

# Epidemic Control with Learning & Optimization

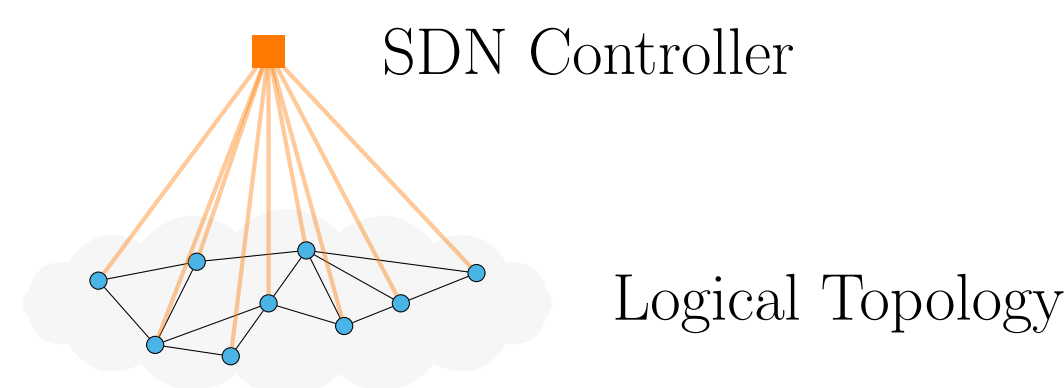
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## Network security

