

# Some applications of biased random-key genetic algorithms in telecommunications

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# Summary

- Specifying a biased random-key genetic algorithm
- Applications in telecommunications
  - Routing in IP networks
  - Design of survivable IP networks with composite links
  - Redundant server location for content distribution
  - Regenerator location
  - Routing & wavelength assignment in optical networks
- Concluding remarks

# Reference



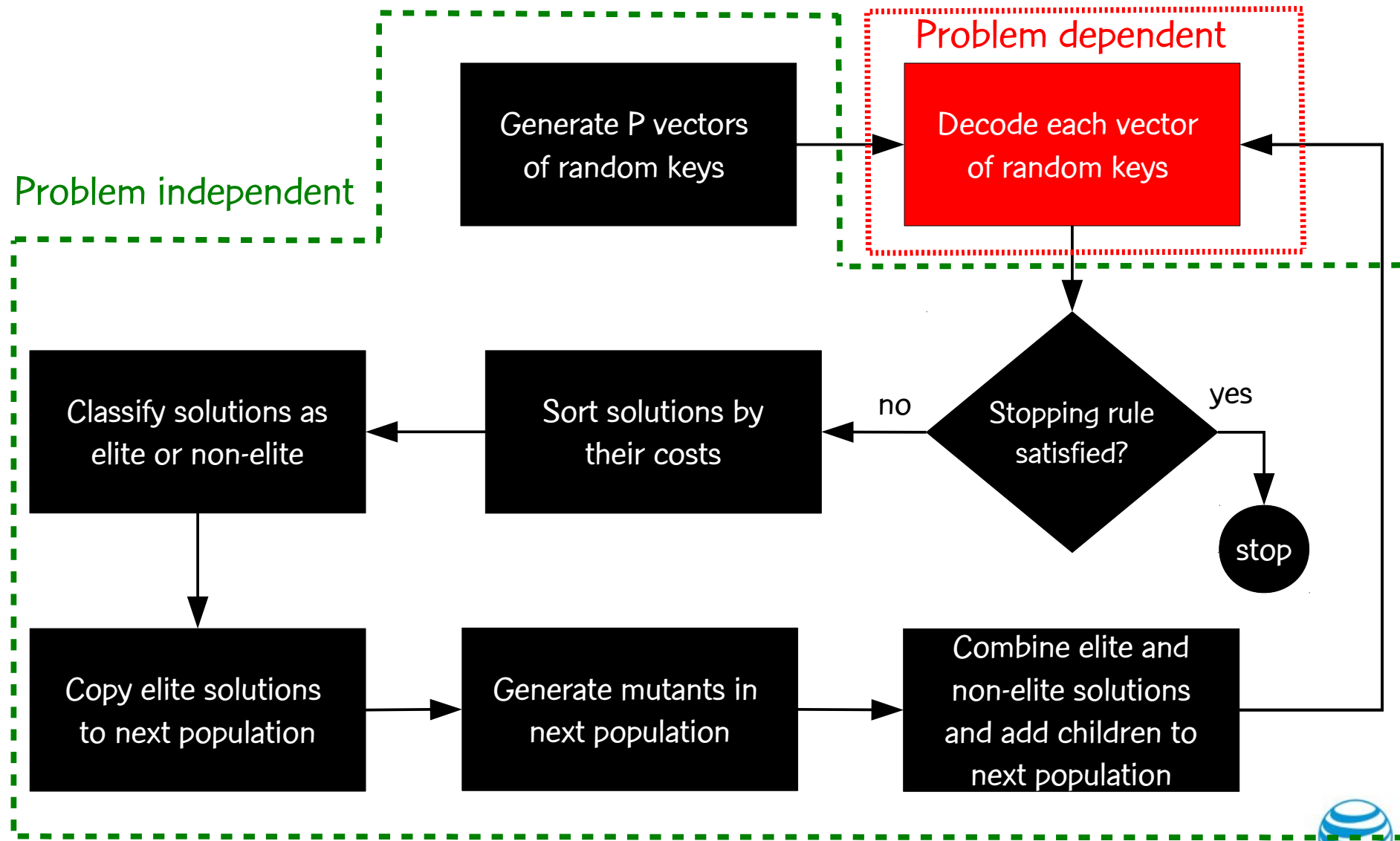
M.G.C.R., “Biased random-key genetic algorithms with applications in telecommunications,” TOP, vol. 20, pp. 120-153, 2012.

Tech report version:

<http://www2.research.att.com/~mgcr/doc/brkga-telecom.pdf>

# Specifying a biased random-key genetic algorithms

# Framework for biased random-key genetic algorithms





# Specifying a biased random-key GA

- Encoding is always done the same way, i.e. with a vector of  $N$  random-keys (parameter  $N$  must be specified)
- Decoder that takes as input a vector of  $N$  random-keys and outputs the corresponding solution of the combinatorial optimization problem and its cost (this is usually a heuristic)
- Parameters:
  - Size of population
  - Size of elite partition
  - Size of mutant set
  - Child inheritance probability
  - Stopping criterion

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  - Stopping criterion: e.g. time, # generations, solution quality, # generations without improvement

# Applications in telecommunications

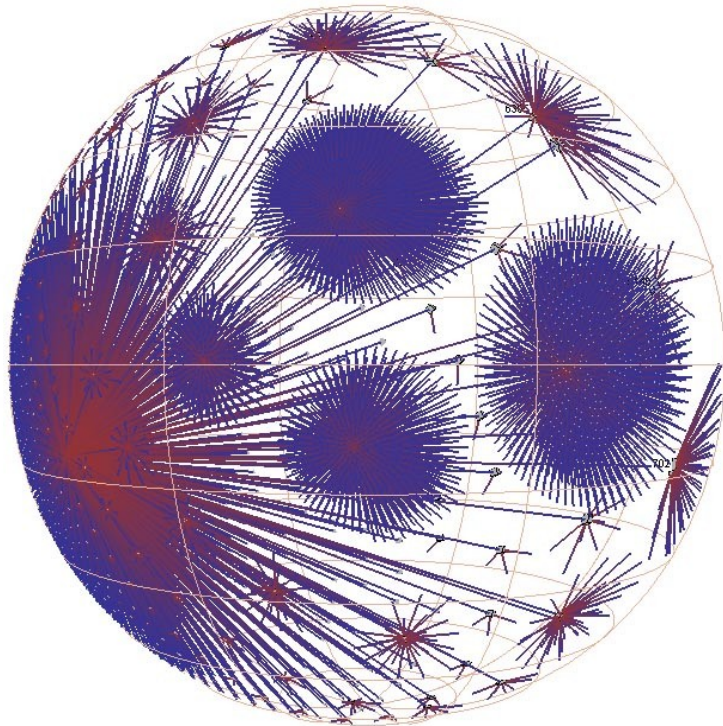
# Applications in telecommunications

- Routing in IP networks
- Design of survivable IP networks
- Redundant server location for content distribution
- Regenerator location
- Routing and wavelength assignment in optical networks

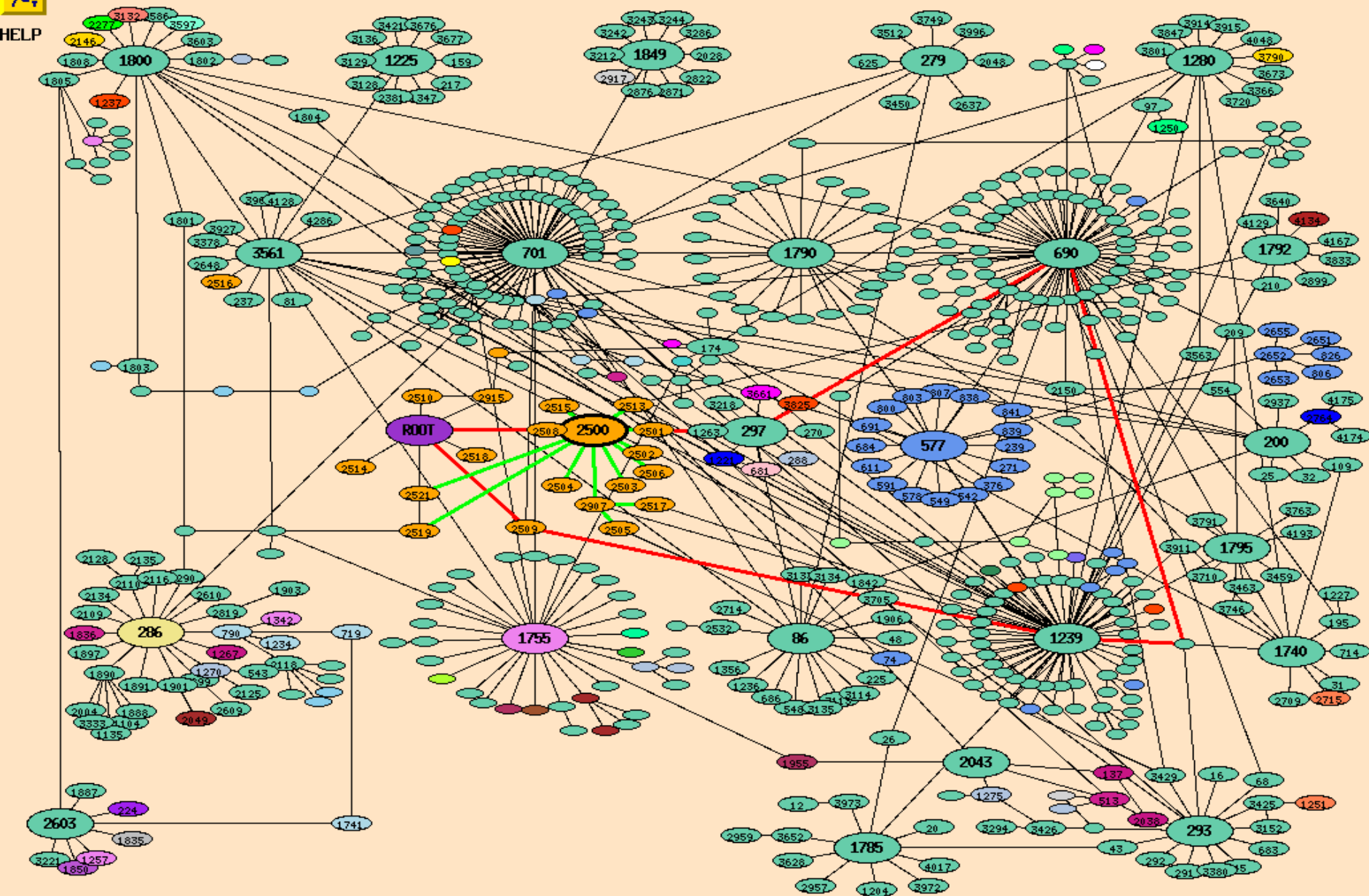
# OSPF routing in IP networks



# The Internet



- The Internet is composed of many (inter-connected) autonomous systems (AS).
- An AS is a network controlled by a single entity, e.g. ISP, university, corporation, country, ...



