







# Epidemic Control through Learning & Optimization

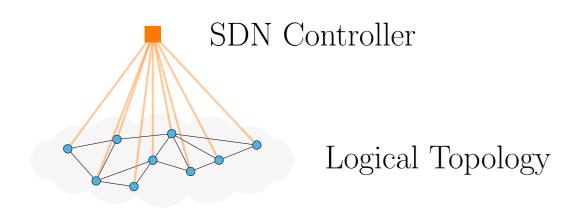
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### Network security

- Networked systems face propagation of malware, cascading hardware failures, DDoS.
- Software-defined networking enables full automated control over network topology.

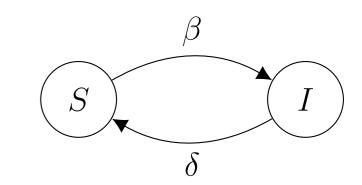


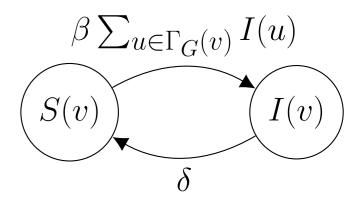
### Model

**Base:** Epidemic models can be used to represent propagating threats: each health status corresponds to a compartment (e.g.: S for susceptible, I for infected).

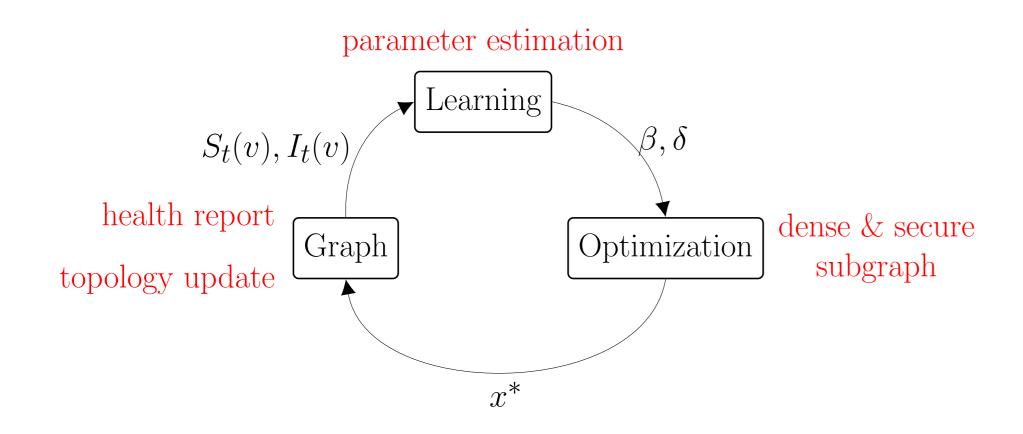
**Refinement:** Standard compartmental models may be refined with network structure: underlying topology is given by an undirected graph G = (V, E).

**Result:** The Markov process has  $|\{S,I\}|^{|V|}$  states and the transition rates for a node depend on the state of its neighbours. The model parameters are  $\beta$  and  $\delta$ .





## Turning a theorem into a control system



### **Definition** (Spectral radius)

The spectral radius of a graph is the largest eigenvalue of its adjacency matrix and satisfies:

$$\frac{1}{n} \sum_{v \in V} \deg_G(v) \le \lambda_{\max}(G) \le \max_{v \in V} \deg_G(v).$$

**Theorem** (Ganesh et al., 2005)

Given a SIS epidemic with parameters  $\beta$  and  $\delta$  on a graph G:

$$\lambda_{\max}(G) < \frac{\delta}{\beta}$$

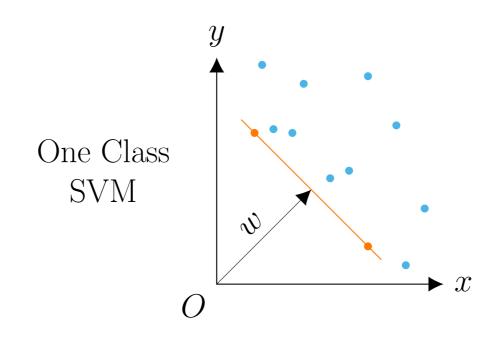
implies that the epidemic dies out in time  $\Theta(\log n)$ .

# Know your enemy: learning epidemic parameters

#### Anomaly detection

- Each node determines its health status by learning.
- One Class SVMs are a family of classifiers used for anomaly detection.
- A OC-SVM is trained on "healthy" data only: the system does not require prior experience of the epidemic to come.

### Maximizing the margin w.r.t. the origin



#### Parameter estimation

Input: Time series of node health data.

**Model:** SIS model with unknown parameters  $\beta$  and  $\delta$ .

Estimate:  $\beta$  and  $\delta$ .

# Approximation algorithms for the secure subgraph problem

#### Closed walks

• Norm inequalities give:

$$\lambda_{\max}(A) = O(||A||_{\log n}).$$

- The number of closed walks of length k is  $||A||_k^k$ .
- Find subgraph with few closed walks of length  $\log n$ .

### Mathematical program

$$\max \sum_{e \in E} x_e$$

$$\sum_{e \in E} x_e A_e \leq \delta/\beta I$$

$$x \in \{0, 1\}^m$$

 $\circ$   $A_e$  is the adjacency matrix of edge e.

### SDP and random matrices

- Continuous relaxation of the mathematical program gives a SDP.
- $\circ$  Optimal solution  $x^*$  used as a distribution.
- Leverage concentration of measure for symmetric random matrices.

### Interlacing polynomials

- Polynomial-valued r.v.s related to the characteristic polynomial of a graph.
- Undirected graphs have real roots: is it a rare property?
- Bounding the spectral radius by bounding roots.