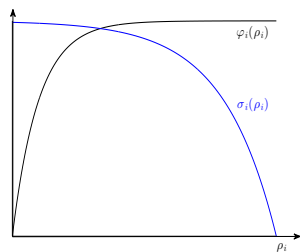
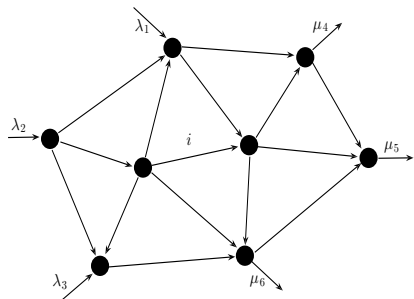
Lighthill-Whitham-Richards traffic model ('55)

$$\frac{\partial}{\partial t} \rho + \frac{\partial}{\partial x} \Phi(\rho) = 0$$



fundamental diagram

Dynamical flow networks

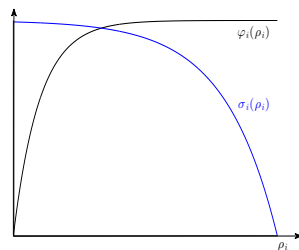
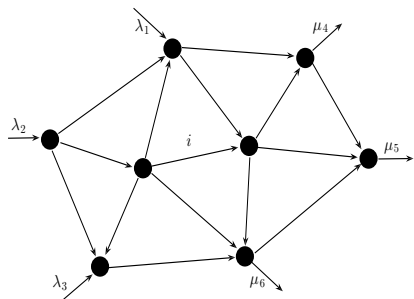


$$\dot{\rho}_i = \lambda_i + \sum_j f_{ji} - \sum_j f_{ij} - \mu_i$$

$$\mu_i + \sum_j f_{ij} \leq \varphi_i(\rho_i)$$

$$\sum_j f_{ji} \leq \sigma_i(\rho_i)$$

Dynamical flow networks



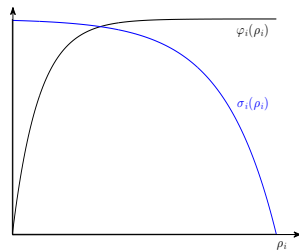
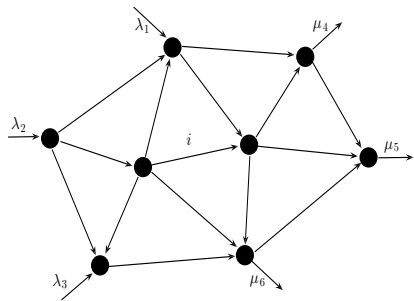
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- Daganzo'92, Gomes&Horowitz'06, Pisarski&CanudasDeWit'12, Varaiya'08, Coogan&Arcak'14, Karafyllis&Papageorgiou'14, ...

Dynamical flow networks



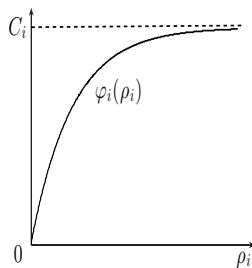
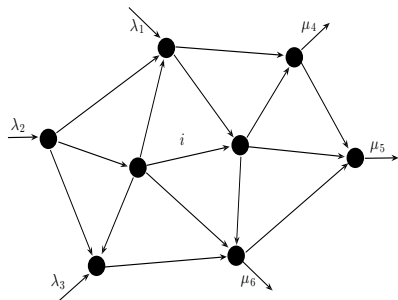
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$$\mu_i + \sum_j f_{ij} \leq \varphi_i(\rho_i)$$

$$\sum_j f_{ji} \leq \sigma_i(\rho_i)$$

► goal: scalable design of f_{ij} with good performance and resilience

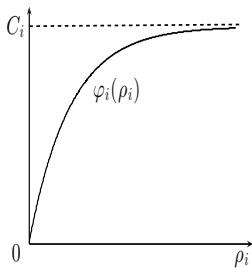
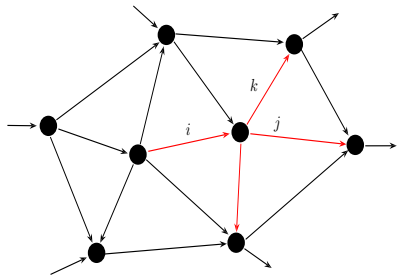
Resilient decentralized routing in dynamical flow networks



$$\dot{\rho}_i = \lambda_i + \sum_j f_{ji} - \sum_j f_{ij} - \mu_i$$

$$\mu_i + \sum_j f_{ij} \leq \varphi_i(\rho_i)$$

Decentralized routing

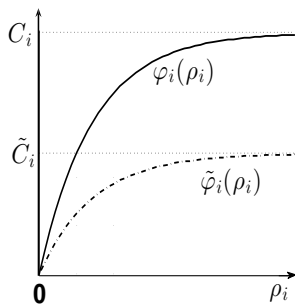


$$\dot{\rho}_i = \lambda_i + \sum_j f_{ji} - \sum_j f_{ij} - \mu_i$$

flow f_{ij} depends only on **local information** ρ^i

$$\begin{array}{ccccc} & f_{ij} & = & \varphi_i(\rho_i) & R_{ij}(\rho^i) \\ & \nearrow & & \uparrow & \nwarrow \\ \text{flow from } i \text{ to } j & & & \text{demand on } i & \text{fraction routed to } j \end{array}$$

