

Resilience Assessment of Hyperconnected Parcel Logistic Networks Under Worst-Case Disruptions

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Joint work with

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Benoit Montreuil

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Georgia Tech College of Engineering

**H. Milton Stewart School of
Industrial and Systems Engineering**



**Physical
Internet
Center**


Problem Motivation

- Parcel delivery industry is one of the **fastest growing industries** in the world.
 - Last year, 87 billion parcels were shipped and delivered.
 - Parcel volume is expected to reach 200 billion in next 5 years.


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 - Asset intensive industry
 - Speed and accuracy of parcel delivery
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- Requires intricate planning
 - Network Design
- Moreover, logistics networks designed with only efficiency considerations
 - Not ideal as disruptions occur

Disruptions

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BUSINESS

FedEx to Ramp Up Spending to Ease Delivery Delays

WORLD ECONOMY

Another shipping crisis looms on

as In southern China

WATERLOO REGION

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MALAYSIA

Parcel Delays Incoming: Floods Causing

Delays in

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- Parcel delivery networks face disruptions:

- Major traffic jams
- Power outages
- Pandemics

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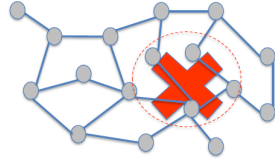
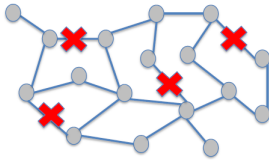


- Disruptions lead to:

- Excess pressure on functional resources
- Late parcel deliveries
- Increased costs

Resilience Evaluation

- **Simulation Models:** Simulation of disruptive events in which the network components fail
 - Random or localized failures
 - Total or partial failures



Resilience Evaluation

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 - Random or localized failures
 - Total or partial failures
- **Analytical models:** Estimate the vulnerability of the network through its structural properties
 - Centrality measures
 - Path lengths, edge-disjoint paths

Stochastic disruptive events only - requires disruption data

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-
- Need to also develop a tool to assess resilience of the **large-scale** networks under such **worst-case disruptive events**
 - Intelligent fictitious adversary (Game-theory based)

Resilient Assessment of Parcel delivery Networks

- **Aim:** To devise a tool that assess the resilience of large-scale logistics networks under worst-case disruptions
 - Through operational costs faced by networks

Resilient Assessment of Parcel delivery Networks

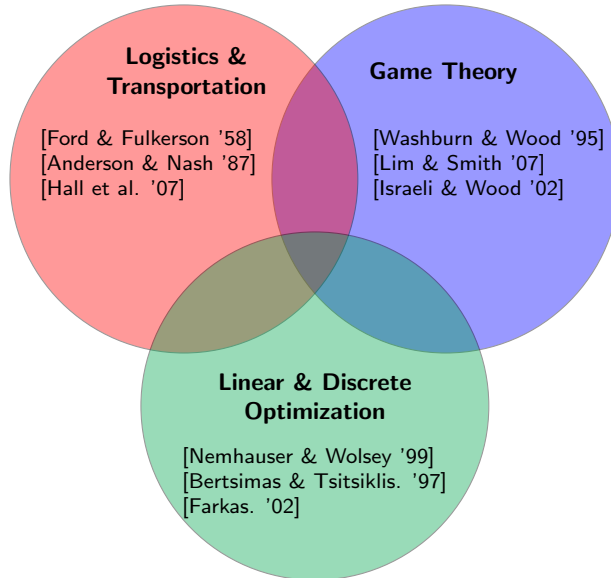
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- We study a **two-person Stackelberg game**:
 - Network components are disrupted to cause most harm (fictitious adversary)
 - Best response to minimize the effects of disruptions (logistics company)
- **Contributions:**
 - Bi-level mixed integer linear program (Network Interdiction Problem)
 - Exact solution technique to assess the resilience
 - Resilience analysis of networks

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- **Agenda:**

To leverage tools from optimization, network science, and game theory to assess the resilience of logistics networks under worst-case disruptions.

