Resilient distributed routing in dynamical flow networks

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#### Resilient infrastructure networks

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



Typical Monday at 18:30

## Static flow networks

di-graph  $(\mathcal{V}, \mathcal{E})$ origin Odestination D capacities  $C_e > 0$  $\mathcal{E}_v^+ = \{ \text{out-links of } v \}$ 

 $\mathcal{E}_v^- = \{\text{in-links of } v\}$ 



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

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

#### Max-flow min-cut theorem



#### Max-flow min-cut theorem



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 
$$rac{\partial}{\partial t}
ho = -
abla_{ extsf{x}}\cdot f$$
 )

## Flow function



$$egin{aligned} \mu_e(0) &= 0 & \sup_{
ho_e \geq 0} \mu_e(
ho_e) = C_e \ & & & & & \ & & & rac{\mathrm{d}}{\mathrm{d}
ho_e} \mu_e > 0 \end{aligned}$$

# Distributed routing

$$\lambda_{\mathbf{v}} = \sum_{\mathbf{e} \in \mathcal{E}_{\mathbf{v}}^{-}} \mathit{f}_{\mathbf{e}}$$
 total flow through  $\mathbf{v}$ 

 $G_e^{\nu} =$ fraction of  $\lambda_{\nu}$  routed to elocal information:  $\rho^{\nu} = \{\rho_e : e \in \mathcal{E}_{\nu}^+\}$ 





# Locally responsive distributed policies

$$\begin{aligned} G^{\nu} : \mathbb{R}^{\mathcal{E}^{+}_{+}}_{+} &\to \mathcal{P}(\mathcal{E}^{+}_{\nu}) \\ \text{(i)} \ \frac{\partial}{\partial \rho_{j}} G^{\nu}_{e} \geq 0 \qquad \forall e \neq j \in \mathcal{E}^{+}_{\nu} \\ \text{(ii)} \ \rho_{e} \to \infty \quad \Rightarrow \quad G^{\nu}_{e}(\rho^{\nu}) \to \end{aligned}$$



## Locally responsive distributed policies

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

Ex.: i-logit

 $G_{e}^{v}(
ho^{v}) \propto lpha_{e} \exp(-eta 
ho_{e})$ 

 $\beta > \mathbf{0}, \ \alpha \in \mathbb{R}^{\mathcal{E}^+_{\mathbf{v}}}_+$ 



## Locally responsive distributed policies



# Dynamical flow networks



$$\frac{\mathrm{d}}{\mathrm{d}t}\rho_{e} = \lambda_{v} G_{e}^{v}(\rho^{v}) - \mu_{e}(\rho_{e}) \qquad e \in \mathcal{E}_{v}^{+}, \ v \in \mathcal{V} \setminus \{D\}$$

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

#### Perturbed dynamical flow networks



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

#### Margin of resilience



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

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

# 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  $\Rightarrow$   $f^*$  globally attractive

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

## Optimal resilience (cont'd)



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

local information constraint  $\implies$  resilience loss

## Optimal resilience (cont'd)



perturbations and distributed routing effect only downstream
 locally responsive policies
  $\downarrow$  locally optimal load rebalance
  $\lambda_v$  e

#### Bounded density capacities



#### Bounded density capacities



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

 $\chi_{\boldsymbol{v}}(\boldsymbol{t}) := 1 - \prod_{e \in \mathcal{E}_{\boldsymbol{v}}^+} (1 - \xi_e(\boldsymbol{t})) \qquad \overline{\xi_e(\boldsymbol{t})} := \mathbb{1}_{[0,\rho_e^{\max})}(\rho_e)$ 



Perturbation localized on  $\mathcal{E}_w^+$ , such that  $\sum_{\mathcal{E}_w^+} \tilde{\mathcal{C}}_e < \lambda_w^*$ 



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

1 acyclic network, any distributed routing

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

2 locally responsive routing,  $d_v^+ \leq 2 \implies \gamma = \Gamma(f^*)$ 

 $\begin{array}{ccc} \textbf{3} \text{ locally responsive routing} & \Longrightarrow & \gamma \geq R(f^*) \end{array}$ 



















#### Slowing down local flow



 $G_i^j(\rho_j, \rho^v) :=$  fraction of flow kept on j

 $1 - G_j^j(
ho_j,
ho^{
m v}) :=$  fraction of flow allowed through  ${
m v}$ 

## Slowing down achieves capacity



Theorem: Acyclic network, locally responsive policy

$$\gamma = C - \lambda_O$$

both with finite and infinite  $\rho_e^{\max}$ 











## Conclusion

#### Summary

- robust distributed routing for dynamical flow networks
- margins of resilience
- role of cascades

#### Current/future directions

- cycles
- multicommodity flows
- value of communication
- resilience of networks with other dynamics