Resilience

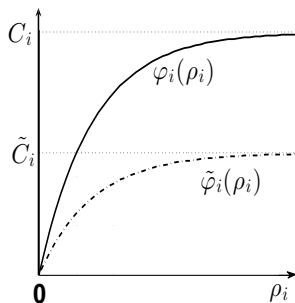


► magnitude of perturbation $\delta := \sum_i |\tilde{\lambda}_i - \lambda_i| + \sum_i |\tilde{C}_i - C_i|$

► throughput loss $\gamma := \liminf_{t \rightarrow +\infty} \frac{1}{t} \int_0^t \sum_i (\tilde{\lambda}_i(s) - \tilde{\mu}_i(s)) ds$

$$\dot{\rho}_i = \tilde{\lambda}_i + \sum_j \tilde{f}_{ji} - \sum_j \tilde{f}_{ij} - \tilde{\mu}_i \quad \tilde{f}_{ij} = \tilde{\varphi}_i(\rho_i) R_{ij}(\rho^i)$$

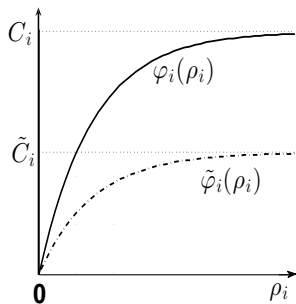
Resilience



margin of resilience := $\inf \{ \delta : \gamma > 0 \}$

- ▶ δ := magnitude of perturbation
- ▶ γ := throughput loss

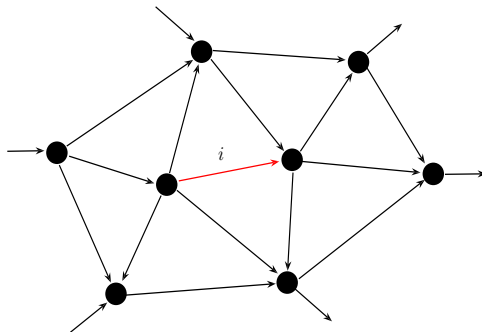
Resilience



margin of resilience $\nu := \inf \{ \delta : \gamma > 0 \}$

- ▶ necessarily: $\nu \leq \min_{\mathcal{U}} \{ C_{\mathcal{U}} - \lambda_{\mathcal{U}} \}$
- ▶ Problem: **max** resilience ν over decentralized routing policies R

Resilience with fixed routing



- ▶ $f_{ij} = R_{ij}\varphi_i(\rho_i)$ with constant R_{ij}
- ▶ start from equilibrium $f_i^* = \varphi_i(\rho_i^*)$

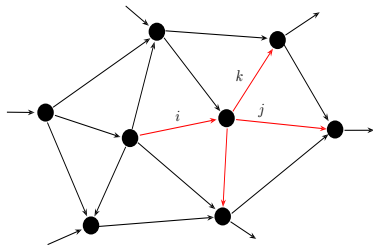
$$\nu = \min_i \underbrace{\{C_i - f_i^*\}}$$

link residual capacity

Resilience with locally responsive routing

(a) $R_{ij}(\rho^i)$ depends on local info

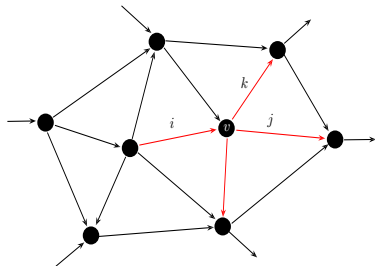
$$R_{ij}(\rho^i) \geq 0 \quad \sum_{j \in \mathcal{E}_i^+} R_{ij}(\rho^i) \equiv 1$$



Resilience with locally responsive routing

(a) $R_{ij}(\rho^i)$ depends on local info

$$R_{ij}(\rho^i) \geq 0 \quad \sum_{j \in \mathcal{E}_i^+} R_{ij}(\rho^i) \equiv 1$$



► **Theorem** [G.C., K.Savla, D.Acemoglu, M.Dahleh, E.Frazzoli, TAC'13]

$$(a) \quad \text{initial equilibrium } f^* \quad \implies \quad \nu \leq \min_v \underbrace{\sum_{j \in \mathcal{E}_v^+} (C_j - f_j^*)}_{\text{node residual capacity}}$$

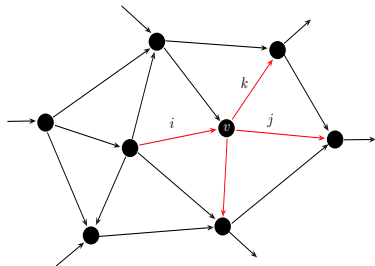
Resilience with locally responsive routing

(a) $R_{ij}(\rho^i)$ depends on local info

$$R_{ij}(\rho^i) \geq 0 \quad \sum_{j \in \mathcal{E}_i^+} R_{ij}(\rho^i) \equiv 1$$

(b) $\frac{\partial}{\partial \rho_k} R_{ij} \geq 0 \quad \forall k \neq j$

$$\rho_j \rightarrow \infty \Rightarrow R_{ij} \rightarrow 0$$



► Example

$$R_{ij}(\rho^i) = \frac{e^{-\beta(\rho_j + \alpha_j)}}{\sum_k e^{-\beta(\rho_k + \alpha_k)}} \quad \beta > 0$$

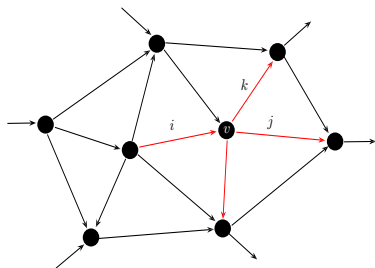
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► **Theorem** [G.C., K.Savla, D.Acemoglu, M.Dahleh, E.Frazzoli, TAC'13]