Related Work



Problem Setting

- **Goal:**

- ① **Leader:** Interdict edges to maximize the commodity delivery costs between all OD pairs
- ② **Follower:** Minimize the commodity delivery costs after edge-interdiction

Problem Setting

- **Given:**

- Directed Graph $G = (H \sqcup S \sqcup T, A)$

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- **Given:**

- Directed Graph $G = (H \cup S \cup T, A)$
- $P \subseteq S \times T$: Set of Origin-Destination pairs



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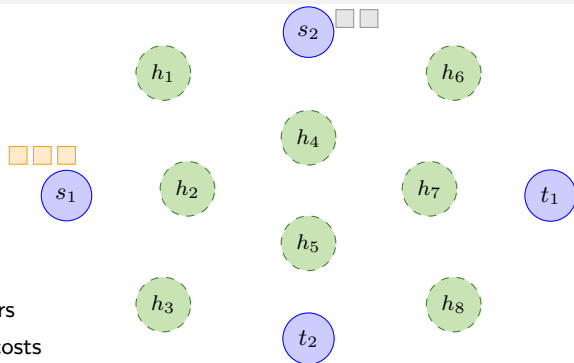
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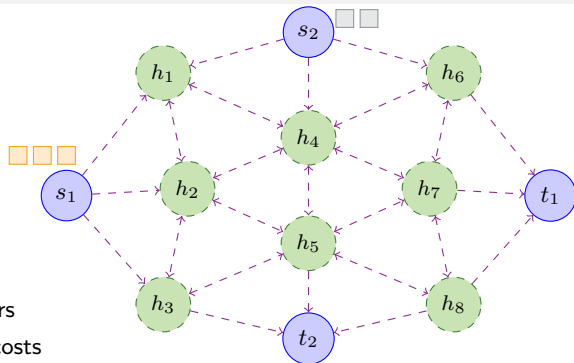
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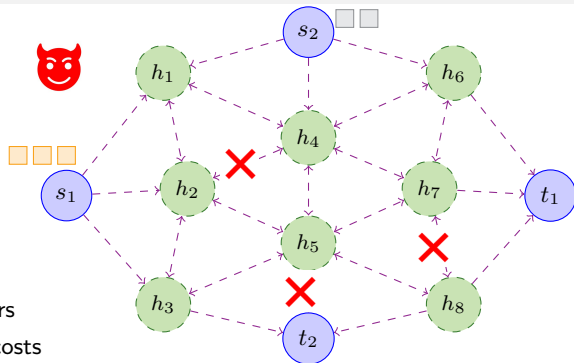
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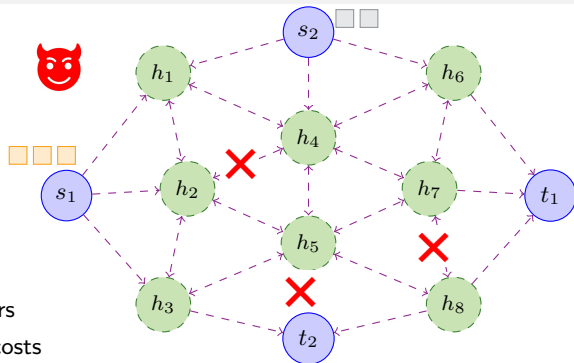
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- $x_{i,j} \in \{0,1\}$: Arc (i,j) is interdicted .