# Routing

- A packet is sent from a origination router S to a destination router T.
- S and T may be in
  - same AS:
  - different ASes:

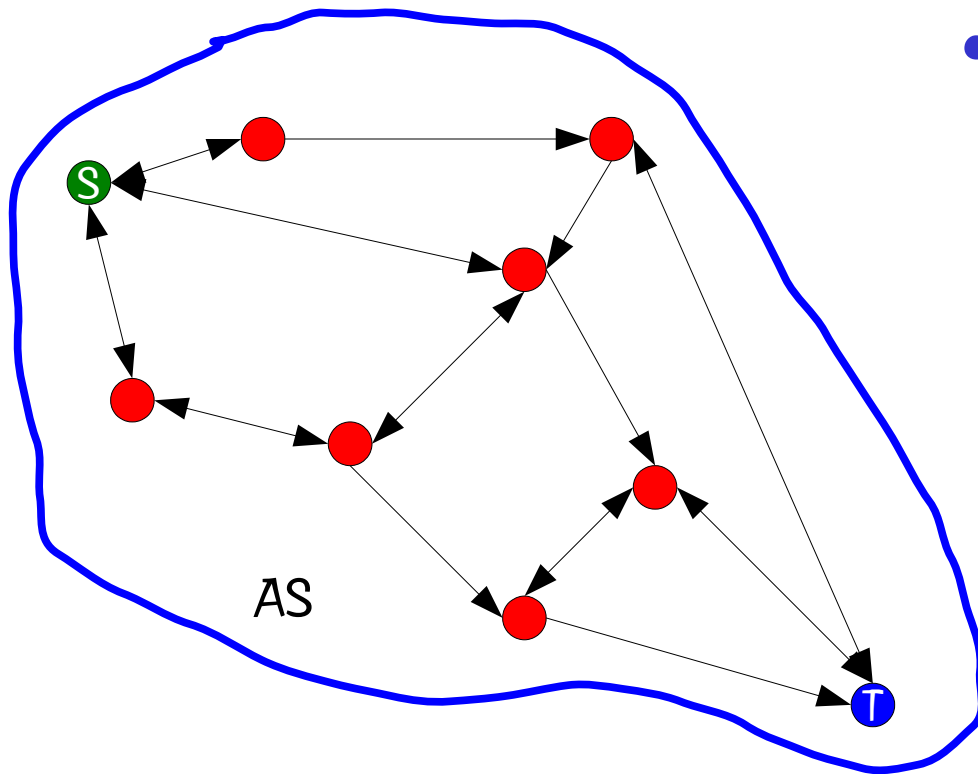
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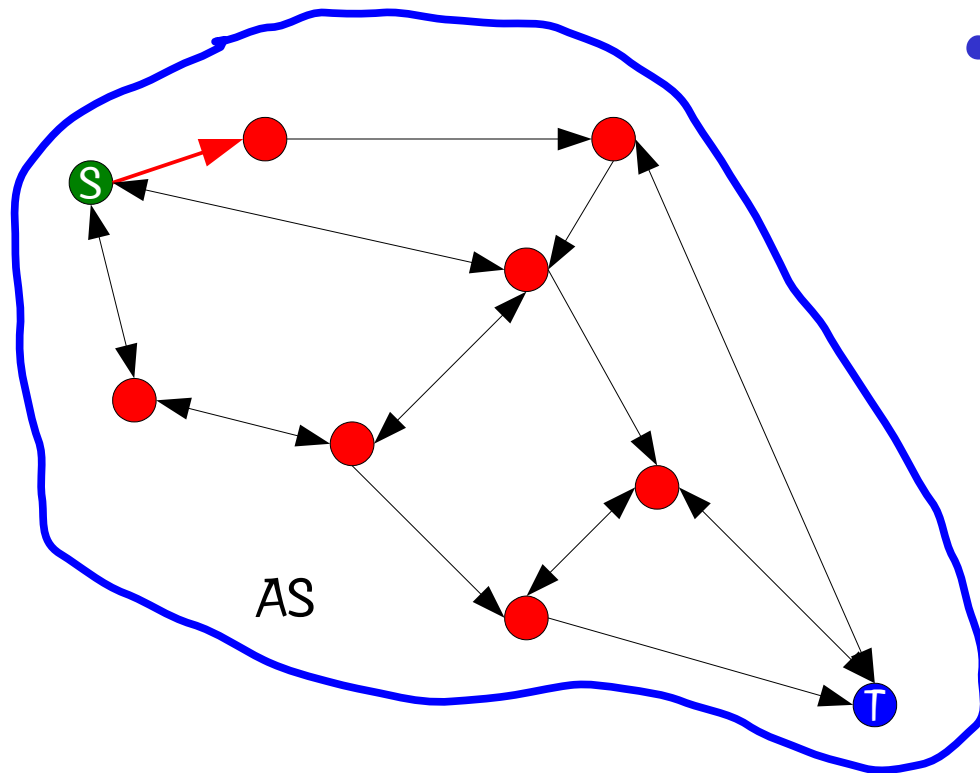
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- IGP (interior gateway protocol) routing is concerned with routing within an AS.