In acyclic networks

$$(a) \text{ and } (b) \quad \Longrightarrow \quad \nu = \min_v \sum_{j \in \mathcal{E}_v^+} (C_j - f_j^*)$$

f^* initial equilibrium

f^* globally attractive

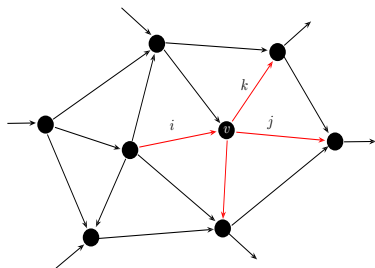
Resilience with locally responsive routing

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$$(a) \text{ and } (b) \quad \Rightarrow \quad \nu = \min_v \sum_{j \in \mathcal{E}_v^+} (C_j - f_j^*)$$

► **shocks** (and information) **propagate only downstream**

sub-additive effects of perturbations

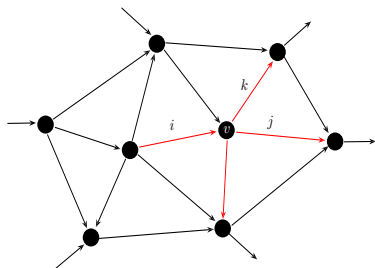
Resilience with locally responsive routing

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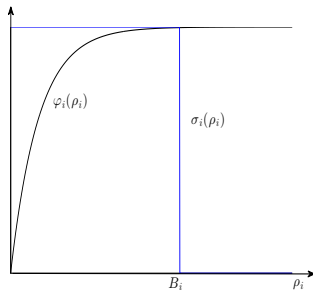
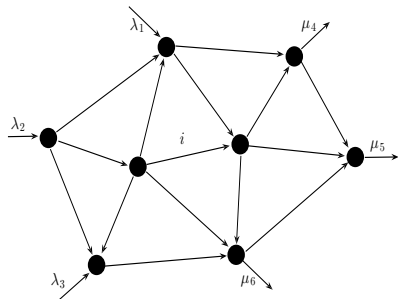


► **Theorem** [G.C., K.Savla, D.Acemoglu, M.Dahleh, E.Frazzoli, TAC'13]

$$(a) \text{ and } (b) \quad \Rightarrow \quad \nu = \min_v \sum_{j \in \mathcal{E}_v^+} (C_j - f_j^*)$$

► typically $\min_v \sum_{j \in \mathcal{E}_v^+} (C_j - f_j^*) \ll \min_u \{C_u - \lambda_u\}$

Dynamical flow networks with cascading failures

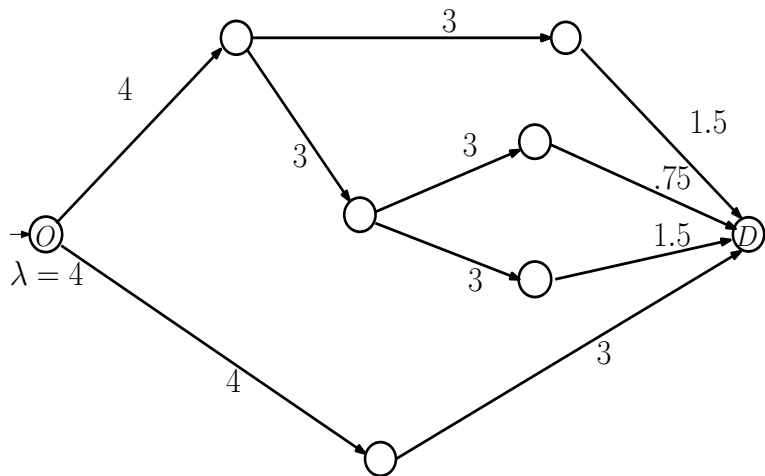


$$\dot{\rho}_i = \lambda_i + \sum_j f_{ji} - \sum_j f_{ij} - \mu_i$$

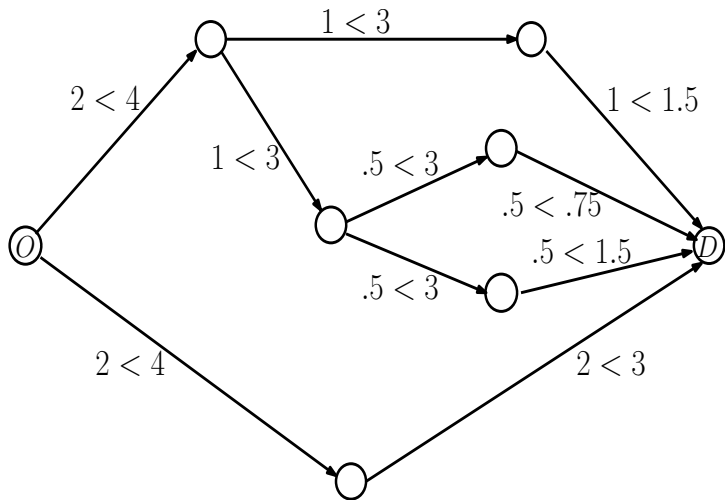
$$f_{ij} = \varphi_i(\rho_i) R_{ij}(\rho^i)$$

