

Optimal Resource Allocation to Control Spreading Processes in Arbitrary Networks

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Project Summary

In a network with a spreading (cascading) process, what's the optimal strategy to allocate scarce protective and corrective control resources?

- Agents in the network are interacting and influencing one another's vulnerability.
- While centrality measures of the agents can be a guide, based on a convex framework, we find a nontrivial resource allocation scheme.
- Our framework is for strongly connected networks of arbitrary structure, including directed networks.

Introduction

Agents in the network are in either susceptible or infected state.

- Protective resources lower infection rate β_i , corrective resources increase recovery rate δ_i .
- Probability of infection evolves according to

$$\frac{dp_i(t)}{dt} = (1 - p_i(t)) \beta_i \sum_{j=1}^n a_{ij} p_j(t) - \delta_i p_i(t). \quad (1)$$

- Control of epidemic outbreak – if the eigenvalue with largest real part of $BA_G - D$ satisfies

$$\Re[\lambda_1(BA_G - D)] \leq -\varepsilon,$$

for some $\varepsilon > 0$, the disease-free equilibrium ($\mathbf{p}^* = \mathbf{0}$) is globally exponentially stable.

Our Model & Problem

Find optimal protective and corrective resource allocation scheme that minimizes the cost and satisfies specified decay rate

$$\text{minimize}_{\{\beta_i, \delta_i\}_{i=1}^n} \sum_{i=1}^n f_i(\beta_i) + g_i(\delta_i) \quad (2)$$

$$\text{subject to } \Re[\lambda_1(\text{diag}(\beta_i) A_G - \text{diag}(\delta_i))] \leq -\varepsilon, \quad (3)$$

$$\underline{\beta}_i \leq \beta_i \leq \bar{\beta}_i, \quad (4)$$

$$\underline{\delta}_i \leq \delta_i \leq \bar{\delta}_i, \quad i = 1, \dots, n, \quad (5)$$

where (2) is the total investment, (3) constrains the decay rate to ε , and (4)-(5) maintain the infection and recovery rates in their feasible limits.

Budget Constrained

$$\text{maximize}_{\varepsilon, \{\beta_i, \delta_i\}_{i=1}^n} \varepsilon \quad (6)$$

$$\text{subject to } \Re[\lambda_1(\text{diag}(\beta_i) A_G - \text{diag}(\delta_i))] \leq -\varepsilon, \quad (7)$$

$$\sum_{i=1}^n f_i(\beta_i) + g_i(\delta_i) \leq C, \quad (8)$$

$$\underline{\beta}_i \leq \beta_i \leq \bar{\beta}_i, \quad (9)$$

$$\underline{\delta}_i \leq \delta_i \leq \bar{\delta}_i, \quad i = 1, \dots, n, \quad (10)$$