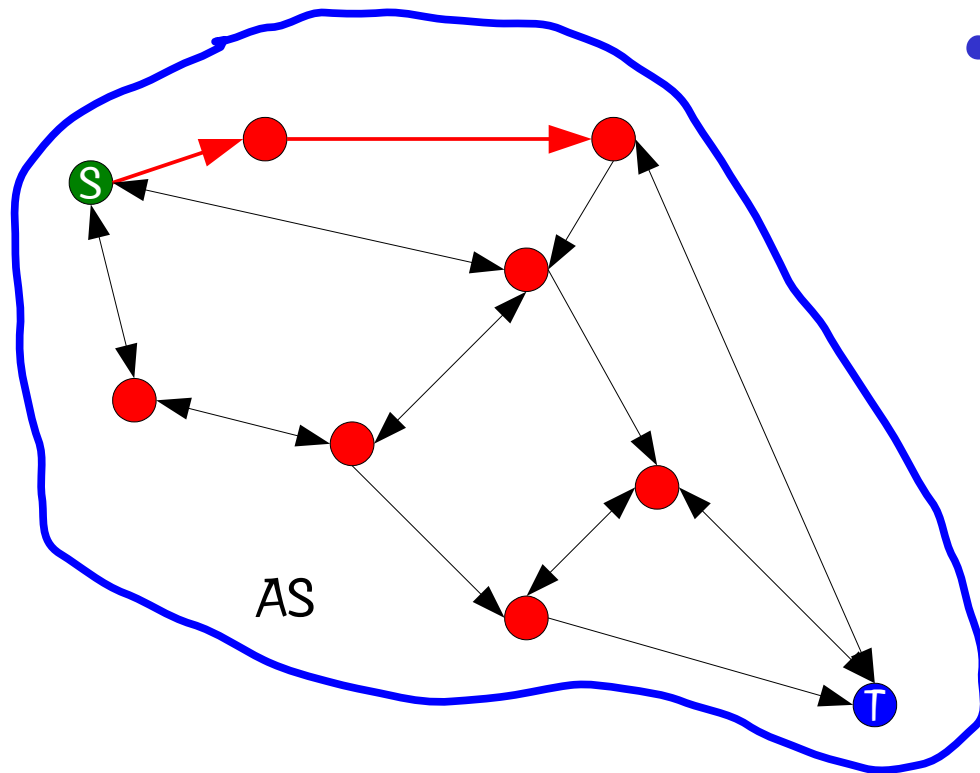


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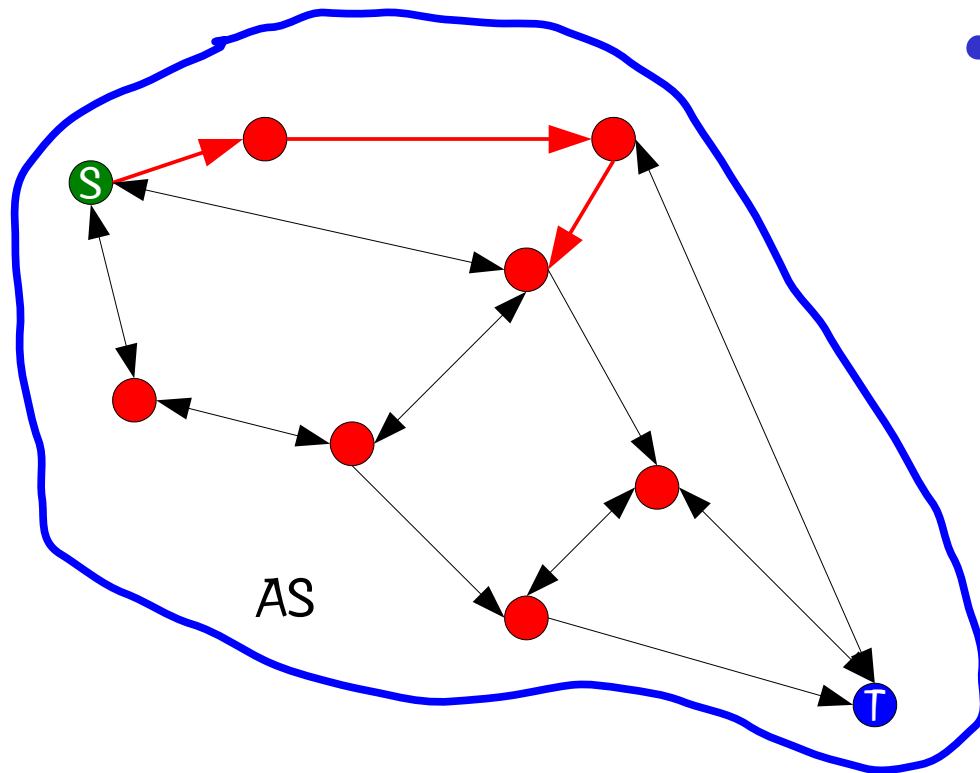
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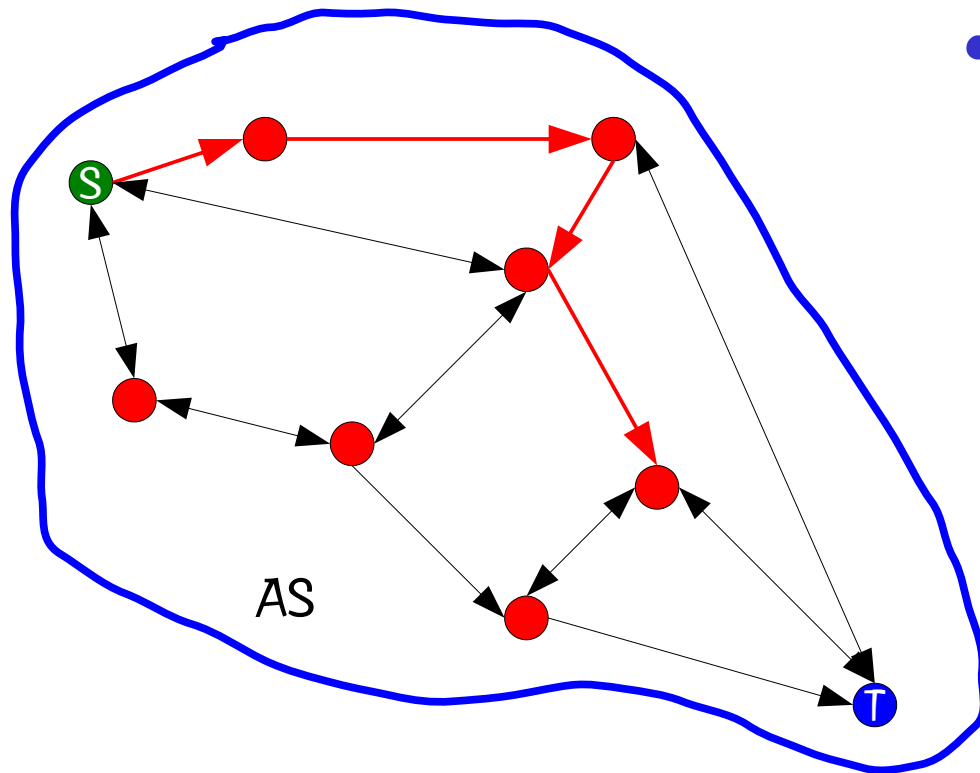
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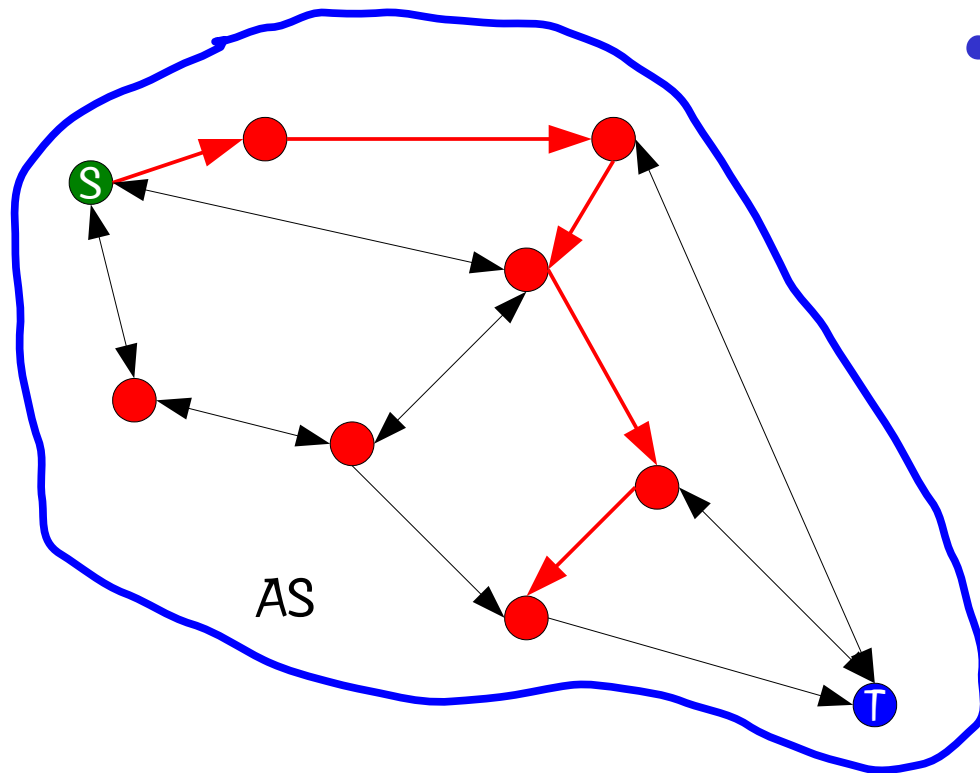
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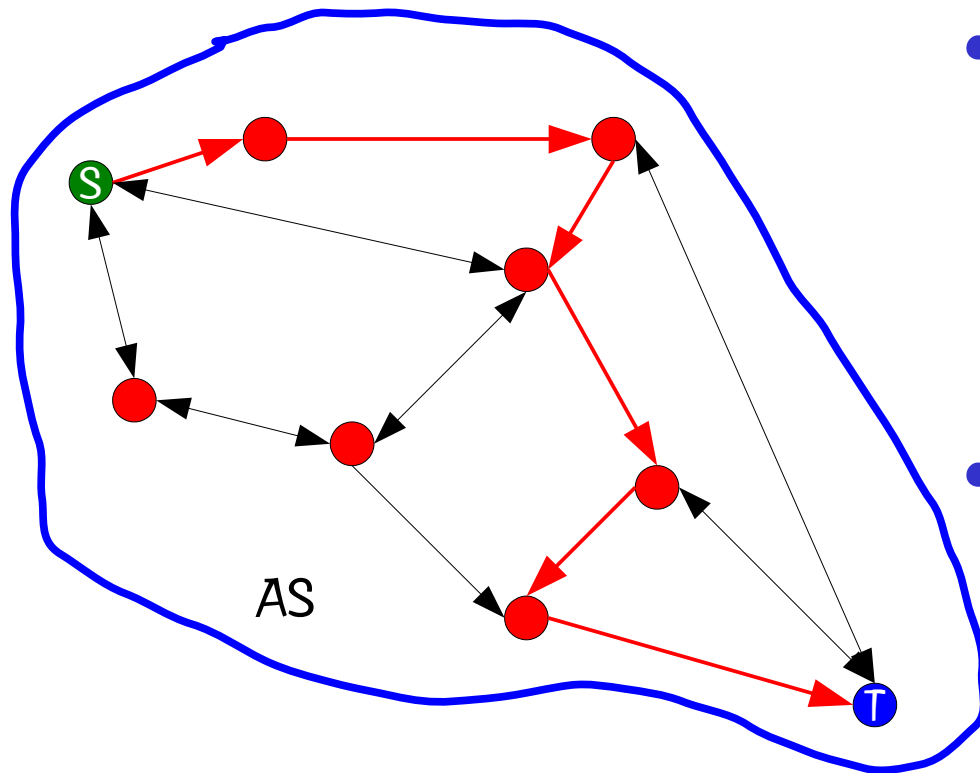
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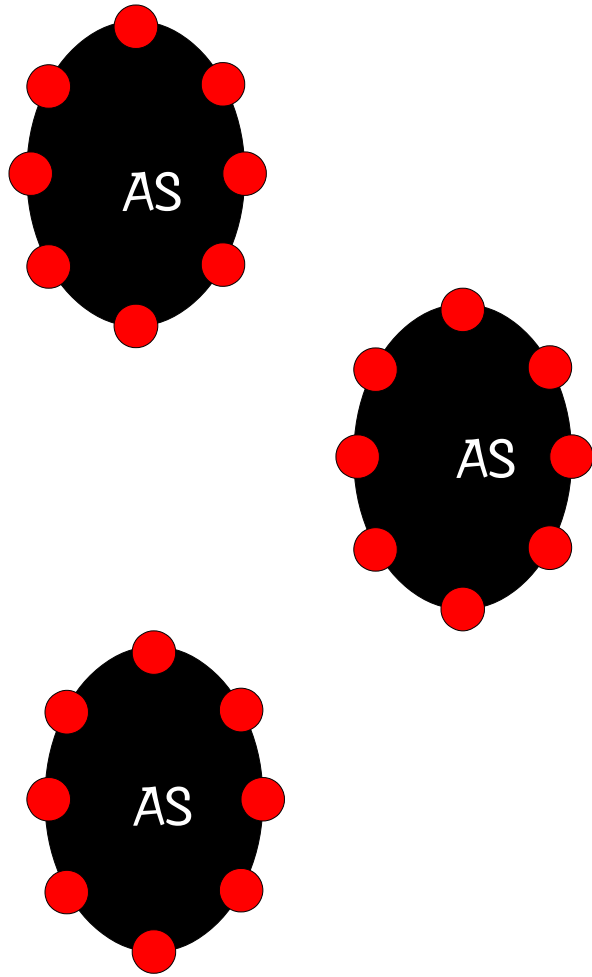
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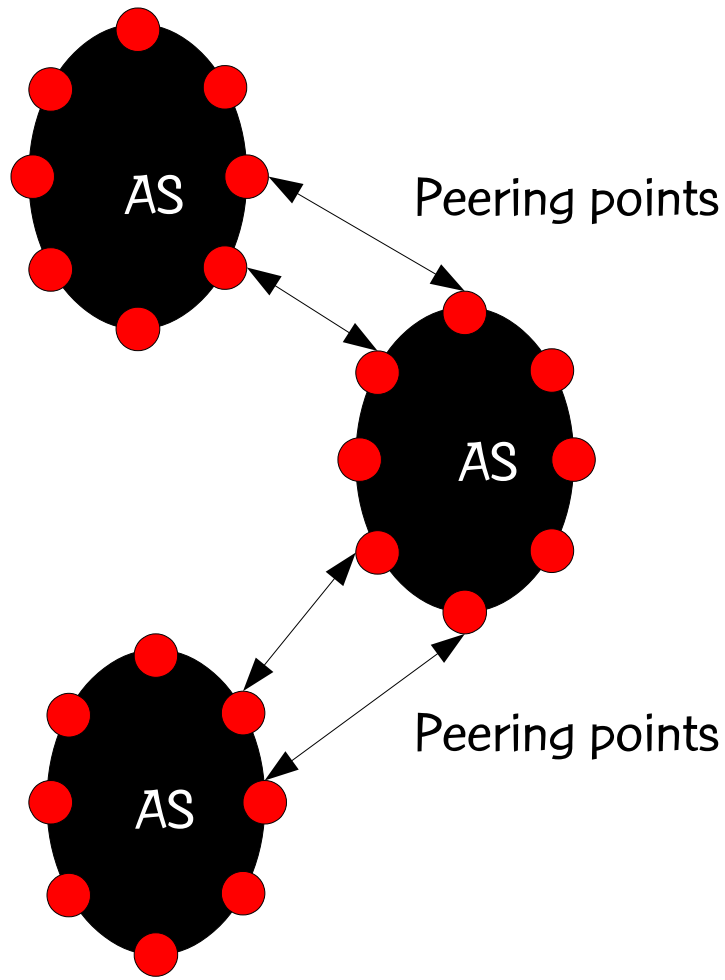