- ▶ link i goes down irreversibly when $\rho_i(t) = B_i$

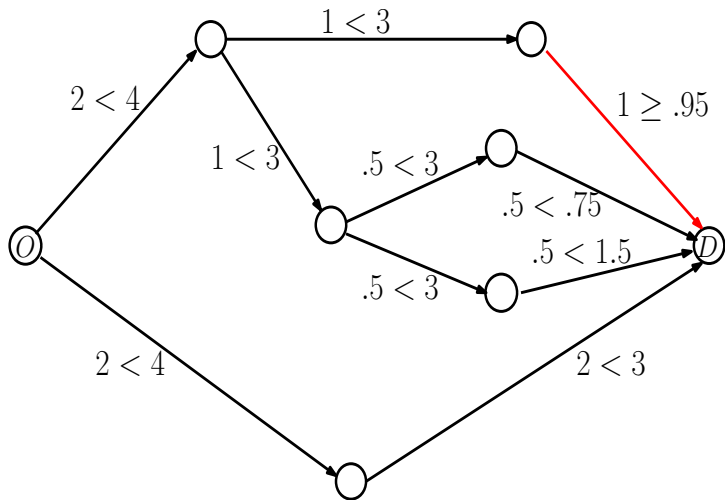
Dynamical flow networks with cascading failures



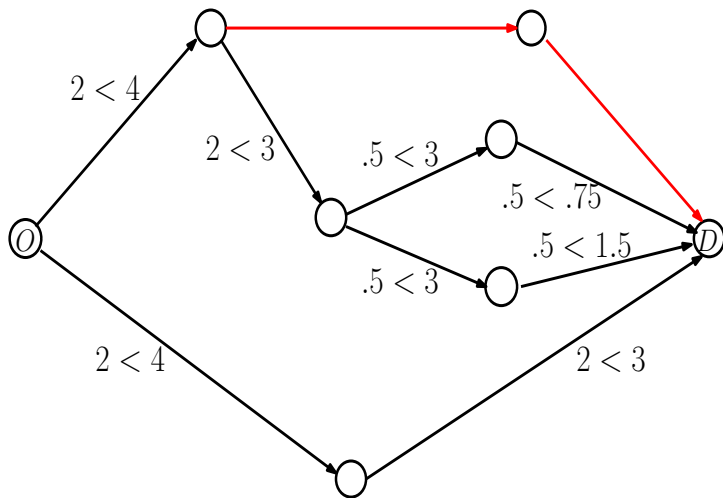
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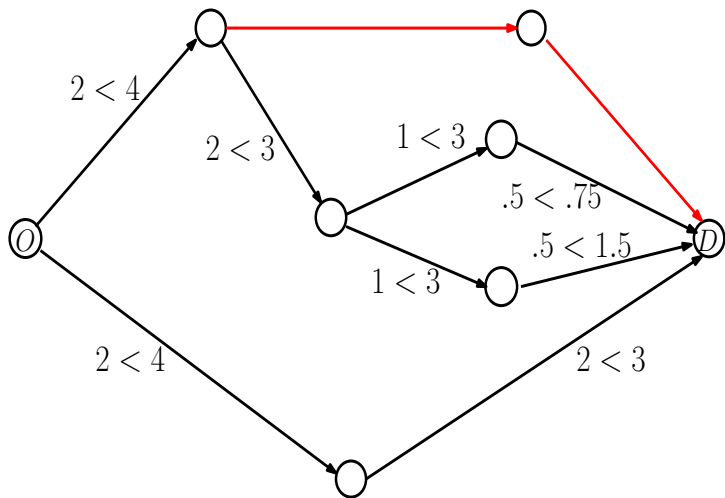
Dynamical flow networks with cascading failures



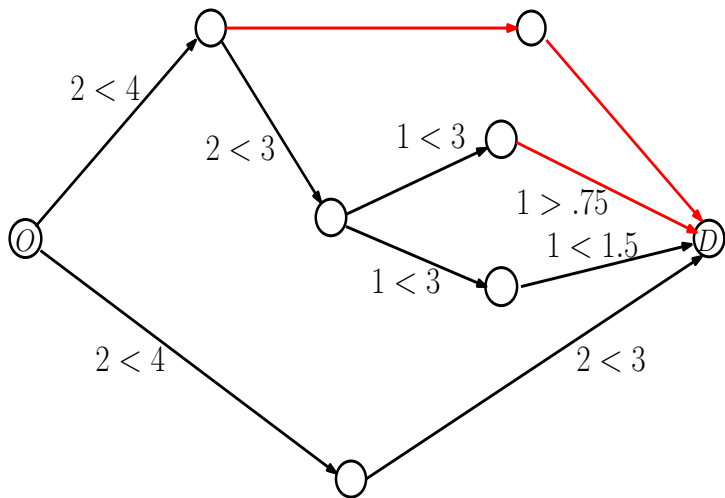
Dynamical flow networks with cascading failures



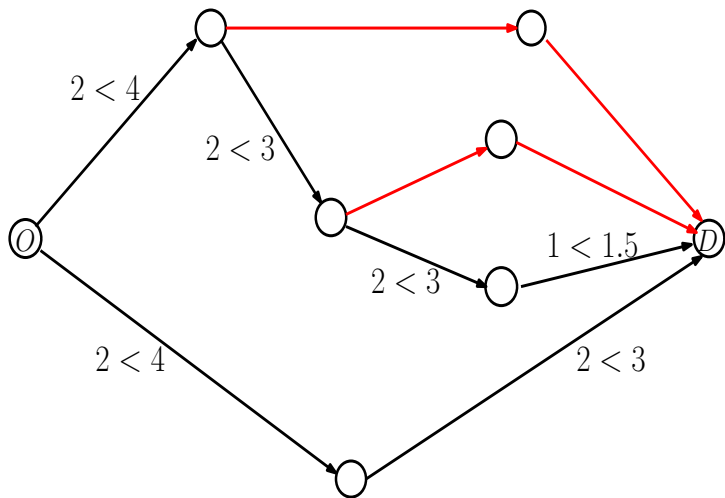
Dynamical flow networks with cascading failures



