

# SARNET optimal protection strategies: a modeling approach

*Stojan Trajanovski*

University of Amsterdam, The Netherlands

Presented by dr. Paola Grosso

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# Outline

Motivation

Model

Problem

Algorithms

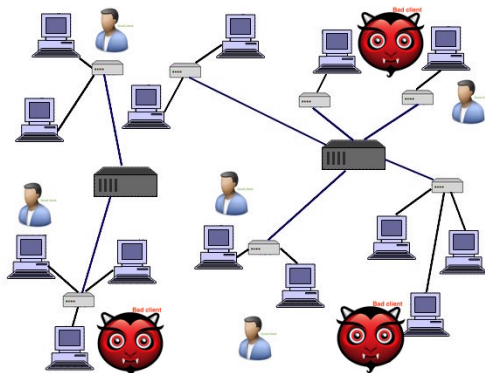
Some Results

Conclusions

# Motivation

## Modeling DDoS attacks

1. A given topology with nodes and links
2. Nodes = {good clients, bad clients, service nodes, other "routers"}
3. Boost the "good traffic" (good clients  $\rightarrow$  service nodes)
4. Shrink the "bad traffic" (bad clients  $\rightarrow$  service nodes)



# Motivation

## ***How to react?* - Finding optimal response**

1. current topology (nodes, links & interconnections)
2. permitted links (that can be made **on** or **off**)
3. current values and box constrains on the link bandwidths
4. filtering a certain flow
5. already determined **good clients** and **bad/attackers**
6. already determined **service nodes**

## **Multiobjective nature:**

1. maximize the flow from **good clients** to the **service nodes**
2. minimize the flow from **attackers** to the **service nodes**
3. under the given constrains

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# Model

## Representing the network as a graph: (given inputs)

- ▶ directed graph  $G = (\mathcal{N}, \mathcal{L})$  with a set of nodes  $\mathcal{N}$  and a set of links  $\mathcal{L} = \{(i, j) | i, j \in \mathcal{N}\}$
- ▶  $k_{ij} \in \{0, 1\}$  represents the initial link  $(i, j)$  presence
- ▶  $c_{ij}$  and  $c_{ij}^{\max}$  are the starting and the maximum allowed capacities of link  $(i, j)$ , respectively
- ▶  $\mathcal{C} \subset \mathcal{N}$  is the set of clients,  $\mathcal{A} \subset \mathcal{N}$  is the set of attackers, and  $\mathcal{S} \subset \mathcal{N}$  are service nodes

**The aim** is to **maximize** the successful flow from the nodes in  $\mathcal{C}$  to  $\mathcal{S}$  to **minimize/protect** from the flow from  $\mathcal{A}$  to  $\mathcal{S}$  with a minimum cost

## Permitted actions:

1. link delete or add (not for all pairs of nodes)
2. bandwidth up or down
3. flow filtering

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# Problem definition

## Decision variables:

$f_{ijkm} \in \mathbb{R}$  is a part of the flow on link  $(i, j)$  carrying a traffic from  $k$  to  $m$

$l_{ij}^+ \in \{0, 1\}$  is 1 if link  $(i, j)$  has been added and 0 otherwise

$l_{ij}^- \in \{0, 1\}$  is 1 if link  $(i, j)$  has been removed and 0 otherwise

$l_{ij} \in \{0, 1\}$  is 1 if link  $(i, j)$  is present in the network and 0 otherwise

$z_{ij} \in \mathbb{R}$  is the increase/decrease of the bandwidth on the link  $(i, j)$

**Maximize** the flow from  $\mathcal{C}$  to  $\mathcal{S}$  and **minimize** the flow from  $\mathcal{A}$  to  $\mathcal{S}$

We formulated **Mixed Bi-linear Integer Programming (MBIP)** optimization problem. MBIP are usually hard to solve.