# BGP Routing



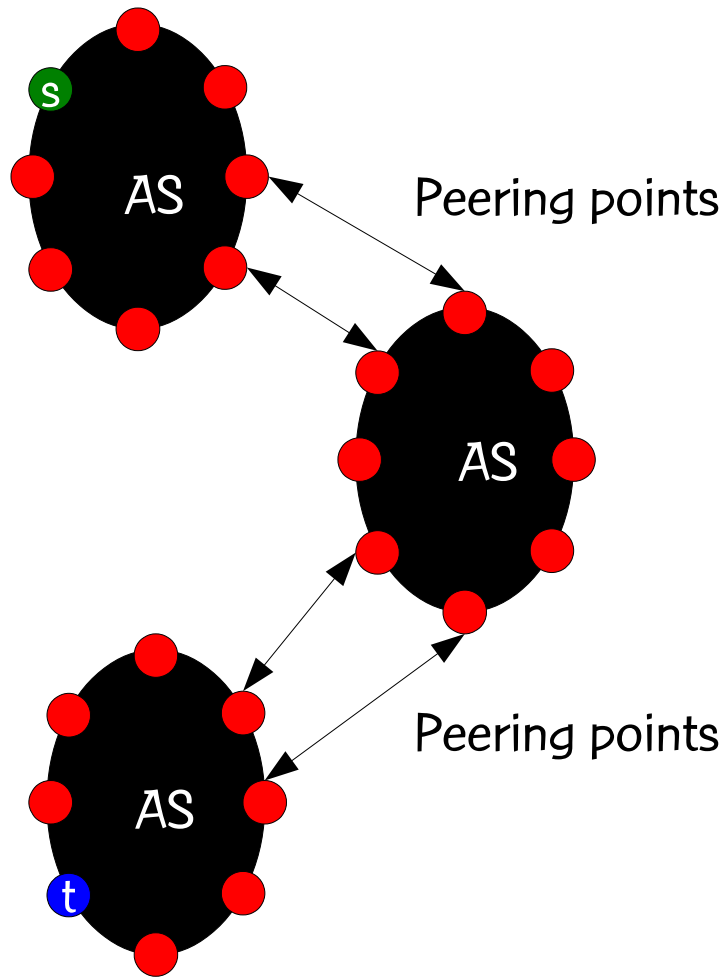
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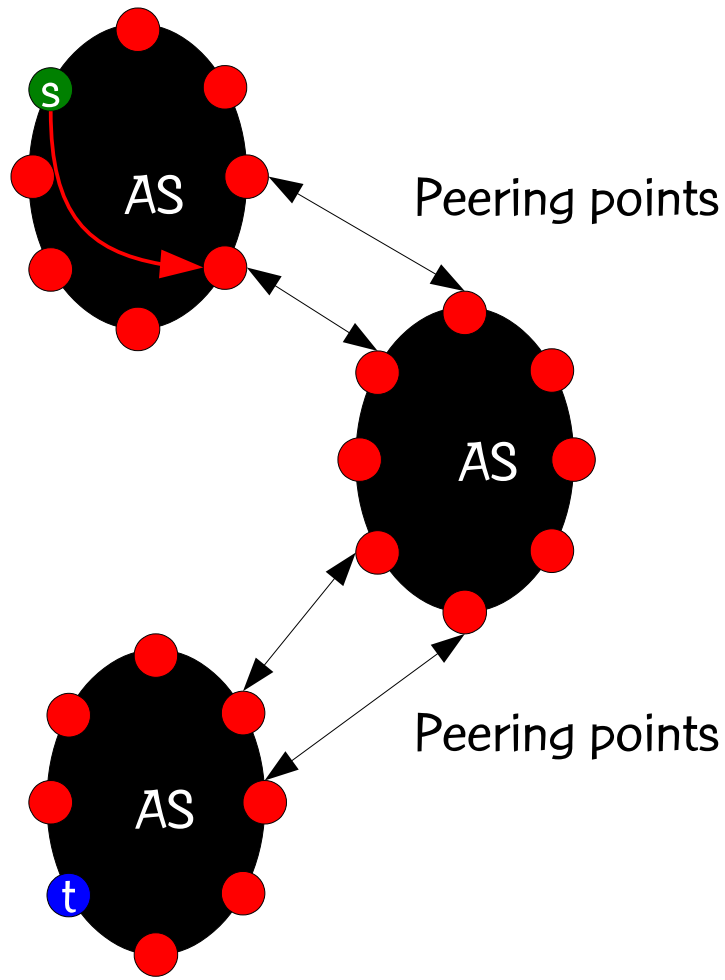
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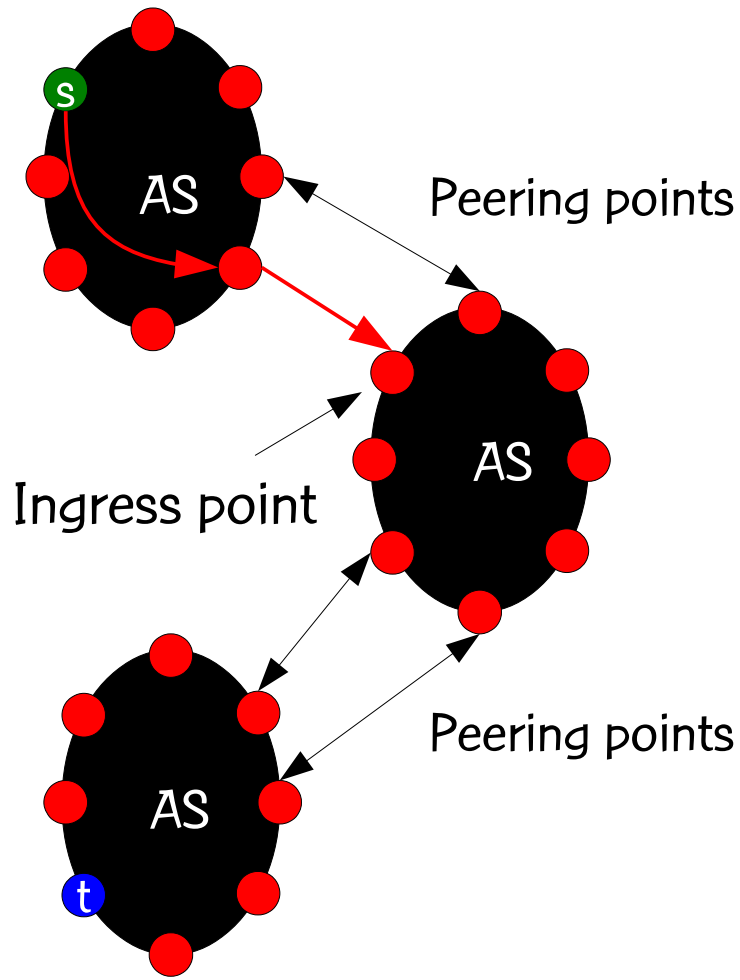
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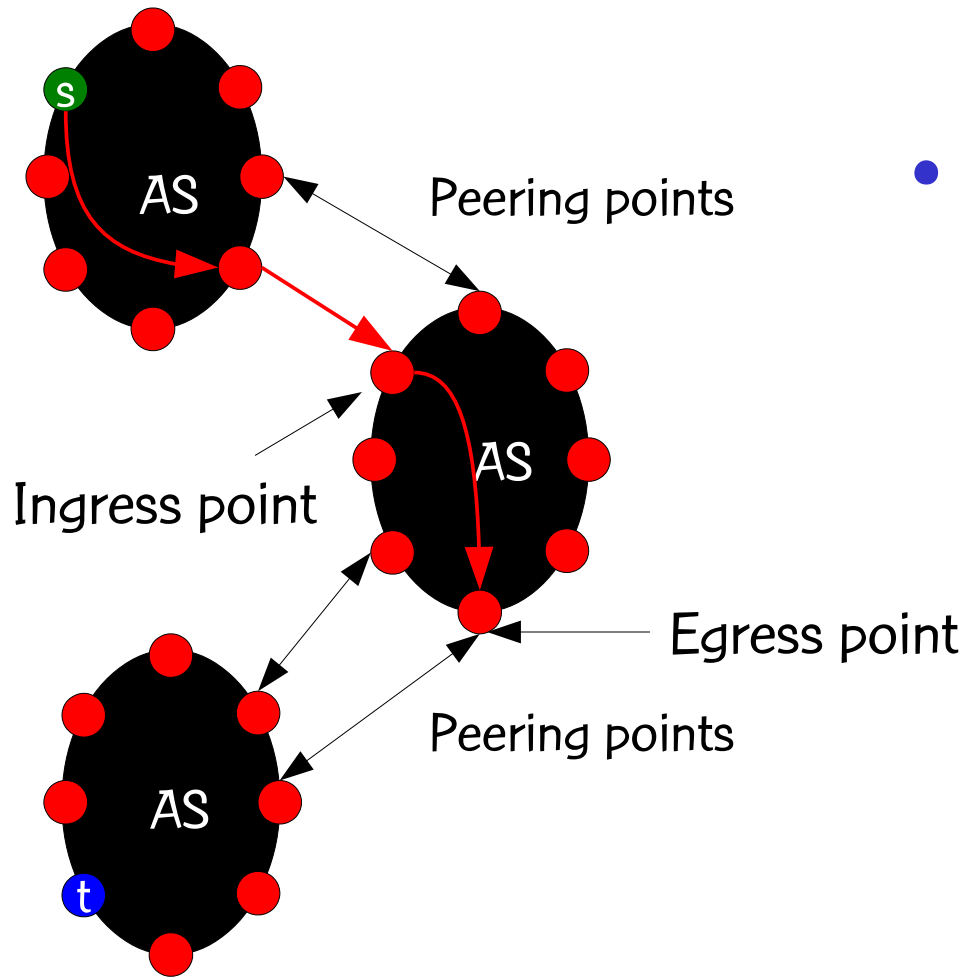
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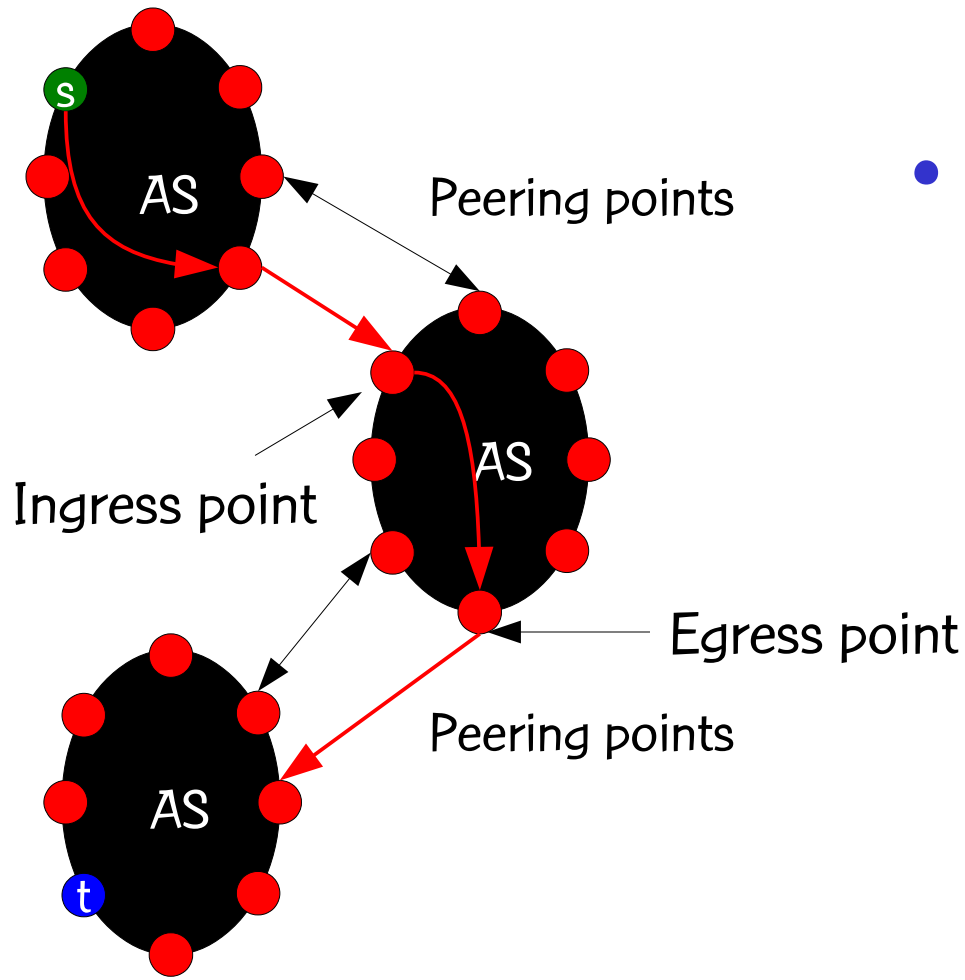
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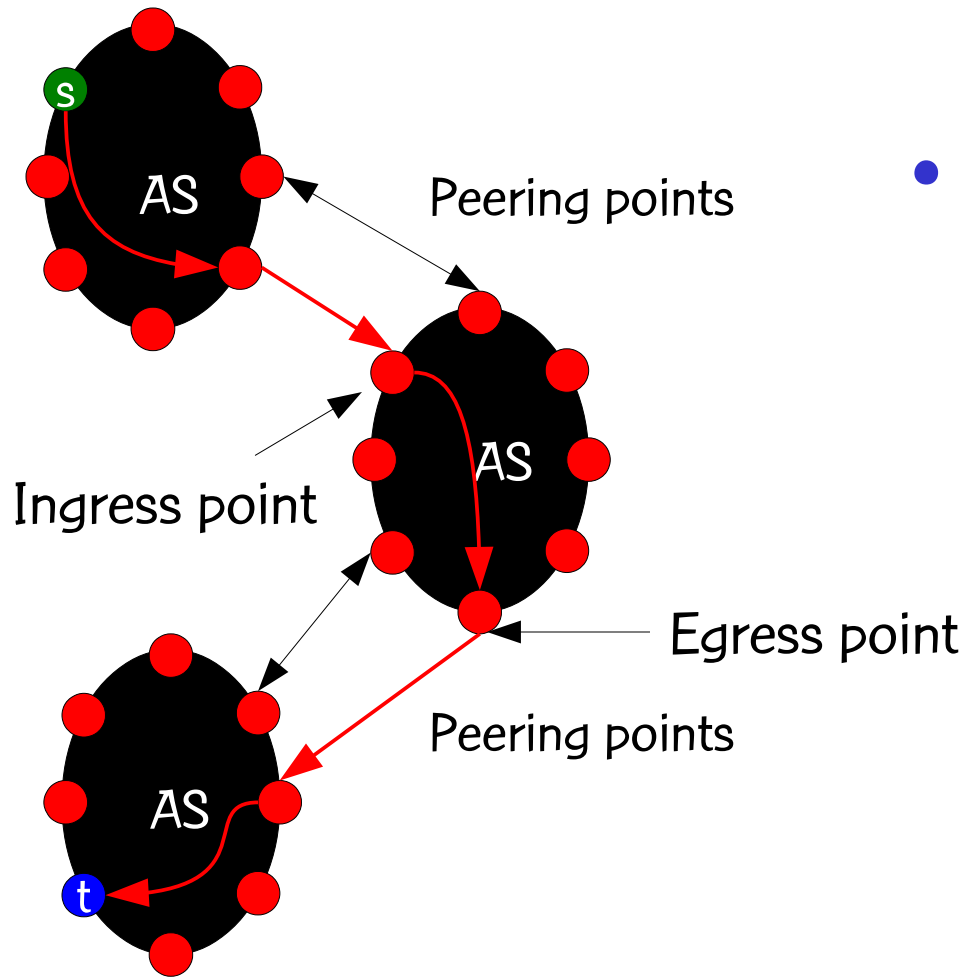