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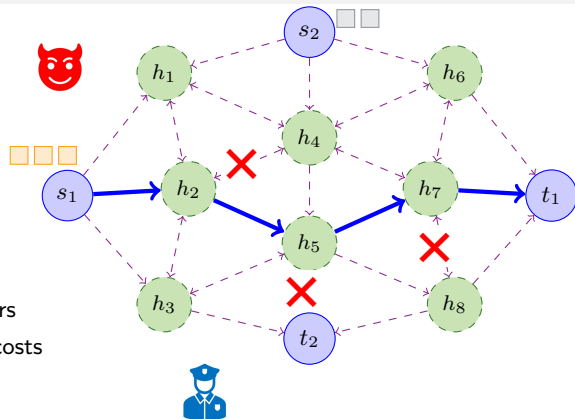
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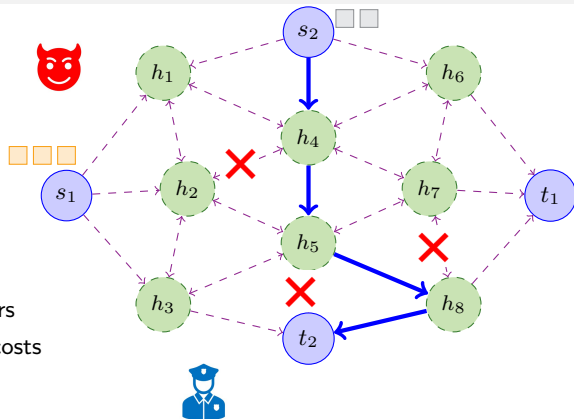
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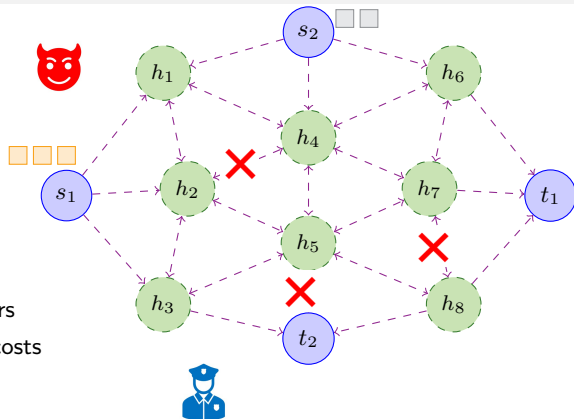
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Goal:

- Leader:** Interdict edges to maximize the commodity delivery costs between all OD pairs
- Follower:** Minimize the commodity delivery costs after edge-interdiction

Decision Variables:

- $x_{i,j} \in \{0, 1\}$: Arc (i, j) is interdicted.
- $f_{i,j}^p \in \mathbb{R}_{\geq 0}$: Commodity flow on arc (i, j) for O-D pair p .



Network Interdiction Problem - Formulation

\mathcal{P} : Set of O-D pairs

\mathcal{H} : Set of hubs

\mathcal{A} : Set of transportation arcs

d_p : Commodity demand
for O-D pair p

Network Interdiction Problem - Formulation

subject to:

$$x_{i,j} \in [0, 1]$$

$$\delta(i, j) \in A$$

Arc interdiction variables

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Network Interdiction Problem - Formulation

subject to:

$$\sum_{(i,j) \in \mathcal{A}} x_{i,j} \leq \beta$$

Interdiction budget

$$x_{i,j} \in [0, 1]$$

$$\delta(i,j) \in \mathcal{A}$$

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$$\max_x$$

Total commodity delivery costs

subject to:

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$$f_{i,j}^p \geq 0,$$

$$\delta(i,j) \in \{0,1\}, \delta p \in P$$

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Commodity flow balance

Interdiction budget

$$f_{i,j}^p \geq 0, \quad \delta(i,j) \in \mathcal{A}, \delta p \in \mathcal{P}$$

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$$\max_x \min_f (c_{i,j} + M x_{i,j}) f_{i,j}^p$$

Total commodity delivery costs

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Dualization Procedure

Bi-level program

s.t:

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Bi-level program

$$\max_x \min_f \sum_{p \in \mathcal{P}} \sum_{(i,j) \in \mathcal{A}} (c_{i,j} + M \cdot x_{i,j}) f_{i,j}^p$$

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$$x_{i,j} \leq \beta \quad (i,j) \in \mathcal{A}$$