# Problem definition

**Objective function:**

$$\max \quad \alpha \sum_{c \in \mathcal{C}, i \in \mathcal{N}, k \in \mathcal{N}, m \in \mathcal{N}} f_{cikm} - \beta \sum_{a \in \mathcal{A}, i \in \mathcal{N}, k \in \mathcal{N}, m \in \mathcal{N}} f_{aikm} \quad (1)$$

**Constraints:**

$$\forall i \in \mathcal{N}, k \in \mathcal{N}, m \in \mathcal{N}, \quad \sum_{j: (i,j) \in \mathcal{L}} f_{ijkm} - \sum_{j: (j,i) \in \mathcal{L}} f_{jikm} = 0, \quad (2)$$

$$\forall (i,j) \in \mathcal{L}, \quad \sum_{k \in \mathcal{N}, m \in \mathcal{N}} f_{ijkm} \leq c_{ij}^{\max} \quad (3)$$

$$\forall (i,j) \in \mathcal{L}, \quad \sum_{k \in \mathcal{N}, m \in \mathcal{N}} f_{ijkm} - z_{ij} = c_{ij} \quad (4)$$

# Problem definition

$$\forall (i, j) \in \mathcal{L}, k \in \mathcal{N}, m \in \mathcal{N}, f_{ijkm} \geq 0 \quad (5)$$

$$\sum_{(i, j) \in \mathcal{L}} l_{ij}^+ \leq C_{\text{add}} \quad (6)$$

$$\sum_{(i, j) \in \mathcal{L}} l_{ij}^- \leq C_{\text{rem}} \quad (7)$$

$$\sum_{(i, j) \in \mathcal{L}, k \in \mathcal{N}, m \in \mathcal{N}} a_{ijkm} \leq C_{\text{filter}} \quad (8)$$

$$\forall (i, j) \in \mathcal{L}, (1 - l_{ij}) \sum_{k \in \mathcal{N}, m \in \mathcal{N}} f_{ijkm} = 0 \quad (9)$$

$$\forall (i, j) \in \mathcal{L}, k \in \mathcal{N}, m \in \mathcal{N}, a_{ijkm} f_{ijkm} = 0 \quad (10)$$

$$\forall (i, j) \in \mathcal{L}, k \in \mathcal{N}, m \in \mathcal{N}, a_{ijkm} - l_{ij} \leq 0 \quad (11)$$

$$\forall (i, j) \in \mathcal{L}, l_{ij} + l_{ij}^- \leq 1 \quad (12)$$

$$\forall (i, j) \in \mathcal{L}, l_{ij}^+ - l_{ij} \leq 0 \quad (13)$$

$$\forall (i, j) \in \mathcal{L}, (1 - l_{ij}) f_{ij} = 0 \quad (14)$$

$$\forall (i, j) \in \mathcal{L}, (l_{ij} + l_{ij}^- - k_{ij})(l_{ij} - l_{ij}^+ - k_{ij}) = 0 \quad (15)$$

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# Algorithms

## Two algorithms:

1. Close-to-exact algorithm (branch & bound)
2. Dedicated heuristic
3. Performance and running time analysis between both

# Close-to-exact algorithm (1)

## Concept of the algorithm

- ▶ non-linear and non-convex constraints, hence hard to solve
  1. there are known special case instances that are NP-hard!
  2. formal proof for more would be a contribution
- ▶ it can still be found close-to-optimal solution!
  1. based on the MBIP formulation
  2. non-polynomial algorithm
  3. branch & bound techniques
- ▶ using **yalmip** in Matlab (that unites several optimization packages)
  - ▶ CPLEX (IBM)
  - ▶ MOSEK
  - ▶ GUROBI
  - ▶ SeDuMi

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# Dedicated heuristic (2)

## Concept of the algorithm

- ▶ it is heuristic, but polynomial time!
- ▶ based on the links "centralities" regarding the flows
- ▶ greedy in nature
- ▶ Overview:
  1. for each potential link, if added in the network
    - 1.1 calculate all pairs max flow for each source and destination (\*)
    - 1.2 compute the weighted objective sum/function (\*\*)
    - 1.3 sort the weighted sums list in descending order (\*\*\*)
  2. for each existing link, if removed from the network
    - 2.1 do (\*), (\*\*) and (\*\*\*) from above
  3. try adding links from the sorted list in 2. until:
    - (i) there is an improvement in the weighted sum/objective flow
    - (ii) there are no more links than the given maximum  $C_{add}$
  4. try removing links from the sorted list in 2. until:
    - (i) there is an improvement in the weighted sum/objective flow
    - (ii) there are no more links than the given maximum  $C_{rem}$
  5. calculate the weighted sum/objective flow with the obtained topology (no link addition/removal constrains)