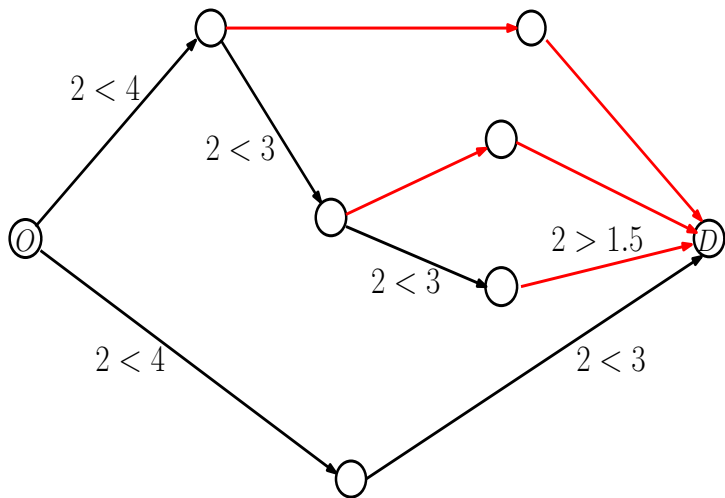
Dynamical flow networks with cascading failures



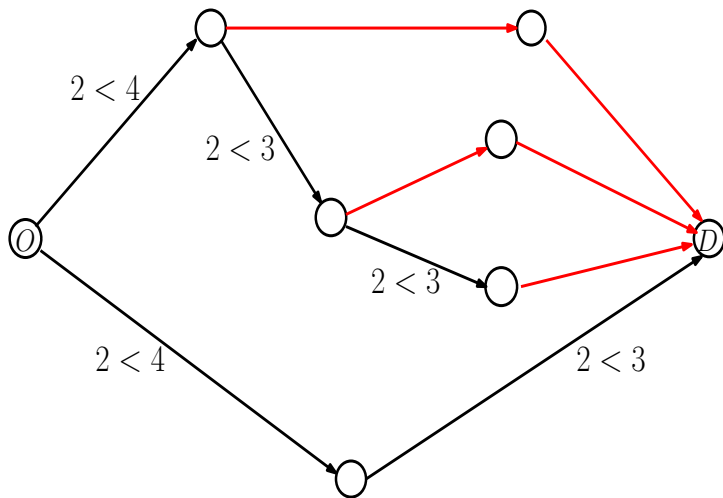
Dynamical flow networks with cascading failures



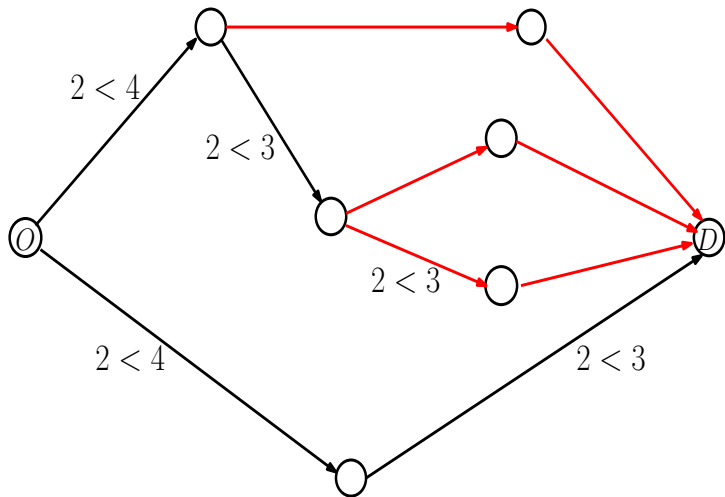
Dynamical flow networks with cascading failures



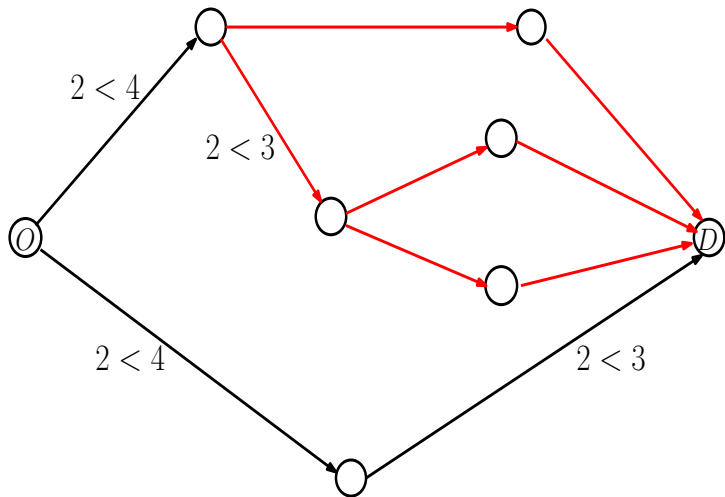
Dynamical flow networks with cascading failures