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# IGP Routing

# OSPF routing

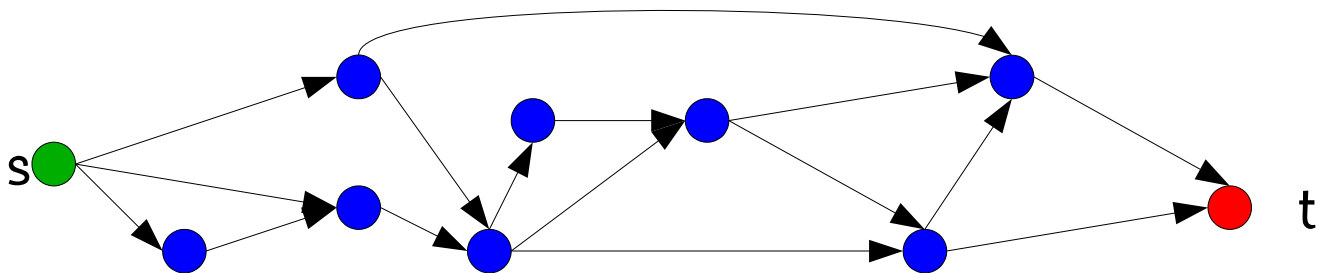
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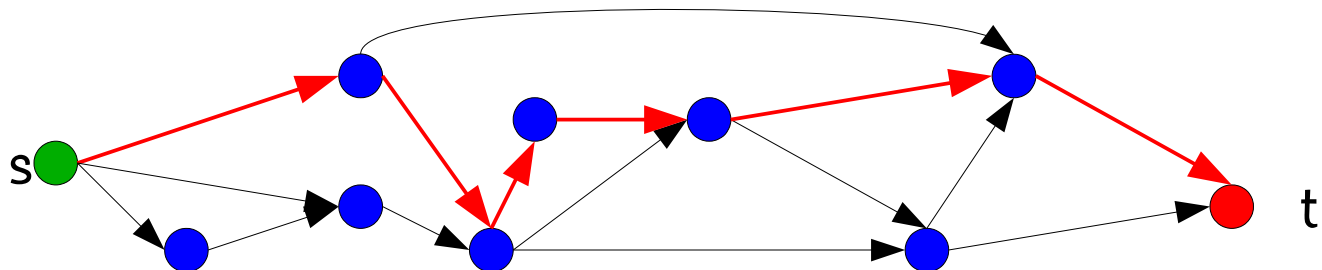
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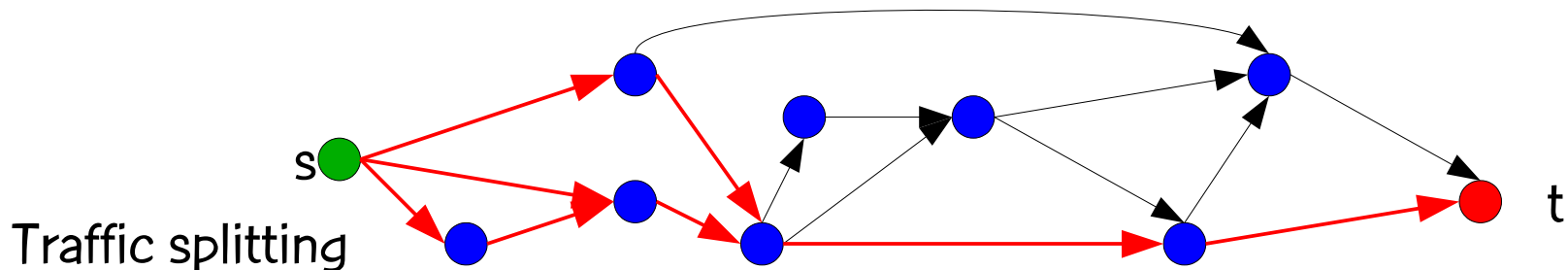
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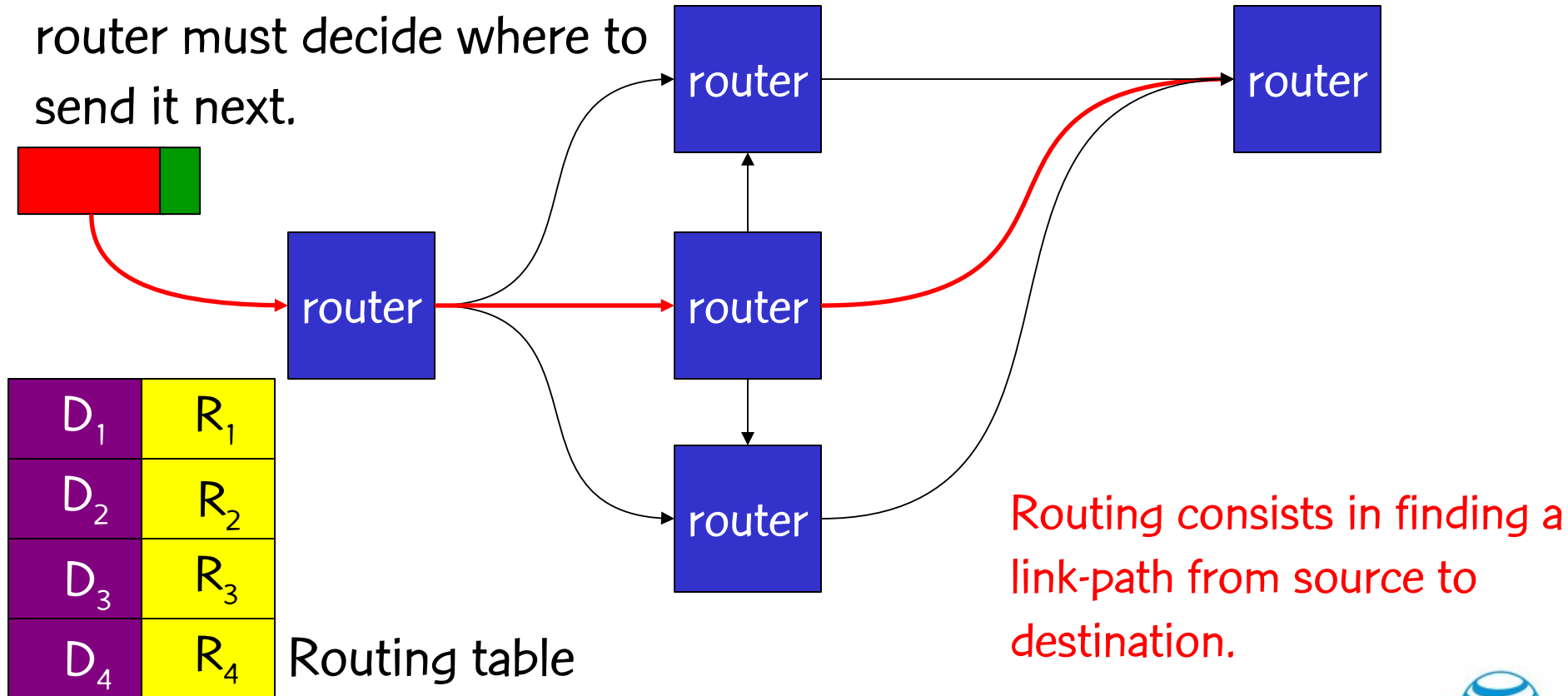
- By setting OSPF weights appropriately, one can do traffic engineering, i.e. route traffic so as to optimize some objective (e.g. minimize congestion, maximize throughput, etc.).
- Some recent papers on this topic:
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# Packet routing

When packet arrives at router, router must decide where to send it next.