## Dedicated heuristic (2): 1/out of 3

### Pseudo code

```
addedLinks  $\leftarrow$  [];  
removedLinks  $\leftarrow$  [];  
for each  $l \in \mathcal{L}$  do  
  tempG  $\leftarrow$  G;  
  totalFlow  $\leftarrow$  0;  
  if  $l$  does not exist in G then  
    for  $r \in$  requests do  
      maxFlow  $\leftarrow$  maxFlow(tempG,r);  
      currentFlow  $\leftarrow$  flow(start(r),end(r));  
      if start(r)  $\in$  "Good clients" then  
        totalFlow  $\leftarrow$  totalFlow +  $\alpha$  currentFlow;  
      else if start(r)  $\in$  "Bad clients" then  
        totalFlow  $\leftarrow$  totalFlow -  $\beta$  currentFlow;  
    end  
    add ( $l$ , totalFlow) in addedLinks;  
  else  
    /* similar code for removedLinks */  
  end  
end
```

## Dedicated heuristic (2): 2/out of 3 (cont.)

```
takeDescendingLinks(addedLinks, Cadd); /*the highest traffic Cadd links*/
takeDescendingLinks(removedLinks, Crem); /*the high. traffic Crem links*/
currentFlow  $\leftarrow$  weightedObjectivemaxFlow(G); tempG  $\leftarrow$  G;
for each entry  $\in$  addedLinks do
|   totalFlow  $\leftarrow$  0; tempG  $\leftarrow$  G + entry.link;
|   for r  $\in$  requests do
|   |   maxFlow  $\leftarrow$  maxFlow(tempG,r);
|   |   currentFlow  $\leftarrow$  flow(start(r),end(r));
|   |   if start(r)  $\in$  "Good clients" then
|   |   |   totalFlow  $\leftarrow$  totalFlow +  $\alpha$  currentFlow;
|   |   else if start(r)  $\in$  "Bad clients" then
|   |   |   totalFlow  $\leftarrow$  totalFlow -  $\beta$  currentFlow;
|   end
|   if totalFlow > currentFlow then
|   |   currentFlow  $\leftarrow$  totalFlow; G  $\leftarrow$  tempG;
|   else
|   |   break;
|   end
end
```



## Dedicated heuristic (2): 3/out of 3 (cont.)

```
tempG  $\leftarrow$  G;  
for each entry  $\in$  removedLinks do  
  | /* similar consecutive removal as the addition in the previous slide */  
end  
currentFlow  $\leftarrow$  weightedObjectivemaxFlow(G);  
return G, addedLinks, removedLinks, currentFlow;
```

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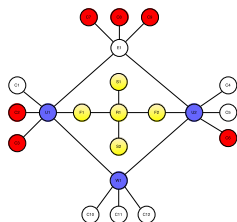
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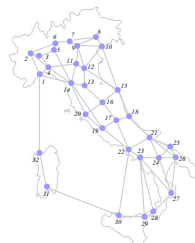
# Used topologies



(a) SARNET topology.



(b) ARPANET.



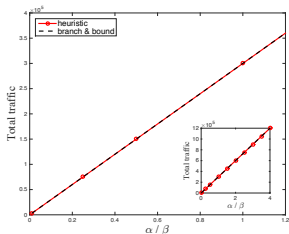
(c) Italian backbone network.

Figure: Used topologies.

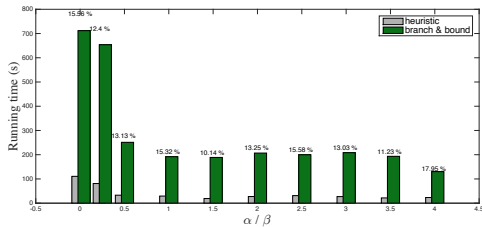
Table: Real networks used in the evaluation.

Networks	$N$	$L$	Description
SARNET	21	22	the project topology
ARPANET	20	32	first packet switching network
ITALY	32	62	main fiber connections in Italy

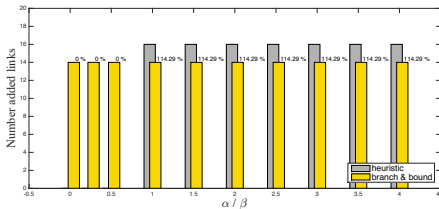
# Results (SARNET, dense topology)



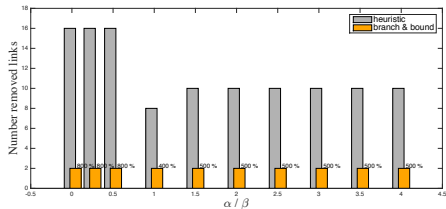
(a) Weighted flow sum.



(b) Running time.