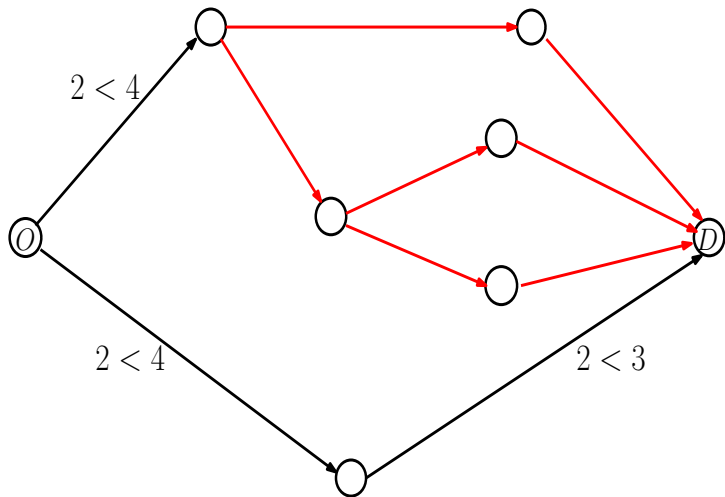
Dynamical flow networks with cascading failures



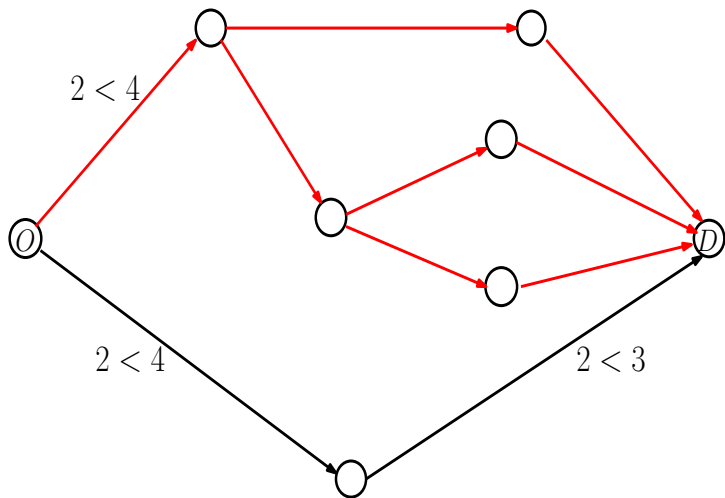
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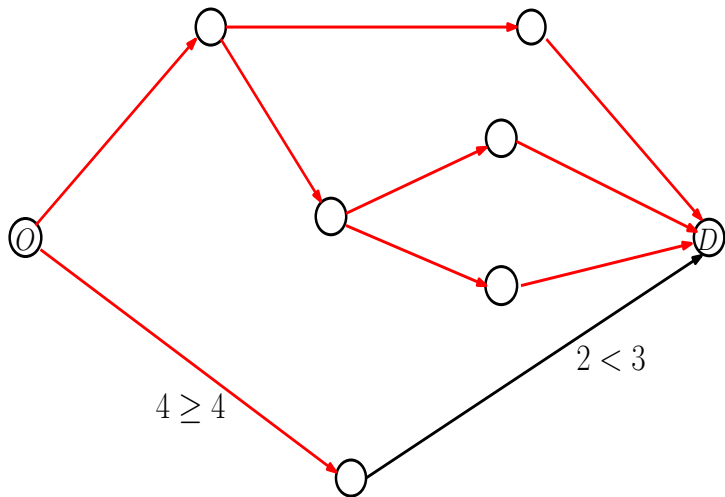
Dynamical flow networks with cascading failures



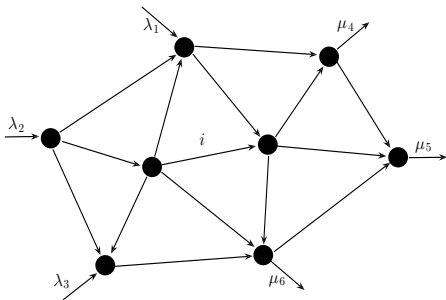
Dynamical flow networks with cascading failures



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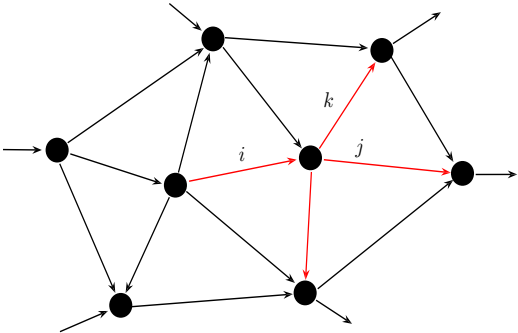
Theorem [K.Savla, G.C., M.Dahleh,'14] In acyclic networks