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Dual of the inner problem

$$\forall p \in \mathcal{P}$$

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Single-level mixed integer program

$$\max_{x, \pi} \sum_{p \in \mathcal{P}} (\pi_s^p - \pi_t^p)$$

s.t.:

$$\pi_i^p - \pi_j^p \leq c_{i,j} + M \cdot x_{i,j} \quad \forall p \in \mathcal{P}, \forall (i,j) \in \mathcal{A}$$

$$x_{i,j} \leq \beta$$

$$(i,j) \in \mathcal{A}$$

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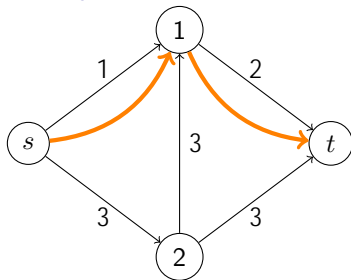
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Large # variables: Problem size reduction required

Edge Interdictions

- Examples:

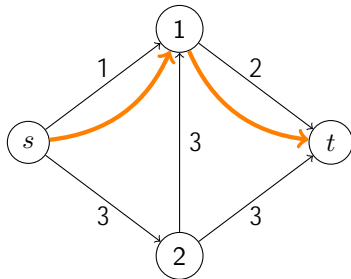


Shortest Path Length: 3

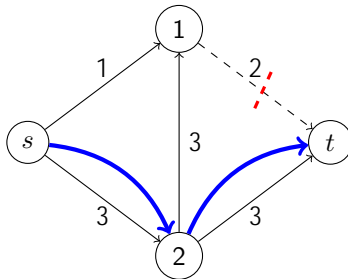
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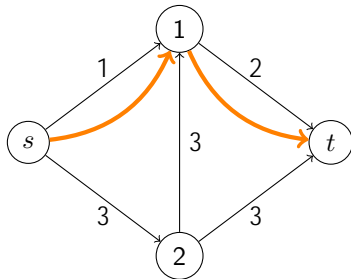
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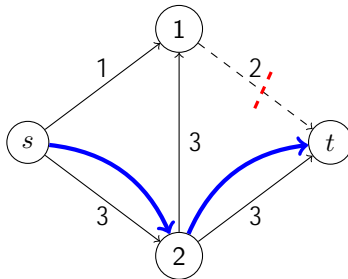
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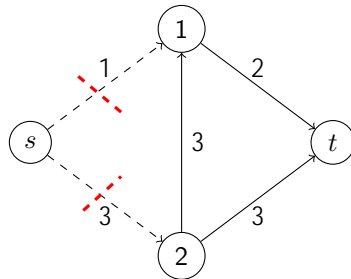
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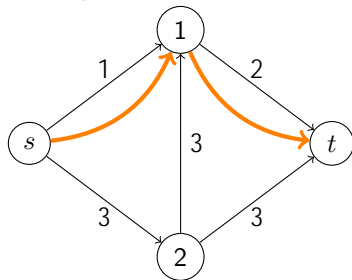
Shortest Path Length: 5
Edge interdictions: 1



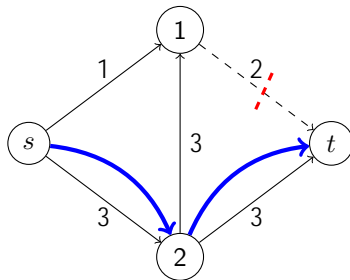
Shortest Path Length: ∞
Edge interdictions: 2

Edge Interdictions

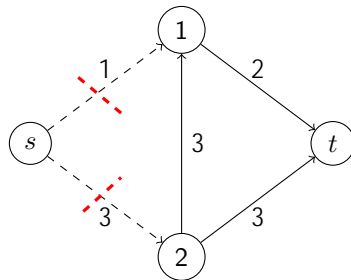
- **Examples:**



Shortest Path Length: 3
Edge interdictions: 0



Shortest Path Length: 5
Edge interdictions: 1



Shortest Path Length: ∞
Edge interdictions: 2

- Which edge(s) to interdict depends upon the **resource availability** at adversary
- **Problem size reduction:** Not generating arc variables that won't be interdicted
 - Search tree to find out such arcs