Network Interactions

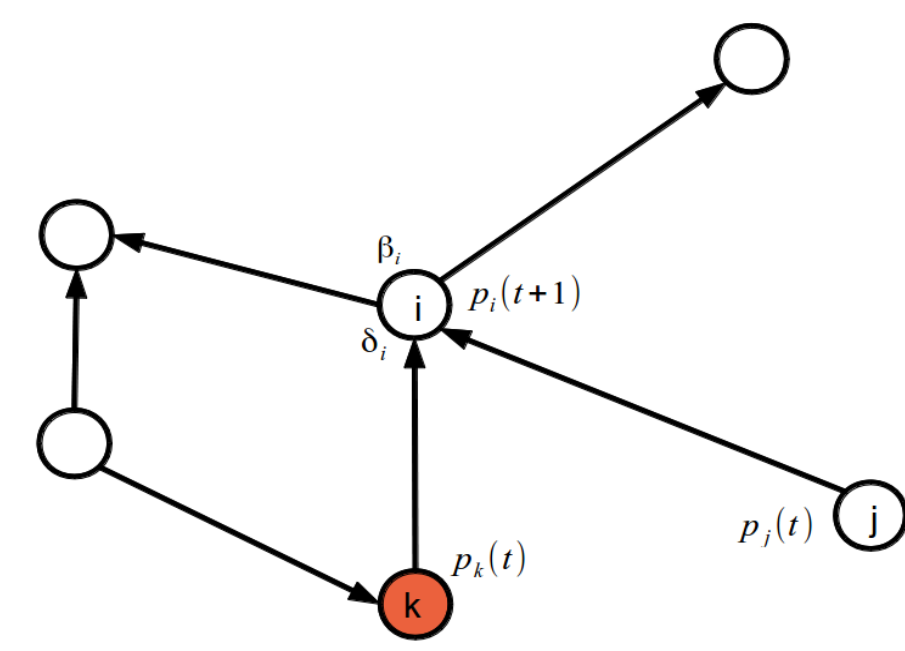


Figure 1: A Network illustrating spread of infection

Cost Functions

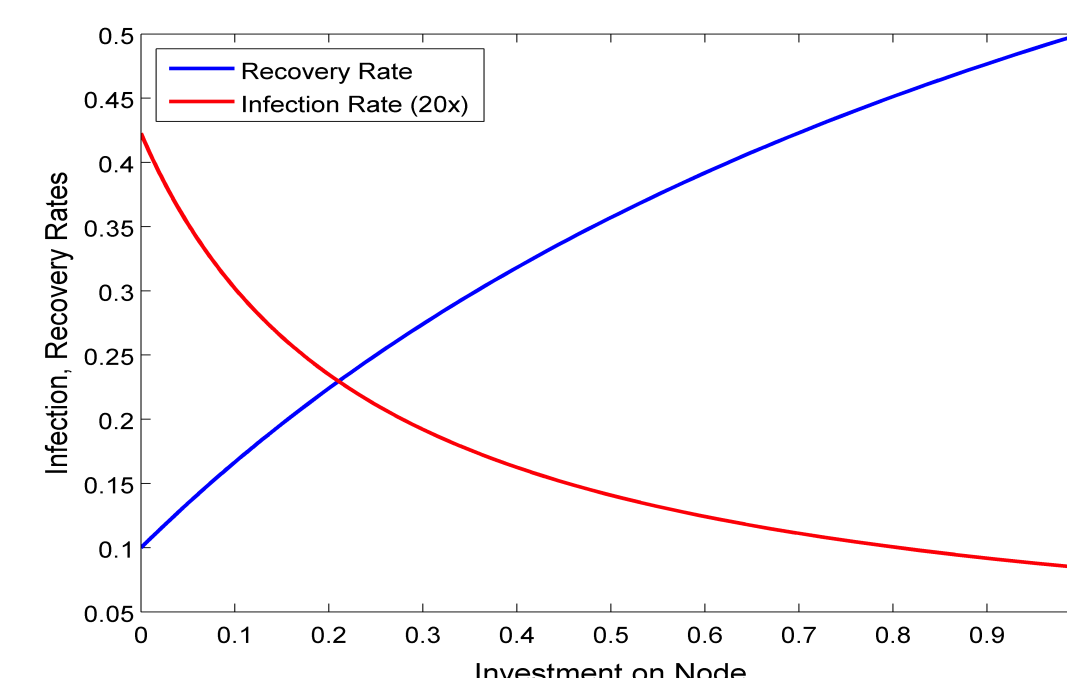


Figure 2: Cost functions for infection and recovery rates

Convex Characterization

Let $M(\mathbf{x})$ comprise posynomials such that $\mathbf{x} \in \mathbb{R}_{++}^k$ and $f_i(\mathbf{x}) \leq 1$, where f_i are posynomials. Then, we can minimize $\lambda_1(M(\mathbf{x}))$ by solving the following Geometric Program

$$\text{minimize}_{\lambda, \{u_i\}_{i=1}^n, \mathbf{x}} \lambda \quad (11)$$

$$\text{subject to } \frac{\sum_{j=1}^n M_{ij}(\mathbf{x}) u_j}{\lambda u_i} \leq 1, \quad i = 1, \dots, n, \quad (12)$$

$$f_i(\mathbf{x}) \leq 1, \quad i = 1, \dots, m. \quad (13)$$

Given a strongly connected graph with posynomial costs, optimal resource allocation for node v_i to solve (6)–(10) are $f_i(\beta_i^*)$ and $g_i(\bar{\Delta} + 1 - \delta_i^*)$, where $\bar{\Delta} \triangleq \max\{\delta_i\}_{i=1}^n$ and β_i^*, δ_i^* are the optimal solution for β_i and δ_i in

$$\text{minimize}_{\{u_i, \beta_i, \delta_i, t_i\}_{i=1}^n} \sum_{k=1}^n f_k(\beta_k) + g_k(t_k) \quad (14)$$

$$\text{subject to } \frac{\beta_i \sum_{j=1}^n A_{ij} u_j + \bar{\delta}_i u_i}{(\bar{\Delta} + 1 - \varepsilon) u_i} \leq 1, \quad (15)$$

$$(t_i + \delta_i) / (\bar{\Delta} + 1) \leq 1, \quad (16)$$

$$\bar{\Delta} + 1 - \delta_i \leq \delta_i \leq \bar{\Delta} + 1 - \delta_i, \quad (17)$$

$$\underline{\beta}_i \leq \beta_i \leq \bar{\beta}_i, \quad i = 1, \dots, n. \quad (18)$$