optimal resilience ν can be computed by dynamic programming

► shocks propagate both up- and downstream

► typically $\nu \ll \min_u \{C_u - \lambda_u\}$

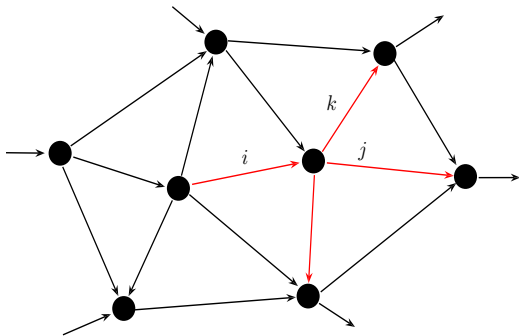
Decentralized routing with flow control



$$f_{ij} = \varphi_i(\rho_i) R_{ij}(\rho^i)$$

flow from i to j max outflow from i fraction routed to j

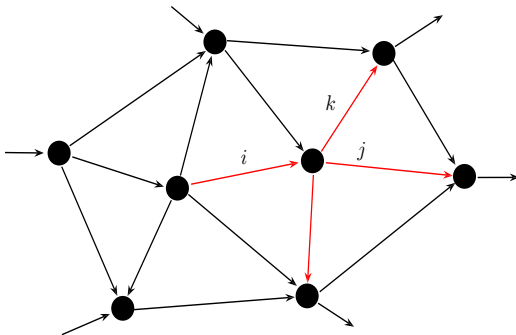
Decentralized routing with flow control



$$f_{ij} = \varphi_i(\rho_i) R_{ij}(\rho^i)$$

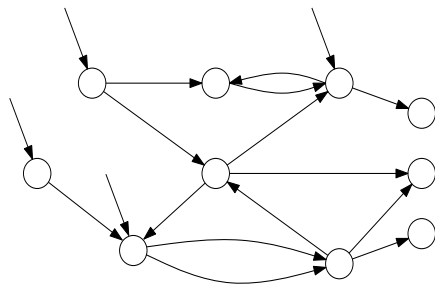
► relax $\sum_{j \in \mathcal{E}_i^+} R_{ij} = 1$ to $\sum_{j \in \mathcal{E}_i^+} R_{ij} \leq 1$, still decentralized

Decentralized monotone routing with flow control



$$R_{ij}(\rho^i) = \frac{e^{-\beta(\rho_j + \alpha_{ij})}}{e^{-\beta(\rho_i + \alpha_{ii})} + \sum_{k \in \mathcal{E}_i^+} e^{-\beta(\rho_k + \alpha_{ik})}}$$

Decentralized monotone routing with flow control

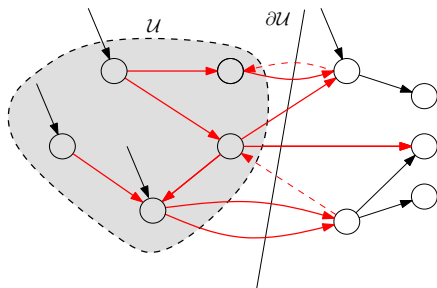


Theorem [G.C.,E.Lovisari,K.Savla,TCONES'14] :

Decentralized monotone routing, both with finite and infinite buffer

► $\min_{\mathcal{U}} \{C_{\mathcal{U}} - \lambda_{\mathcal{U}}\} > 0 \implies \exists \text{ equilibrium } \rho^* \text{ s.t. } \rho(t) \rightarrow \rho^* \quad \forall \rho(0)$

Decentralized monotone routing with flow control

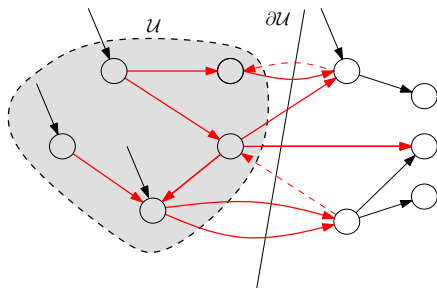


Theorem [G.C., E.Lovisari, K.Savla, TCONES '14] :

Decentralized monotone routing, infinite buffer

► $\min_{\mathcal{U}} \{C_{\mathcal{U}} - \lambda_{\mathcal{U}}\} < 0 \implies$ minimal throughput loss (graceful degradation)

Decentralized monotone routing with flow control

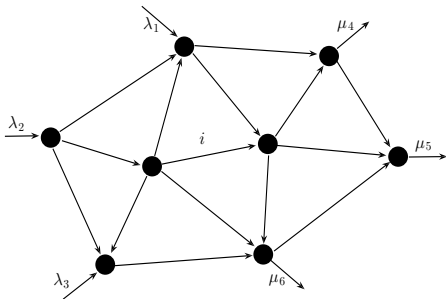


Theorem [G.C.,E.Lovisari,K.Savla,TCONES '14] :

Decentralized monotone routing, finite buffer

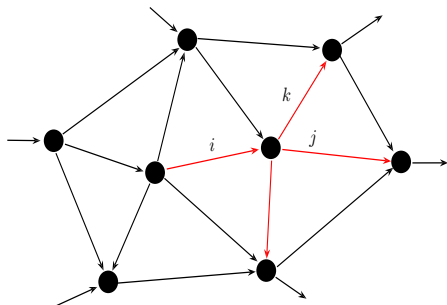
► $\min_{U} \{C_U - \lambda_U\} < 0 \implies$ links left of cut fail simultaneously

Beyond throughput and resilience



- ▶ decentralized routing + flow control achieves optimal resilience
implicitly propagating information through the network
- ▶ for other performance measures (e.g., delay at equilibrium)
communication / distributed optimization layer necessary

What equilibrium?



$$R_{ij}(\rho^i) = \frac{e^{-\beta(\rho_j + \alpha_{ij})}}{e^{-\beta(\rho_i + \alpha_{ii})} + \sum_{k \in \mathcal{E}_i^+} e^{-\beta(\rho_k + \alpha_{ik})}}$$

$$\alpha_{ij} = -\rho_j^* - \beta^{-1} \log(f_{ij}^*) \quad \alpha_{ii} = -\rho_i^* - \beta^{-1} \log(\varphi_i(\rho_i^*) - \sum_j f_{ij}^*)$$

