Resilient distributed routing in dynamical flow networks

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### Resilient infrastructure networks

- $\blacktriangleright$  good in business as usual, prone to disruptions
	- $\triangleright$  cascade effects
- $\Rightarrow$  network vulnerability  $>>$   $\sum$  component vulnerabilities



Typical Monday at 18:30 p.m. Monday 11, 2011, at 6:30 p.m. Monday 11, 2011, at 7:30 p.m. Monday, July 11, 2011, at 7:30 p.m. Monday

#### Static flow networks

di-graph  $(\mathcal{V}, \mathcal{E})$ origin O destination D capacities  $C_e > 0$  $\mathcal{E}_{\mathsf{v}}^+=\{\mathsf{out}\text{-links of }\mathsf{v}\}$  $\mathcal{E}_{v}^{-} = \{\text{in-links of } v\}$ 



equilibrium flow:  $f = \{f_e\}$  such that  $0 \le f_e < C_e$  and

$$
\sum_{e\in \mathcal{E}_v^-} f_e = \sum_{e\in \mathcal{E}_v^+} f_e \qquad \forall v\neq \mathit{O}, \mathit{D}
$$

### Max-flow min-cut theorem



#### Max-flow min-cut theorem



 $\triangleright$  static, centralized, global information

#### Dynamical flow networks





### Dynamical flow networks



$$
\frac{\mathrm{d}}{\mathrm{d}t} \rho_e = \lambda_v \quad G_e^v \quad - \quad f_e
$$

(cf. with 
$$
\frac{\partial}{\partial t}\rho = -\nabla_x \cdot f
$$
)

### Flow function



$$
\mu_e(0) = 0 \qquad \sup_{\rho_e \ge 0} \mu_e(\rho_e) = C_e
$$

$$
\frac{d}{d\rho_e} \mu_e > 0
$$

### Distributed routing

$$
\lambda_v = \sum_{e \in \mathcal{E}_v^-} f_e \text{ total flow through } v
$$

 $G_{e}^{V} =$  fraction of  $\lambda_{V}$  routed to  $e$ local information:  $\rho^{\vee} = \{ \rho_e : e \in \mathcal{E}_{\mathsf{v}}^+ \}$ 





### Locally responsive distributed policies

$$
G^{v}: \mathbb{R}_{+}^{\mathcal{E}_{+}^{+}} \to \mathcal{P}(\mathcal{E}_{v}^{+})
$$
\n(i) 
$$
\frac{\partial}{\partial \rho_{j}} G_{e}^{v} \ge 0 \qquad \forall e \ne j \in \mathcal{E}_{v}^{+}
$$
\n(ii) 
$$
\rho_{e} \to \infty \quad \Rightarrow \quad G_{e}^{v}(\rho^{v}) \to 0
$$



#### Locally responsive distributed policies

$$
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$$
\rho_{e} \to \infty \quad \Rightarrow \quad G_{e}^{\vee}(\rho^{\vee}) \to 0
$$

Ex.: i-logit

 $G_{\mathbf{e}}^{\vee}(\rho^{\vee}) \propto \alpha_{\mathbf{e}} \exp(-\beta \rho_{\mathbf{e}})$  $e^{\nu}$ ( $\rho^{\nu}$ )  $\propto \alpha_e \exp(-\beta \rho_e)$   $\beta > 0, \ \alpha \in \mathbb{R}_+^{\mathcal{E}^+_{+}}$ 



### Locally responsive distributed policies

$$
G^{\vee}: \mathbb{R}_{+}^{\mathcal{E}_{+}^{+}} \to \mathcal{P}(\mathcal{E}_{\nu}^{+})
$$
\n(i)  $\frac{\partial}{\partial \rho_{j}} G_{e}^{\vee} \geq 0 \quad \forall e \neq j \in \mathcal{E}_{\nu}^{+}$ \n(ii)  $\rho_{e} \to \infty \quad \Rightarrow \quad G_{e}^{\vee}(\rho^{\vee}) \to 0$ \n  
\n(i)  $\Rightarrow$  network is monotone (i.e., Jacobian is Metzler)  $\Rightarrow \dots$   
\n $\Rightarrow \dots || \ ||_{1}$  contraction  $\Rightarrow \lim_{t \to \infty} \lambda_{D}(t)$  independent of  $\rho(0)$ 

#### Dynamical flow networks



$$
\frac{\mathrm{d}}{\mathrm{d}t}\rho_e = \lambda_v G_e^v(\rho^v) - \mu_e(\rho_e) \qquad \qquad e \in \mathcal{E}_v^+, \ v \in \mathcal{V} \setminus \{D\}
$$

transferring if  $\lim_{t\to\infty} \lambda_D(t) = \lambda_O$ 

#### Perturbed dynamical flow networks



 $\mathsf{Magnitude} \qquad \delta := \sum_{\mathsf{e}} \delta_{\mathsf{e}} = \sum_{\mathsf{e}} || \mu_{\mathsf{e}}(\,\cdot\,) - \tilde{\mu}_{\mathsf{e}}(\,\cdot\,) ||_{\infty}$ 