Illustrations on Global AirTraffic

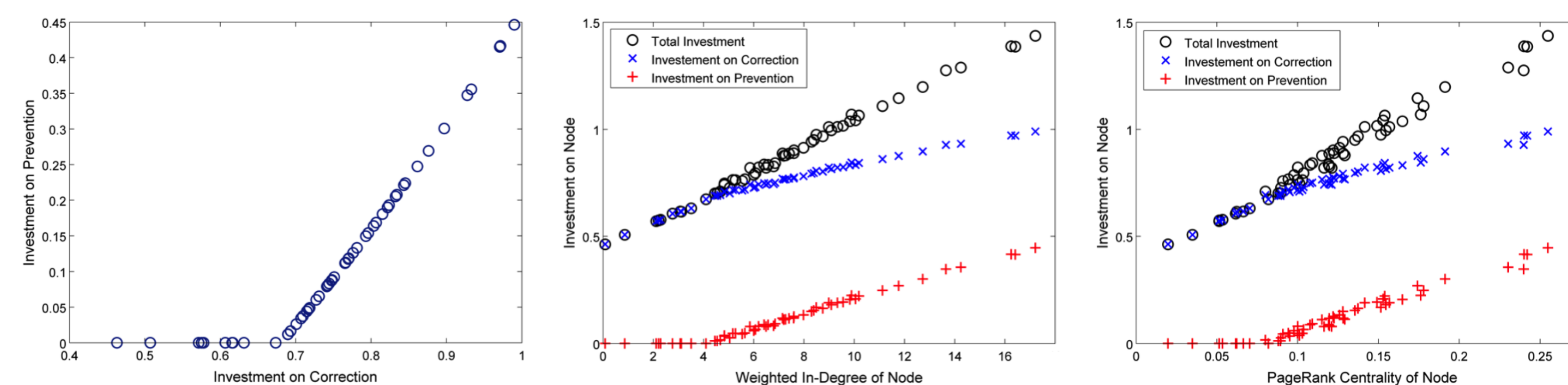


Figure 3: Results from the Budget-constrained resource allocation problem

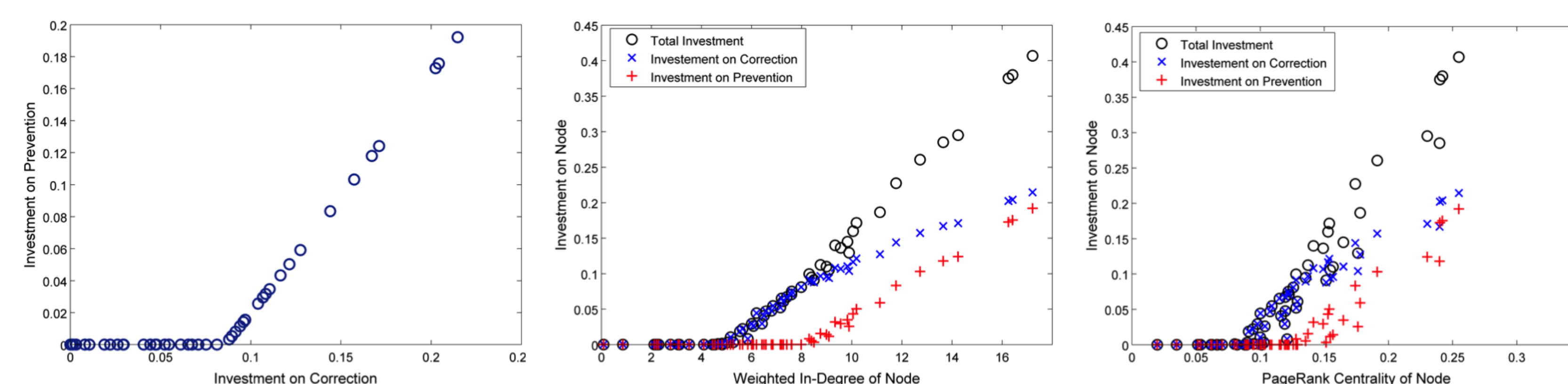


Figure 4: Results from the Rate-constrained resource allocation problem

Simulation parameters

- We applied results to a global air transportation network (56 airports with incoming passenger traffic ≥ 10 million per year)
- Bounds on infection and recovery rates used were chosen so that in the absence of protection resources, $\lambda_1(\beta_i A_G - \delta_i I) = 0.1 > 0$