# OSPF routing

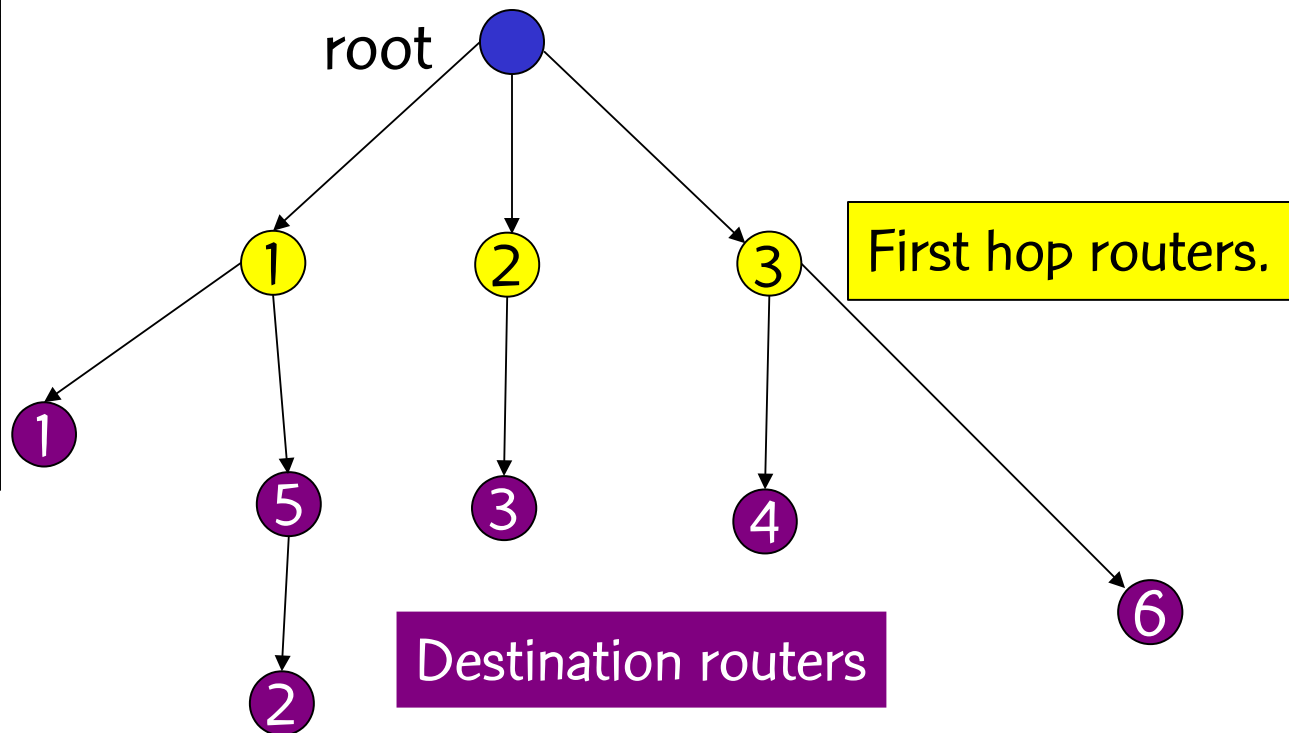
- Assign an integer weight  $\in [1, w_{max}]$  to each link in AS. In general,  $w_{max} = 65535 = 2^{16} - 1$ .
- Each router computes tree of shortest weight paths to all other routers in the AS, with itself as the root, using Dijkstra's algorithm.

# OSPF routing

Routing table

$D_1$	$R_1$
$D_2$	$R_1$
$D_3$	$R_2$
$D_4$	$R_3$
$D_5$	$R_1$
$D_6$	$R_3$

Routing table is filled with first hop routers for each possible destination.

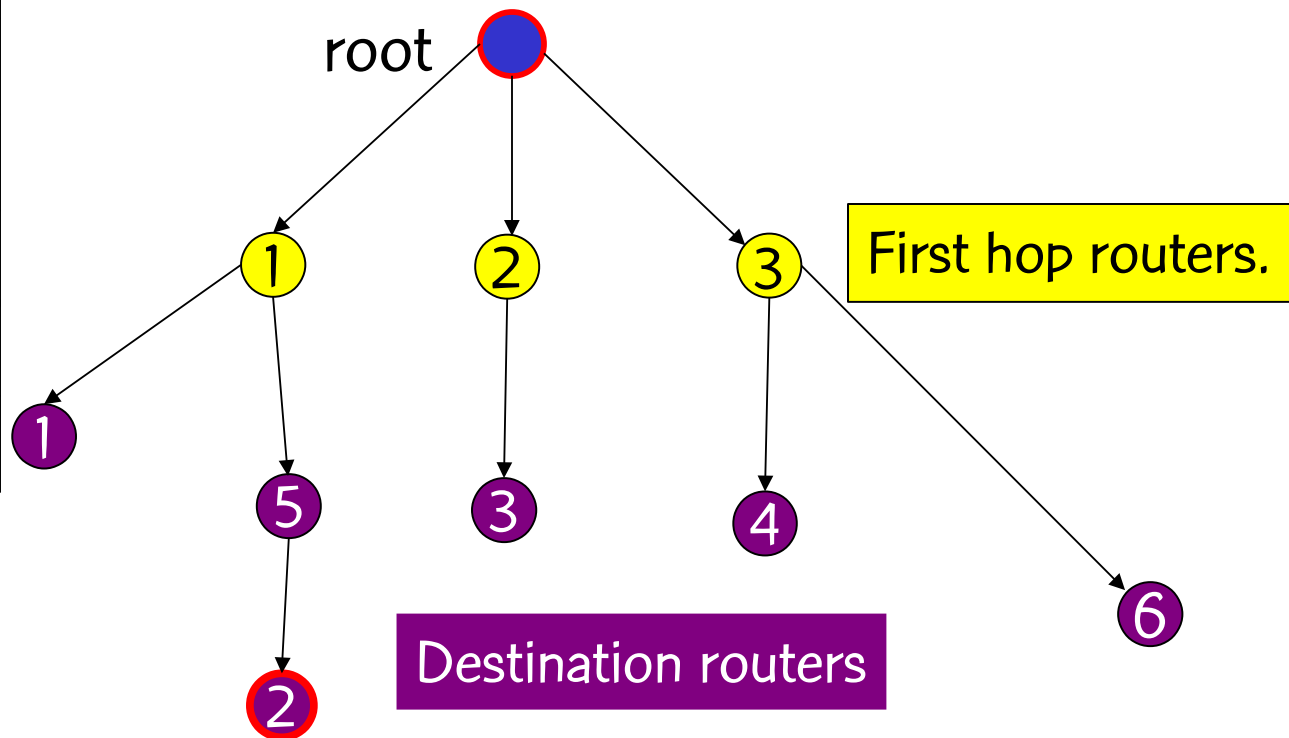


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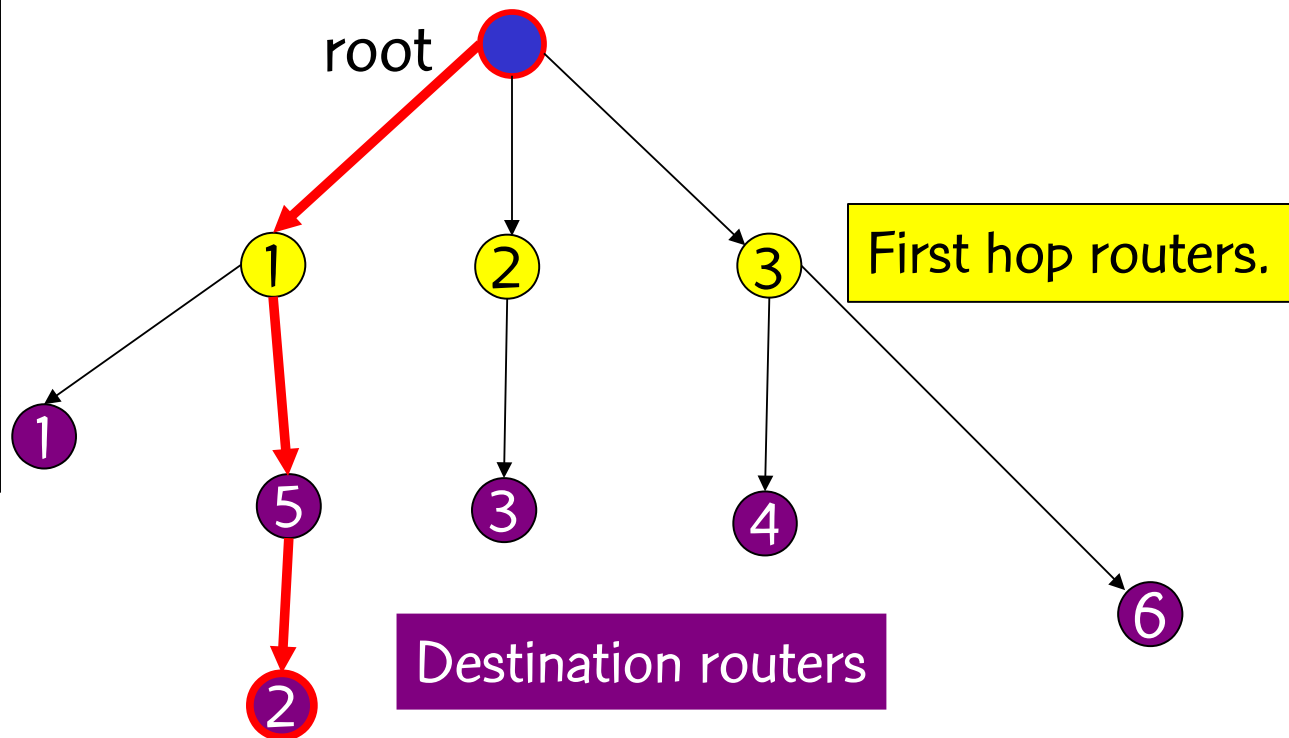


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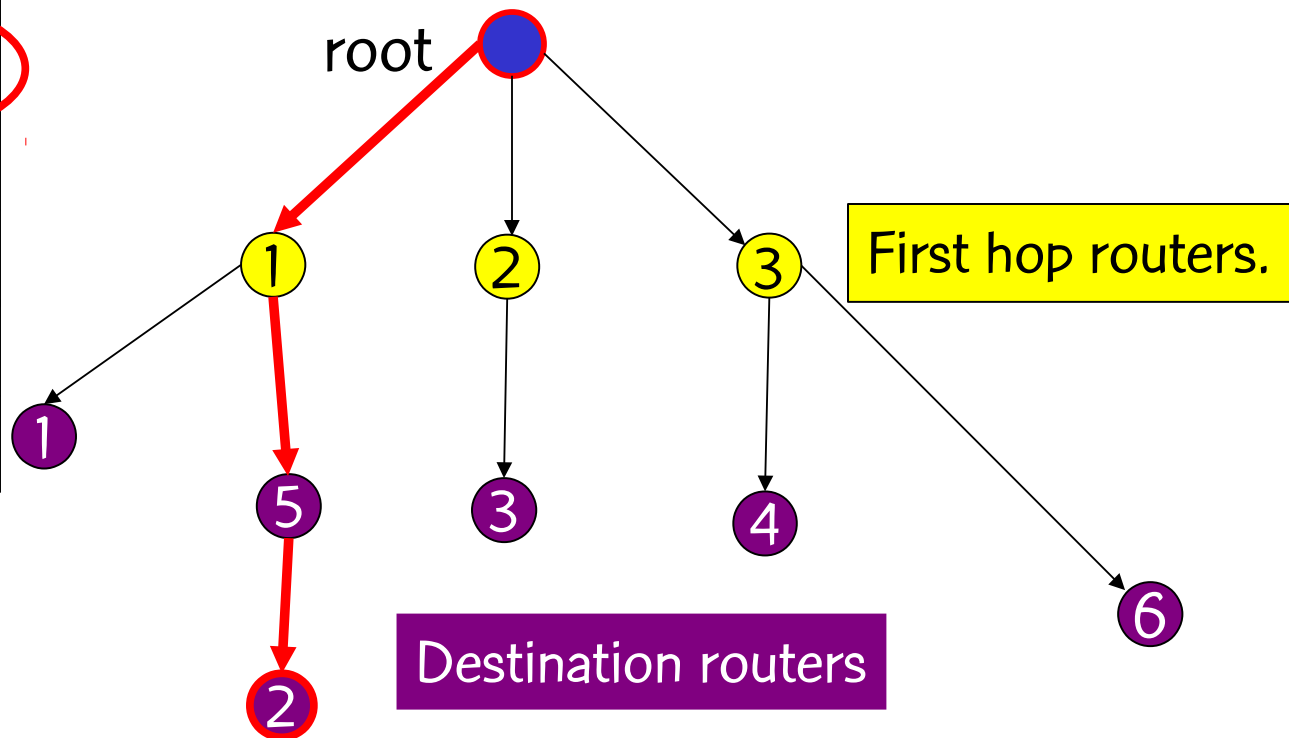