Case Study

- Major Parcel Delivery Company in China
 - Several millions of parcels handled every week
 - **Implementation Scale:** Central China
- Potential locations for logistics hub construction
 - **Logistics significance:** Major cities, highway intersections, and existing city-based inbound/outbound hubs
- Regulations set by Chinese government
 - 11-hour driving limit per day
 - **Allowable transportation edges:** 5.5 hours of travel time
- Topology-optimized networks designed through minimizing
 - single shortest path - Lean network
 - k -shortest paths
 - k -shortest edge-disjoint path

Computational Performance

Computational Performance

Method
Pre-processing + Gurobi
Gurobi

Computational Performance

Method	# Hubs	$ \mathcal{A} $
Pre-processing + Gurobi	60	1005
	70	1074
	80	1344
	90	1429
Gurobi	60	1005
	70	1074
	80	1344
	90	1429

Computational Performance

Method	# Hubs	$ \mathcal{A} $	$ \mathcal{A}' $
Pre-processing + Gurobi	60	1005	586
	70	1074	656
	80	1344	767
	90	1429	817
Gurobi	60	1005	-
	70	1074	-
	80	1344	-
	90	1429	-

Computational Performance

Method	# Hubs	$ \mathcal{A} $	$ \mathcal{A}' $	# Variables
Pre-processing + Gurobi	60	1005	586	3137
	70	1074	656	3227
	80	1344	767	3312
	90	1429	817	3403
Gurobi	60	1005	-	18207
	70	1074	-	18641
	80	1344	-	21290
	90	1429	-	23388

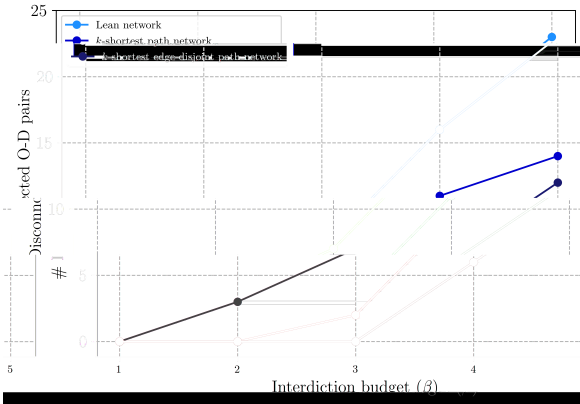
Computational Performance

Method	# Hubs	$ \mathcal{A} $	$ \mathcal{A}' $	# Variables	# Constraints
Pre-processing + Gurobi	60	1005	586	3137	4288
	70	1074	656	3227	4571
	80	1344	767	3312	4836
	90	1429	817	3403	5912
Gurobi	60	1005	-	18207	150,976
	70	1074	-	18641	163,420
	80	1344	-	21290	212,830
	90	1429	-	23388	261,777

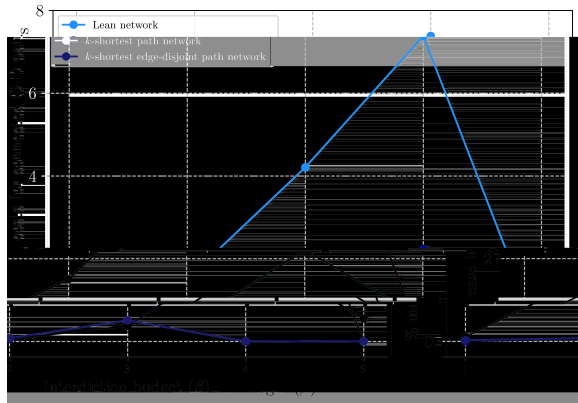
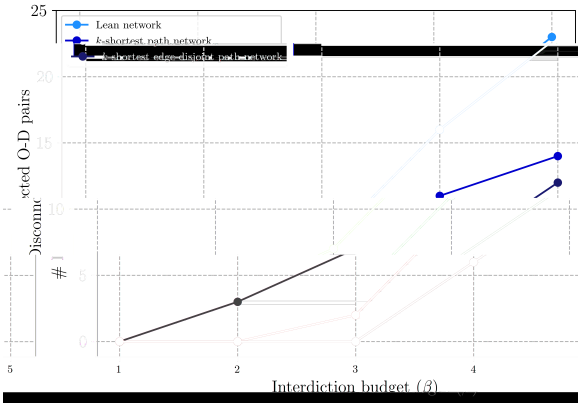
Computational Performance