Conclusions

- A convex framework indicates network centrality can guide allocation scheme, though it's not sufficient
- Objective function determines whether or not one or both resources are applied
- Observed (5 mppy and 8 mppy) thresholds for resource allocation

Challenges in CPNs

- Interdependence of different network layers – cyber (virtual) and physical makes them even more vulnerable.
- CPN configurations are not fixed; hence, not necessarily a one-shot allocation problem

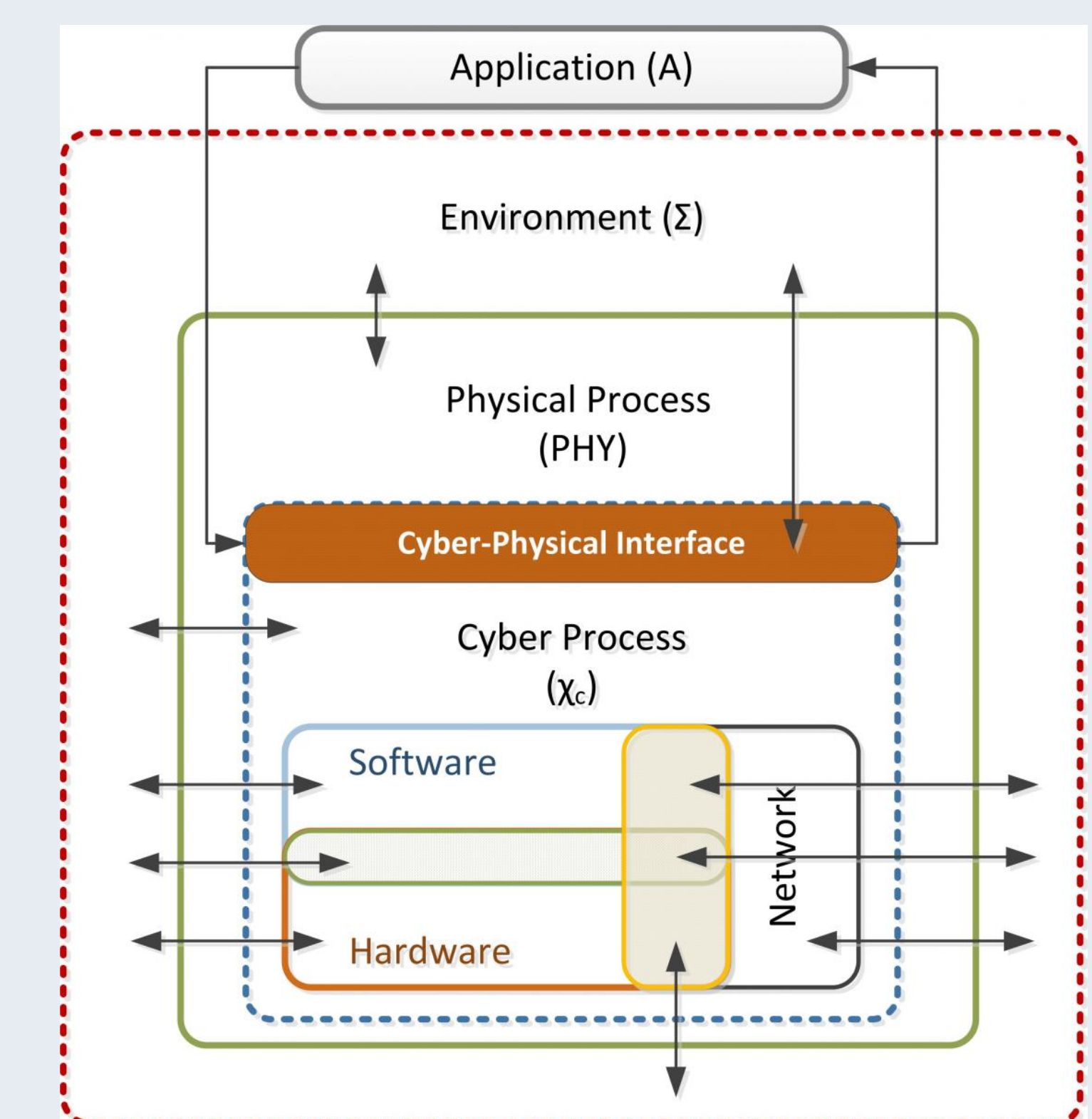


Figure 5: An illustration of layers in CPNs

- Efficient distributed solutions are needed
- Reliability and intermittent information in CPS.
- Allocation strategies that allow for graceful degradation?

References

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