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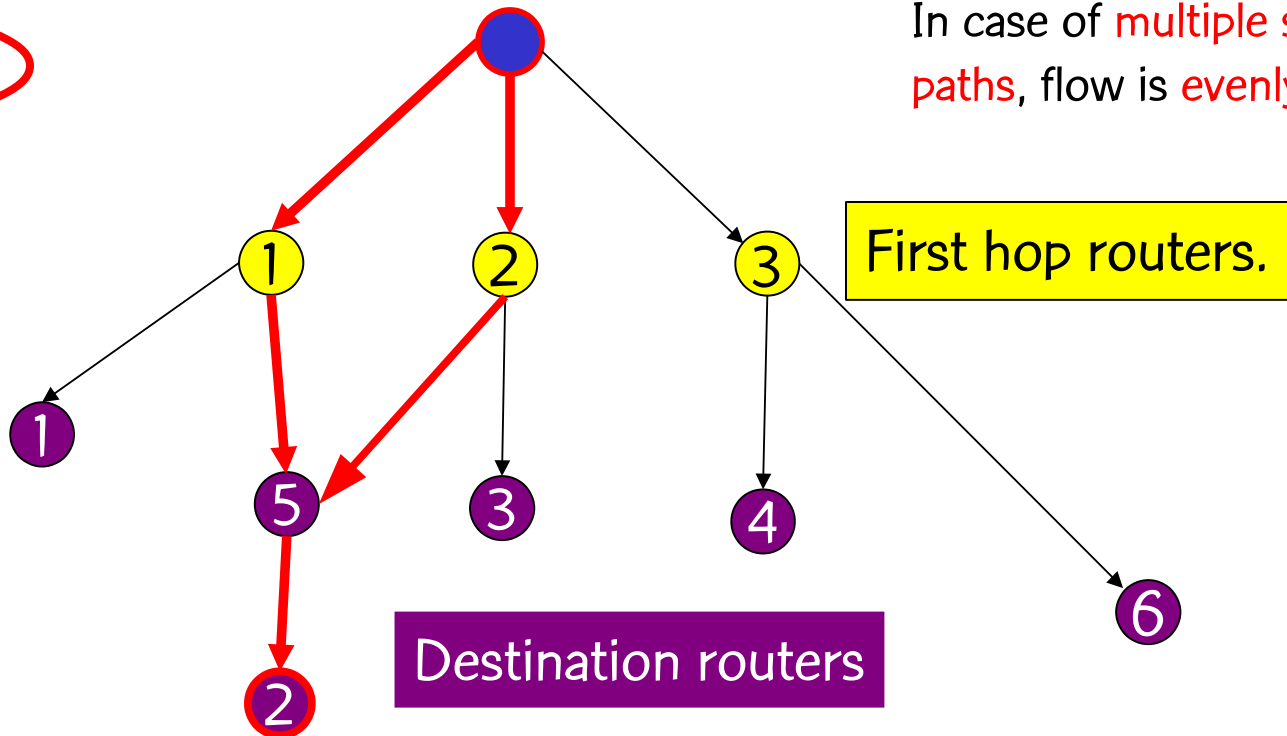




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Routing table is filled with first hop routers for each possible destination. In case of **multiple shortest paths**, flow is **evenly split**.

# OSPF weight setting

- OSPF weights are assigned by network operator.
  - CISCO assigns, by default, a weight proportional to the inverse of the link bandwidth (Inv Cap).
  - If all weights are unit, the weight of a path is the number of hops in the path.
- We propose two BRKGA to find good OSPF weights.

# Minimization of congestion

- Consider the directed capacitated network  $G = (N, A, c)$ , where  $N$  are routers,  $A$  are links, and  $c_a$  is the capacity of link  $a \in A$ .
- We use the measure of Fortz & Thorup (2000) to compute congestion:

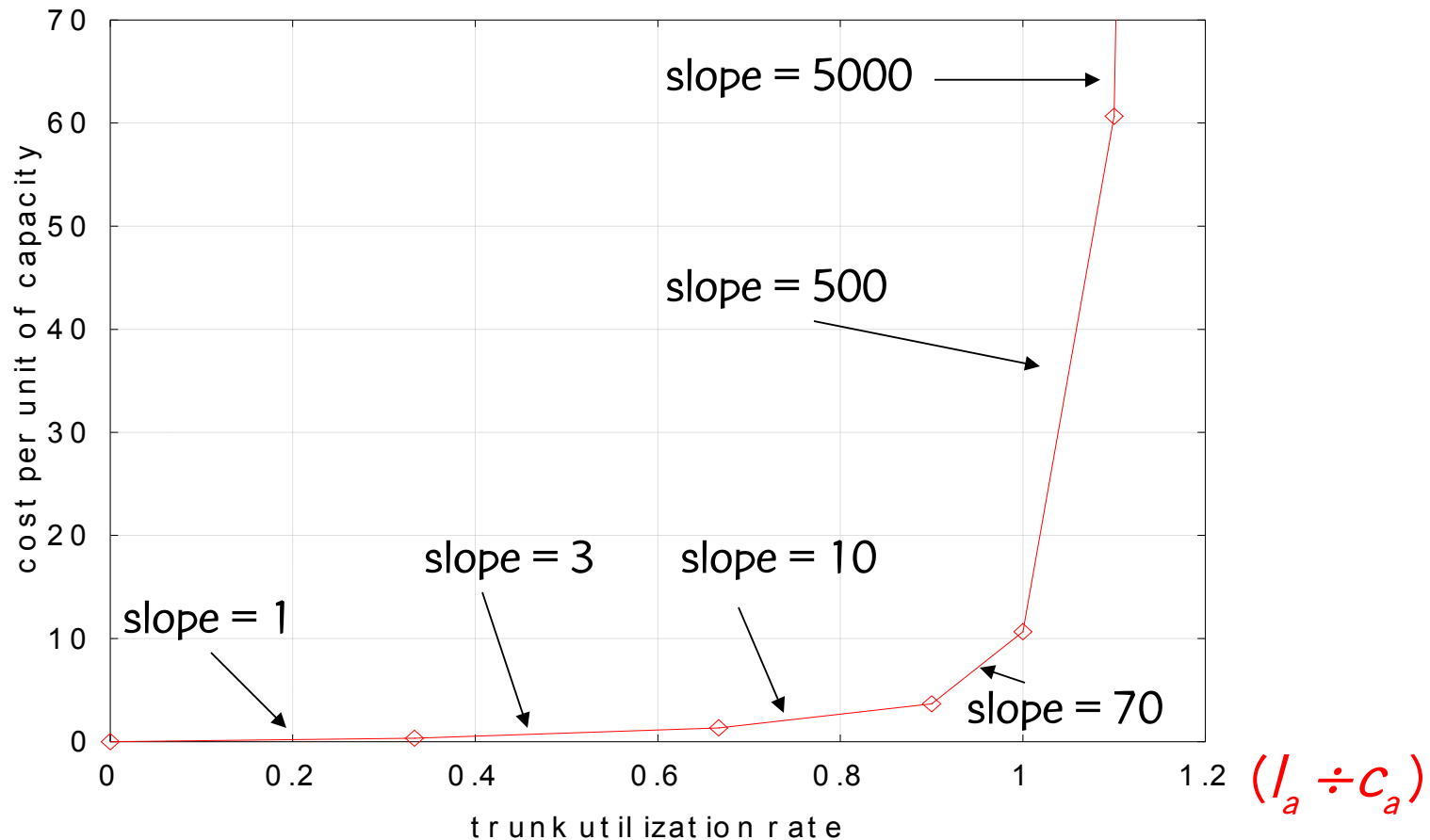
$$\Phi = \Phi_1(I_1) + \Phi_2(I_2) + \dots + \Phi_{|A|}(I_{|A|})$$

where  $I_a$  is the load on link  $a \in A$ ,

$\Phi_a(I_a)$  is piecewise linear and convex,

$\Phi_a(0) = 0$ , for all  $a \in A$ .

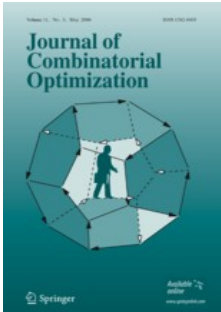
# Piecewise linear and convex $\Phi_a(I_a)$ link congestion measure



# OSPF weight setting problem

- Given a directed network  $G = (N, A)$  with link capacities  $c_a \in A$  and demand matrix  $D = (d_{s,t})$  specifying a demand to be sent from node  $s$  to node  $t$ :
  - Assign weights  $w_a \in [1, w_{max}]$  to each link  $a \in A$ , such that the objective function  $\Phi$  is minimized when demand is routed according to the OSPF protocol.

# BRKGA for OSPF routing in IP networks



M. Ericsson, M.G.C.R., & P.M. Pardalos, “**A genetic algorithm for the weight setting problem in OSPF routing**,” J. of Combinatorial Optimization, vol. 6, pp. 299–333, 2002.