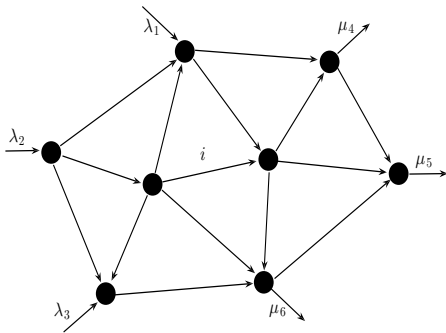
Equilibrium optimization

$$\min_{x,y} \sum_i \Psi_i(x_i)$$

$$\lambda_i + \sum_j f_{ji} = \sum_j f_{ij} + \mu_i$$

$$\mu_i + \sum_j f_{ij} \leq \varphi_i(x_i)$$

$$\lambda_i + \sum_j f_{ji} \leq \sigma_i(x_i)$$



[E.Lovisari,G.C.,A.Rantzer,K.Savla,'14]

If φ_i , σ_i , and $-\psi_i$ are concave (linear):

- ▶ equilibrium optimization is convex (linear) problem

suitable for distributed solutions (ADMM)

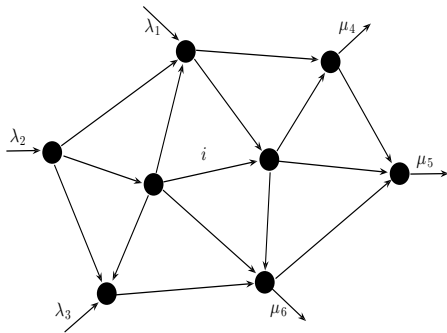
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$$\mu_i + \sum_j f_{ij} \leq \varphi_i(x_i)$$

$$\lambda_i + \sum_j f_{ji} \leq \sigma_i(x_i)$$



[E.Lovisari,G.C.,A.Rantzer,K.Savla,'14]

If φ_i , σ_i , and $-\psi_i$ are concave (linear):

- ▶ equilibrium optimization is convex (linear) problem

analogous problem when routing is fixed (or constrained)

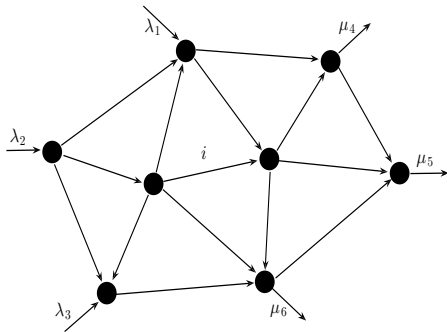
Optimal control

$$\min_{x,y} \sum_i \int_0^T \Psi_i(x_i(t)) dt$$

$$\dot{x}_i = \lambda_i + \sum_j (f_{ji} - f_{ij}) + \mu_i$$

$$\mu_i + \sum_j f_{ij} \leq \varphi_i(x_i)$$

$$\lambda_i + \sum_j f_{ji} \leq \sigma_i(x_i)$$

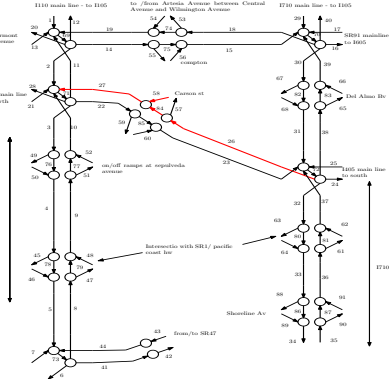
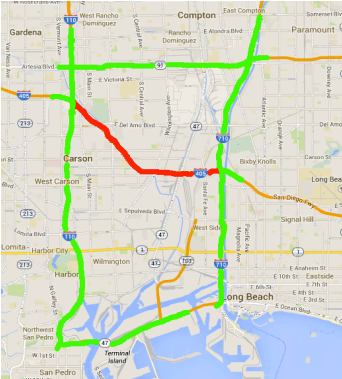


[E.Lovisari,G.C.,A.Rantzer,K.Savla,'14]

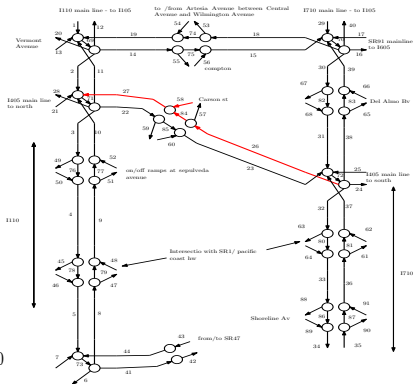
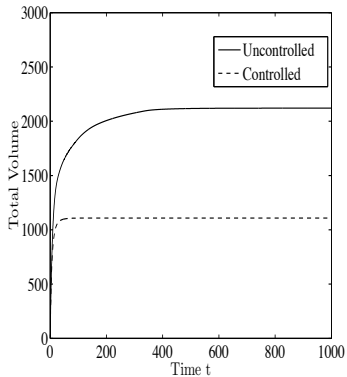
If φ_i , σ_i , and $-\psi_i$ are concave (linear):

- ▶ optimal control is convex (linear) problem

Highway net in Long Beach



Highway net in Long Beach



Conclusion

Dynamical flow networks:

- ▶ resilience and cascading failures with decentralized routing
- ▶ distributed equilibrium optimization and optimal control
- ▶ proofs: exploit monotonicity and l_1 -contraction

Dynamical flow networks beyond transportation:

- ▶ production networks
- ▶ distribution networks