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

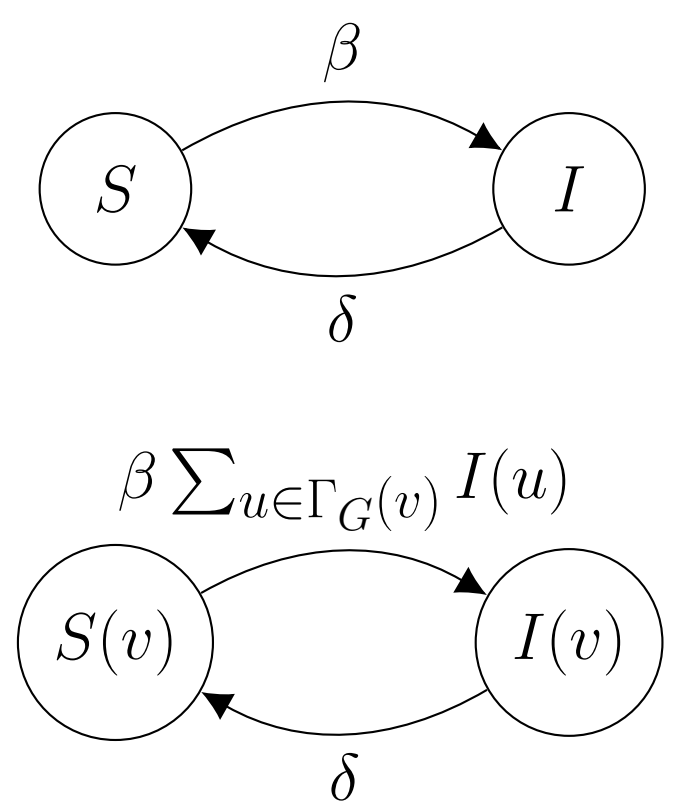


## Model

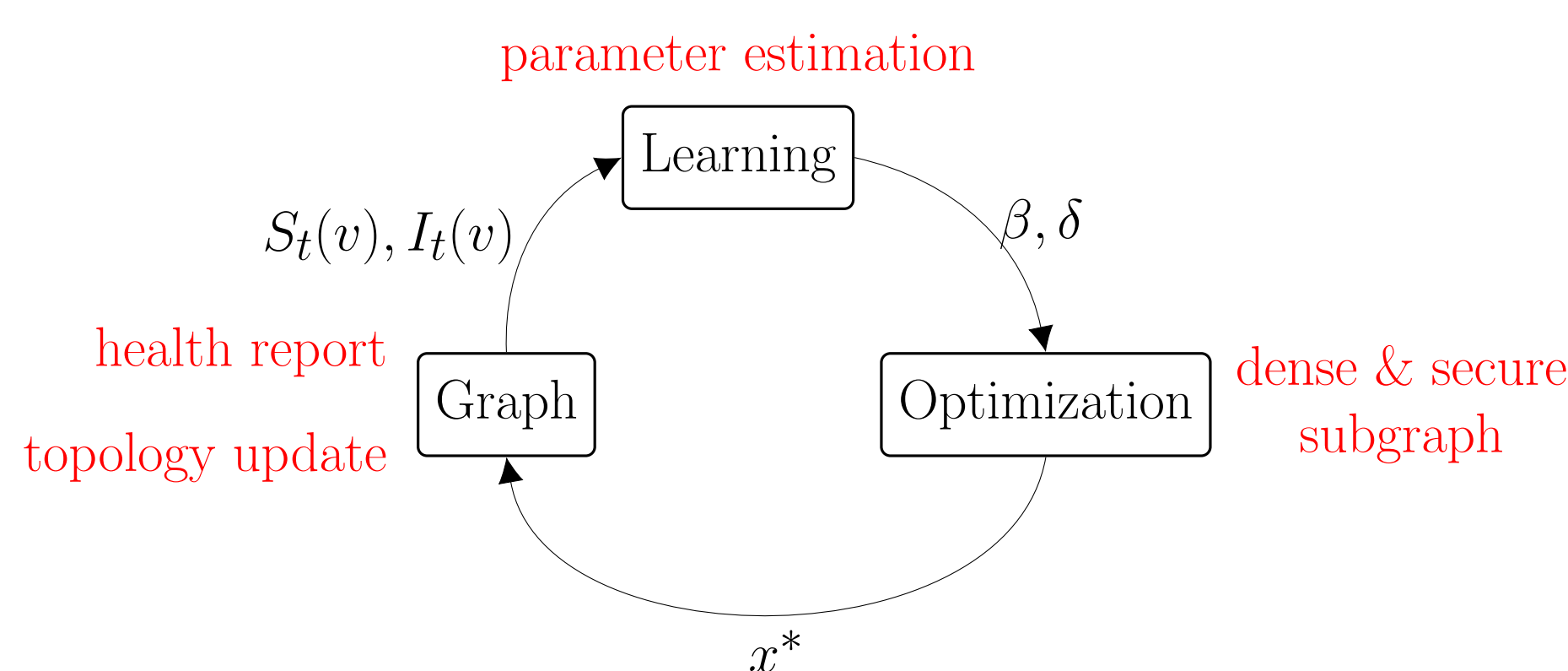
**Base:** Epidemic models can be used to represent propagating threats: each health status corresponds to a compartment (e.g.:  $S$  for susceptible,  $I$  for infected).

**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:

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

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

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

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

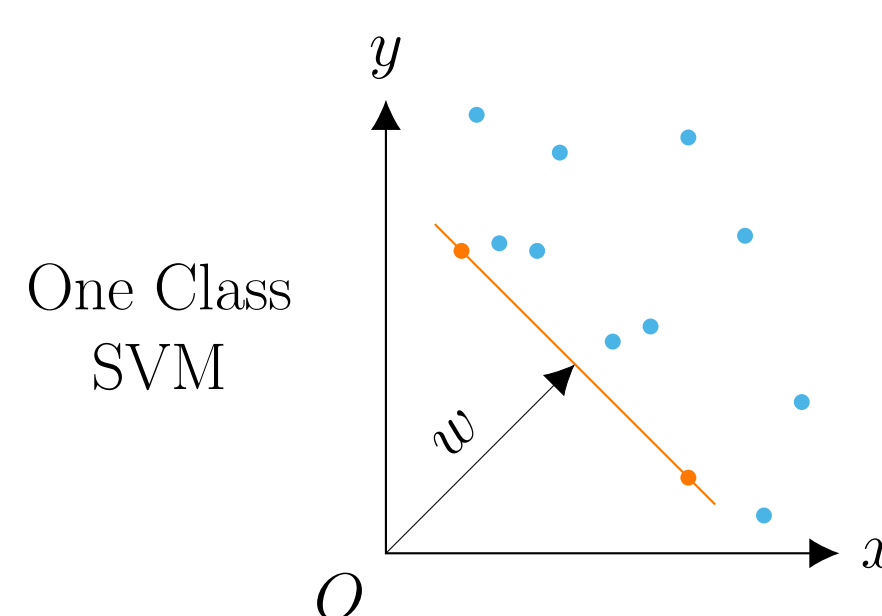
implies that the epidemic dies out in time  $O(\log n)$ .

## 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

$$\begin{aligned} \max \quad & \sum_{e \in E} x_e \\ \sum_{e \in E} x_e A_e & \preceq \delta / \beta I \\ x & \in \{0, 1\}^m \end{aligned}$$

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

### SDP and random matrices

- Continuous relaxation of the mathematical program gives a SDP.
- 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.