#### Dynamic Management of Network Risk from Epidemic Phenomena

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# **Analyzing Epidemics**

- Classic models (SIS, SIR) now generalized to probabilistic models of infection (Ganesh et al. 2005)
- Widely applicable digital/biological viruses, network router faults, social media influence, etc.
- Control
  - Optimization approaches explicitly include budget constraints (Gourdin et al. 2011, Preciado et al. 2013, Preciado et al. 2014)
  - Our methods also deal with decentralization and robustness

Model Framework

Proposed Approach

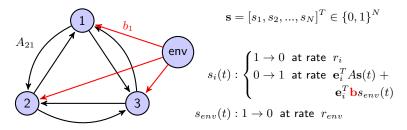
Experiments

Conclusions & Future Work

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### System Dynamics

• SIS epidemic model as a continuous-time Markov process



• Instantaneous energy of infection

$$P(\mathbf{1}^T \mathbf{s}(t) > 0) \le \sqrt{N} \|\mathbf{z}(t)\|_2$$

$$\dot{\mathbf{z}}(t) = D\mathbf{z}(t) + \mathbf{b}e^{-r_{env}t}s_{env}(0),$$
  
$$\mathbf{z}(0) = \mathbf{s}(0), \quad D := A - \text{diag}(\mathbf{r})$$

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#### **Problem Setup**

• Control environmental impact on system via limited budget at discrete intervals

- Discretize dynamics:  $\mathbf{x}(k) := \mathbf{z}(kh)$ 

- Control b w.r.t budget constraints

$$\mathbf{u}(k) = (\mathbf{b} - \mathbf{w}(k))e^{-r_{env}kh}s_{env}(0)$$
$$\mathbf{0} \leq \mathbf{w}(k) \leq \mathbf{b}, \quad \|\mathbf{w}(k)\|_1 \leq c,$$

Minimize cumulative energy of infection via MPC

$$\int_0^\infty P(\mathbf{1}^T \mathbf{s}(t) > 0) dt \leq \sqrt{N} \int_0^\infty \|\mathbf{z}(t)\|_2 dt \approx \sqrt{N} \sum_{k=0}^T \|\mathbf{x}(k)\|_2$$

minimize 
$$J_m := \sqrt{N} \sum_{k=m+1}^{T+m} \|\mathbf{x}(k)\|_2$$
  
subject to (dynamics, constraints)

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# Problem Setup (contd.)

- Centralized solution is inefficient for large  ${\cal N}$  and network connectivity might not be known perfectly
- Decentralization split system into M (possibly unequal) subsystems
- Robustness off-diagonal blocks of  ${\cal A}$  are known only within some uncertainty region

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#### **Reduced-Order Models**

- Each subsystem models the other subsystems' dynamics through reduced-order models (decentralization/accuracy tradeoff)
- Standard model reduction procedure (e.g. via balanced truncation, Safonov et al. 1988)
  - Procedure outputs compression and expansion operators
  - Analogous to similarity transformation
- Local problem for subsystem i (with state  $\mathbf{x}_r^i$ , control  $\mathbf{u}_r^i$ ):

minimize 
$$J_m^i := \sqrt{N} \sum_{k=m+1}^{T+m} \|\mathbf{x}_r^i(k)\|_2$$
  
subject to (reduced dynamics, reduced constraints)

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#### **Robust Formulation**

• Polytopic/"scenario" uncertainty sets (efficiency/robustness tradeoff)

$$\mathcal{A}_{mn} = \{ C | C = \sum_{k=1}^{L_{mn}} \mu_k A_{mn}(k), \ \mu_k \ge 0, \ \sum_{k=1}^{L_{mn}} \mu_k = 1 \}$$

- Straightforward generalization for model reduction via generalized balanced truncation (replace Lyapunov eq. with LMI)
- Robust counterpart for local problem  $(\min \sup_{\mathcal{A}} J_m^i)$  is an SOCP
  - Requires linearizing dynamics s.t.  $\mathbf{x}^i_r(k)$  is affine in A

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### Experiments

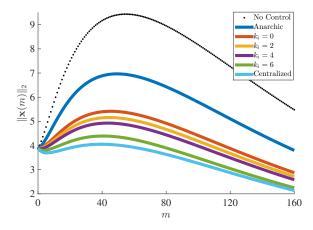
- N = 24, M = 3, equal subsystem sizes, random adjacency matrices and recovery rates
- Environment heals, but at a slower rate than the system -  $s_{env}(0)=1,\,r_{env}=0.2<-\lambda_i(D)\in[0.33,1]$
- We vary the order of reduced models,  $k_i = \{0,2,4,6,8\}$
- Compare with no control, anarchy (each subsystem has budget c/M)

Experiments

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### Experiments (contd.)

- Cooperation/dynamic budget allocation assuages overshoot
- Larger  $k_i$  yields better performance



Experiments

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### **Conclusions & Future Work**

- Developed framework for dynamic network protection incorporating budget constraints, decentralization, and robustness to uncertainty
- Tradeoffs between efficiency/robustness and decentralization/optimality
- Many avenues worth further research
  - Uncertainty sets with greater scalability
  - Optimal decentralized schemes for partitioning budgets between subsystems
  - Dynamic network topologies