Method	# Hubs	$ \mathcal{A} $	$ \mathcal{A}' $	# Variables	# Constraints	Total Time (sec)	Optimality Gap (%)
Pre-processing + Gurobi	60	1005	586	3137	4288	54	0
	70	1074	656	3227	4571	59	0
	80	1344	767	3312	4836	62	0
	90	1429	817	3403	5912	82	0
Gurobi	60	1005	-	18207	150,976	time limit	504.5
	70	1074	-	18641	163,420	time limit	831.4
	80	1344	-	21290	212,830	time limit	1392.5
	90	1429	-	23388	261,777	time limit	1965.2

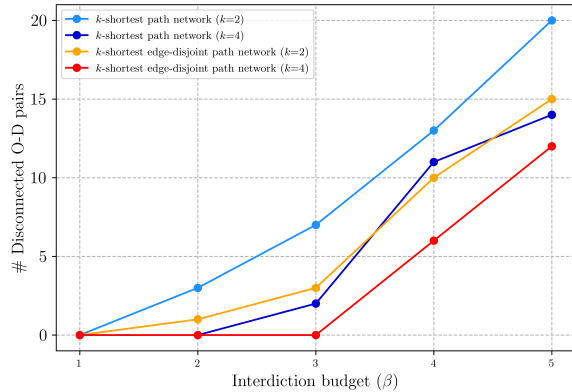
Resilience Analysis of Networks



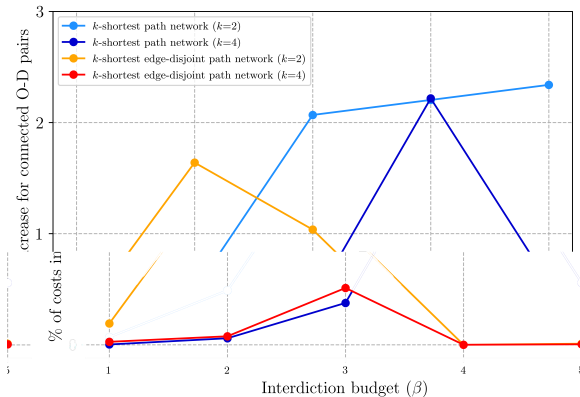
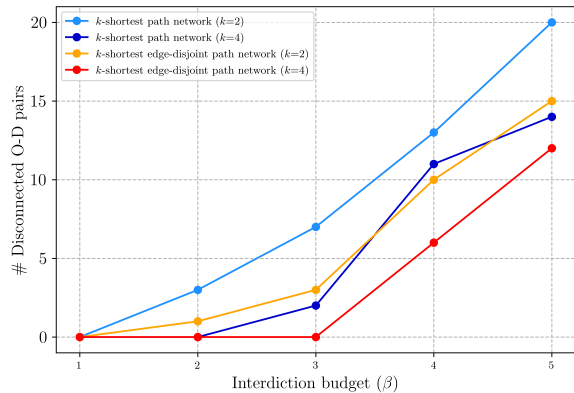
Resilience Analysis of Networks



Resilience Analysis of Networks



Resilience Analysis of Networks



Summary

Contributions:

- Resilience assessment of large-scale networks under worst-case disruptions
- Network interdiction problem with a bi-level mixed integer formulation
- Dualization procedure and search tree strategy to reduce the problem size drastically
- Computational performance comparison of the solution methodology against off-the-shelf solver
- Resilience analysis of the topology-optimized networks

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- Worst-case disruption in Network Design

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Thank you!

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