#### Margin of resilience



perturbed network:  $\frac{\mathrm{d}}{\mathrm{d}t}\tilde{\rho}_{\mathrm{e}} = \tilde{\lambda}_{\mathrm{v}}\, \mathcal{G}_{\mathrm{e}}^{\mathrm{v}}(\tilde{\rho}^{\mathrm{v}}) - \tilde{\mu}_{\mathrm{e}}(\tilde{\rho}_{\mathrm{e}}) \qquad \mathrm{e} \in \mathcal{E}_{\mathrm{v}}^{+}, \ v \neq D$ 

 $\gamma := \inf \mathsf{magnitude}$  of disruption s.t.  $\lim_{t \to +\infty} \tilde{\lambda}_D(t) < \lambda_O$ 

### Optimal resilience

residual node capacity for limit flow  $f^*$ 

$$
R(f^*) := \min_{v \neq D} \sum_{e \in \mathcal{E}_v^+} C_e - f_e^*
$$



#### Optimal resilience



Theorem: acyclic network, limit flow  $f^*$ 

1. distributed routing

 $\gamma \leq R(f^*)$ 

2. locally responsive routing  $\Rightarrow$  f<sup>\*</sup> globally attractive

$$
\gamma = R(f^*)
$$

## Optimal resilience (cont'd)



$$
R(f^*) := \min_{v \neq D} \sum_{e \in \mathcal{E}_v^+} C_e - f_e^*
$$

$$
1 \leq \frac{C - \lambda_O}{R(f^*)}
$$
 can be arbitrarily large

local information constraint =⇒ resilience loss

### Optimal resilience (cont'd)



$$
R(f^*) := \min_{v \neq D} \sum_{e \in \mathcal{E}_v^+} C_e - f_e^*
$$

 $\overrightarrow{v}$   $\overrightarrow{v}$   $\overrightarrow{e}$ λ $_{n}$  $\triangleright$  perturbations and distributed routing effect only downstream  $\blacktriangleright$  locally responsive policies ⇓ locally optimal load rebalance



#### Bounded density capacities



#### Bounded density capacities



$$
e = (v, w) \Rightarrow \frac{d}{dt} \rho_e = \chi_v(t) \lambda_v G_e^v(\rho^v) - \chi_w(t) \mu_e(\rho_e)
$$

 $\chi_{\tiny V}(t) := 1 - \prod_{e \in {\mathcal E}_\mathbf{v}^+} (1 - \xi_e(t)) \qquad \xi_e(t) := \mathbbm{1}_{[0,\rho_e^{\rm max})}(\rho_e)$ 



Perturbation localized on  $\mathcal E^+_{\sf w}$ , such that  $\sum_{\mathcal E_{\sf w}^+} \widetilde{\mathsf C}_{\sf e}<\lambda_{\sf w}^*$ 



Node  $w$  overloaded  $\Rightarrow$  every  $e \in \mathcal{E}^+_{w}$  drops



Hence  $w$  drops together with all  $e \in \mathcal{E}_w^-$ 



Possibly other nodes downstream become overloaded...



...and eventually drop



Either a new equilibrium flow is eventually achieved



...or not.



If the perturbation overloads the origin...



... no more flow passes through the network.



Theorem:  $\rho_e^{\text{max}} < +\infty \quad \forall e \quad \Longrightarrow \quad \lim_{t \to \infty} \lambda_D(t) \in \{0, \lambda_O\}$ 

1 acyclic network, any distributed routing

 $\gamma \leq \Gamma(f^*)$   $\Gamma(f^*) := \dots$  backward induction

 $2$  locally responsive routing,  $d_{\rm v}^+\leq 2\quad\Longrightarrow\quad \gamma=\mathsf{\Gamma}(f^*)$ 3 locally responsive routing  $\implies \gamma > R(f^*)$ 



















#### Slowing down local flow



 $\mathcal{G}^{j}_{j}(\rho_{j},\rho^{\nu}):=$  fraction of flow kept on  $j$ 

 $1-G^{j}_{j}(\rho_{j},\rho^{\mathsf{v}})$  := fraction of flow allowed through  $\mathsf{\nu}$ 

#### Slowing down achieves capacity



Theorem: Acyclic network, locally responsive policy

$$
\gamma = \mathcal{C} - \lambda_{\mathcal{O}}
$$

both with finite and infinite  $\rho_{\bm{e}}^{\textsf{max}}$ 











#### Conclusion

#### Summary

- $\triangleright$  robust distributed routing for dynamical flow networks
- $\triangleright$  margins of resilience
- $\triangleright$  role of cascades
- Current/future directions
- $\blacktriangleright$  cycles
- $\blacktriangleright$  multicommodity flows
- $\triangleright$  value of communication
- $\triangleright$  resilience of networks with other dynamics