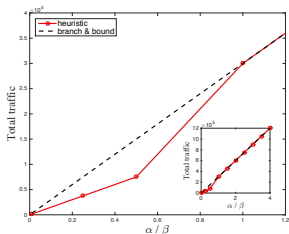
(c) Number of added links.



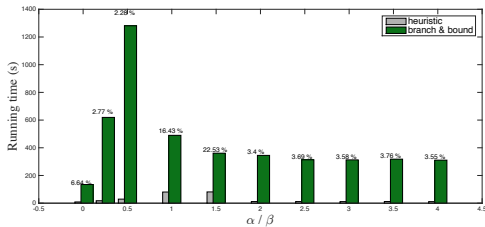
(d) Number of removed links.

Figure: SARNET comparison (dense topology).

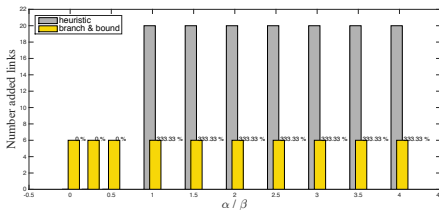
# Results (SARNET, sparse topology)



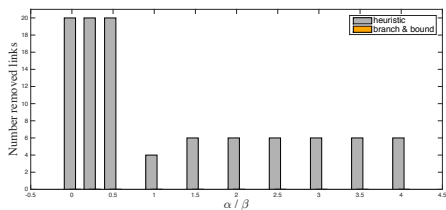
(a) Weighted flow sum.



(b) Running time.



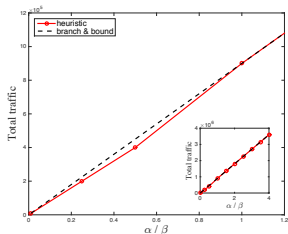
(c) Number of added links.



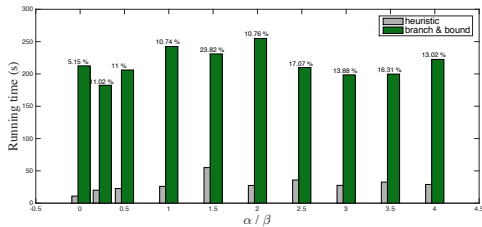
(d) Number of removed links.

Figure: SARNET comparison (sparse topology).

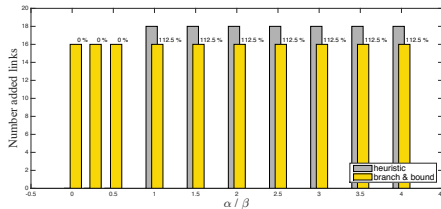
# Results (ARPANET, dense topology)



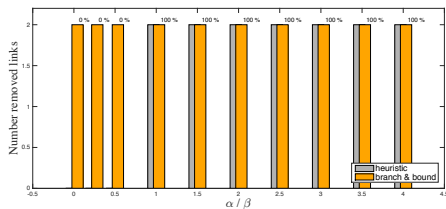
(a) Weighted flow sum.



(b) Running time.



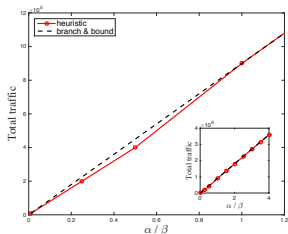
(c) Number of added links.



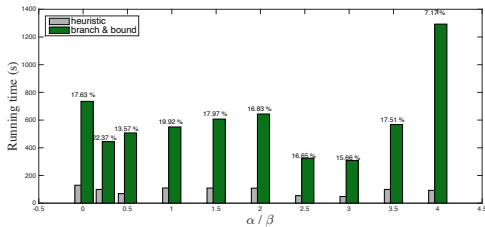
(d) Number of removed links.

Figure: ARPANET comparison (dense topology).

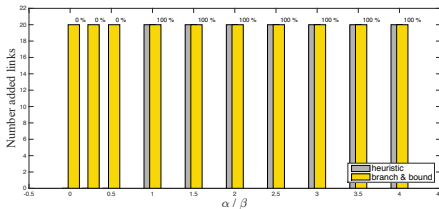
# Results (ARPANET, sparse topology)



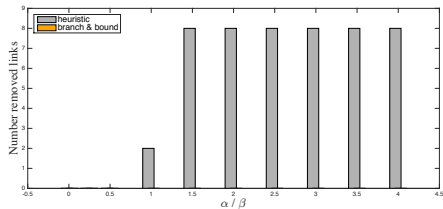
(a) Weighted flow sum.



(b) Running time.



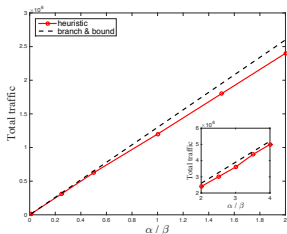
(c) Number of added links.



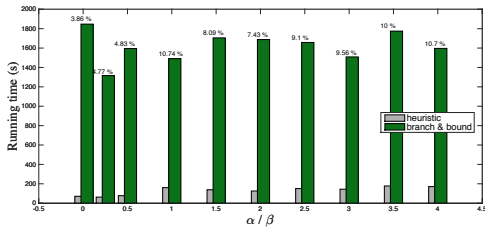
(d) Number of removed links.

Figure: ARPANET comparison (sparse topology).

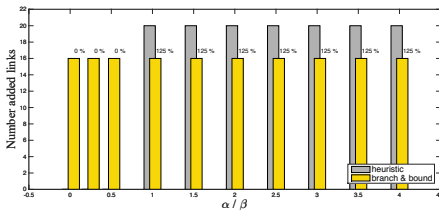
# Results (Italy, dense topology)



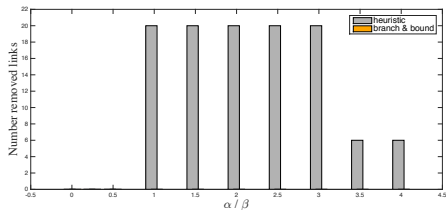
(a) Weighted flow sum.



(b) Running time.



(c) Number of added links.

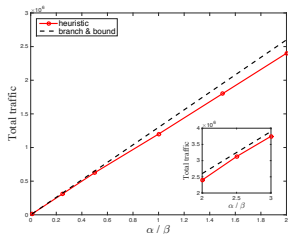


(d) Number of removed links.

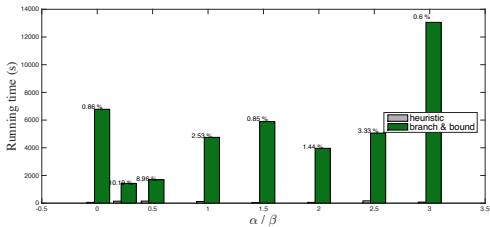
Figure: Italy comparison (dense topology).



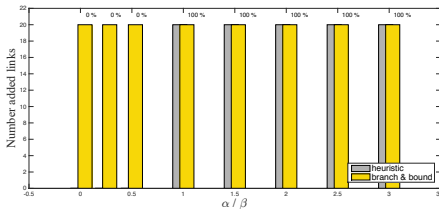
# Results (italy, sparse topology)



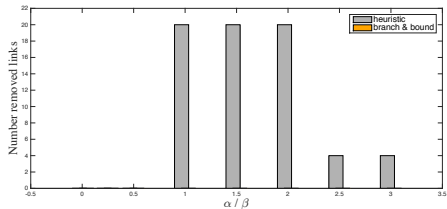
(a) Weighted flow sum.



(b) Running time.



(c) Number of added links.



(d) Number of removed links.

Figure: Italy comparison (sparse topology).

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## Contributions

- ▶ flow network models have been proposed
- ▶ two algorithms for solving the problem:
  1. Close-to-exact algorithm (branch & bound, bi-linear mixed programming)
  2. Dedicated greedy heuristic
- ▶ the greedy heuristic shows **surprisingly** good performance
  1. the two algorithms are **closed** in objective function performance (especially for  $\alpha \geq \beta$ )
  2. the heuristic is **significantly faster** than the MBIP - by factors 5 - 10
  3. 2 algorithms give different solutions:
    - (i) numbers of added links similar for  $\alpha \geq \beta$
    - (ii) MBIP tends to not added as many links as the heuristic!  
*reason the removal appears after the addition, hence "most of the job has been done" - perhaps trying variants*

# Conclusions

## Possible future steps

- ▶ complexity of the problem
  1. known to be NP-hard for the general case
  2. proving the NP-hardness on some particular cases (only link addition or removal ...)
- ▶ integration with the SC demo and the real response of the strategies
- ▶ modeling the inter-domain issues
- ▶ modeling the virtualization

# Questions?



email: [s.trajanovski@uva.nl](mailto:s.trajanovski@uva.nl)