Tech report version:

<http://www2.research.att.com/~mgcr/doc/gaospf.pdf>

# BRKGA for OSPF routing in IP networks

Ericsson, R., & Pardalos (J. Comb. Opt., 2002)

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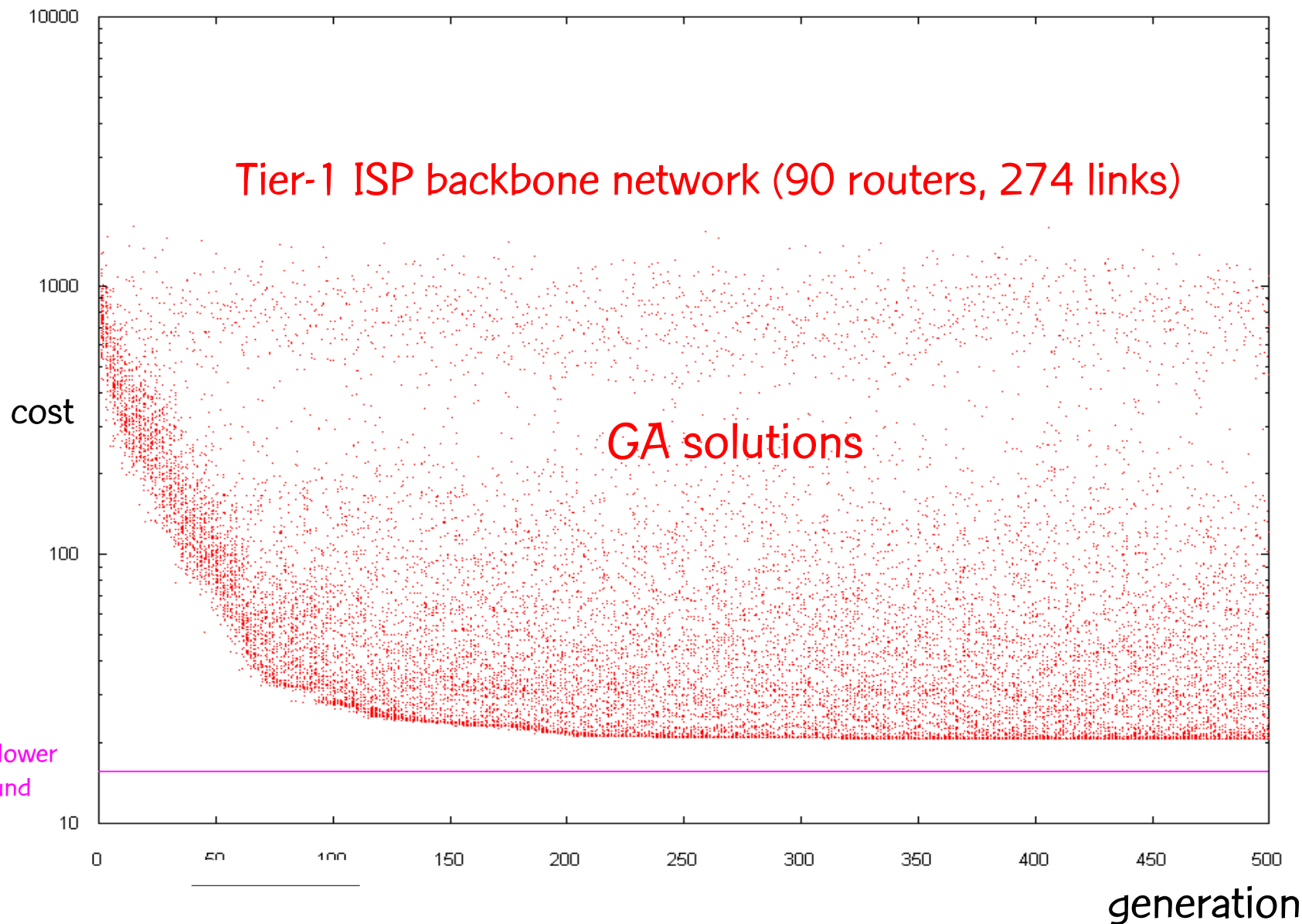
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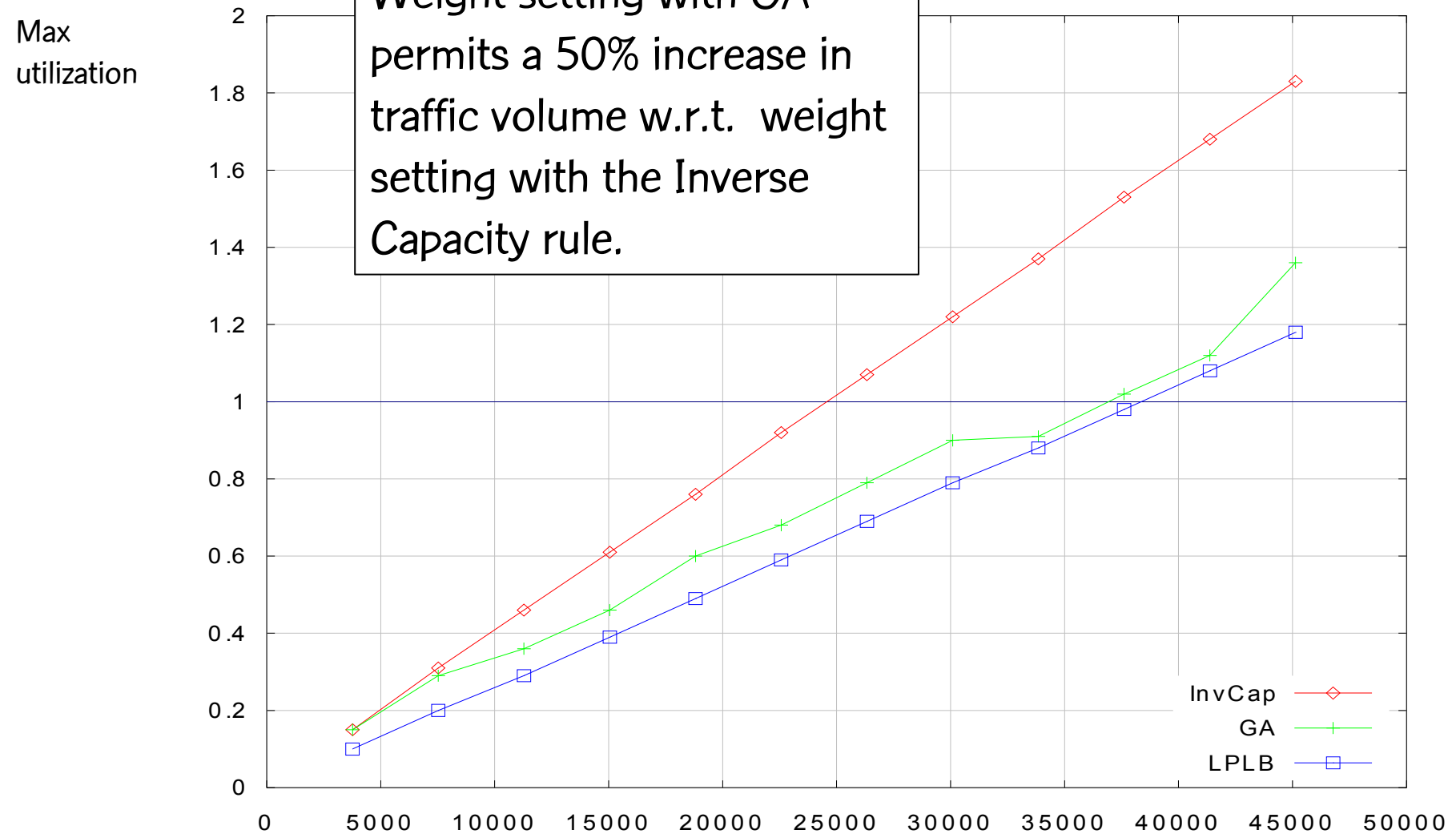


## Tier-1 ISP backbone network (90 routers, 274 links)

Weight setting with GA permits a 50% increase in traffic volume w.r.t. weight setting with the Inverse Capacity rule.

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# Improved BRKGA for OSPF routing in IP networks



L.S. Buriol, M.G.C.R., C.C. Ribeiro, and M. Thorup, “**A hybrid genetic algorithm for the weight setting problem in OSPF/IS-IS routing**,” Networks, vol. 46, pp. 36–56, 2005.

Tech report version:

<http://www2.research.att.com/~mgcr/doc/hgaospf.pdf>

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Buriol, R., Ribeiro, and Thorup (Networks, 2005)

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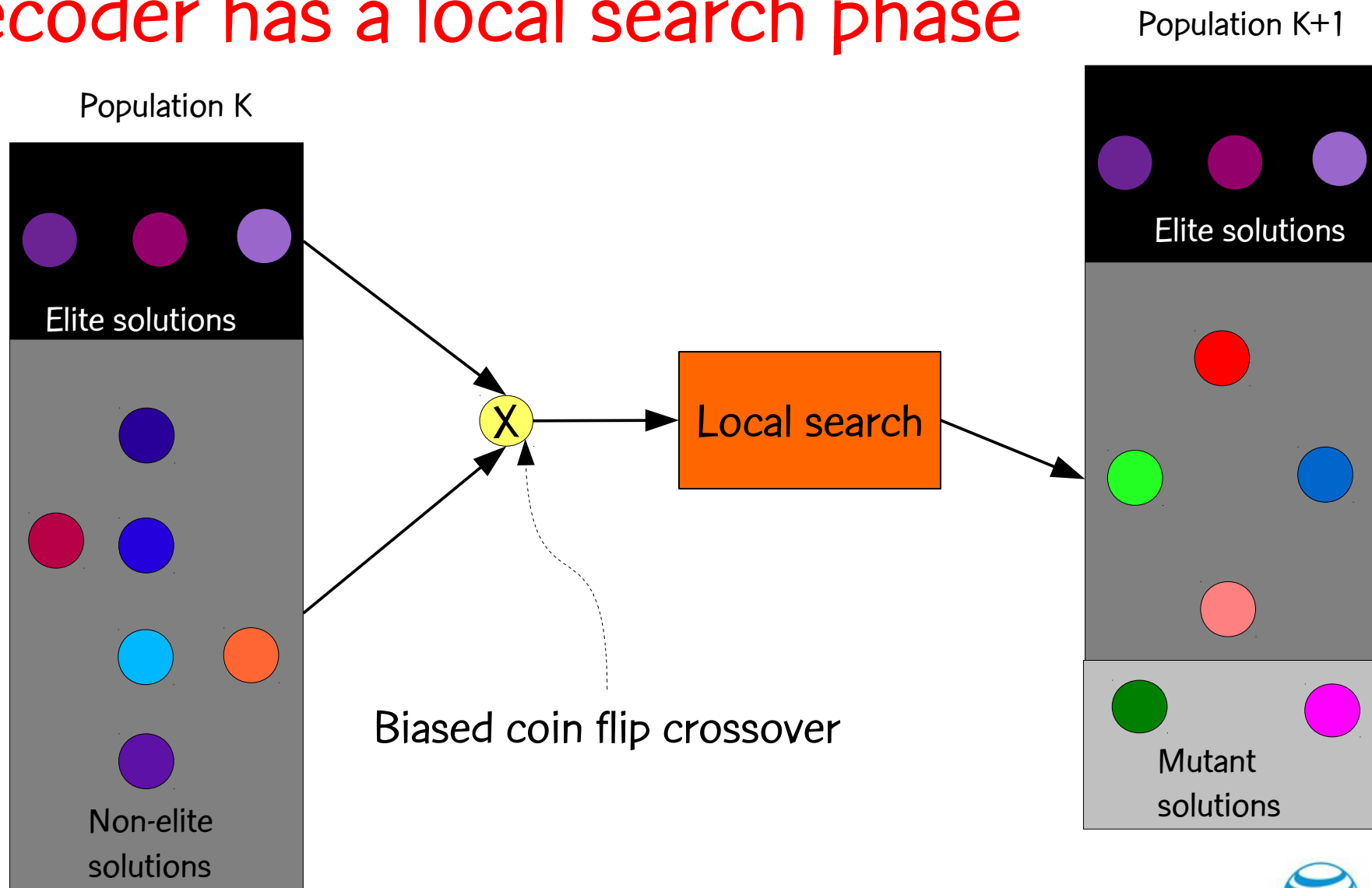
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  - Apply fast local search to improve weights.

# Decoder has a local search phase



# Fast local search

- Let  $A^*$  be the set of five arcs  $a \in A$  having largest  $\Phi_a$  values.

# Fast local search

- Let  $A^*$  be the set of five arcs  $a \in A$  having largest  $\Phi_a$  values.
- Scan arcs  $a \in A^*$  from largest to smallest  $\Phi_a$ :

# Fast local search

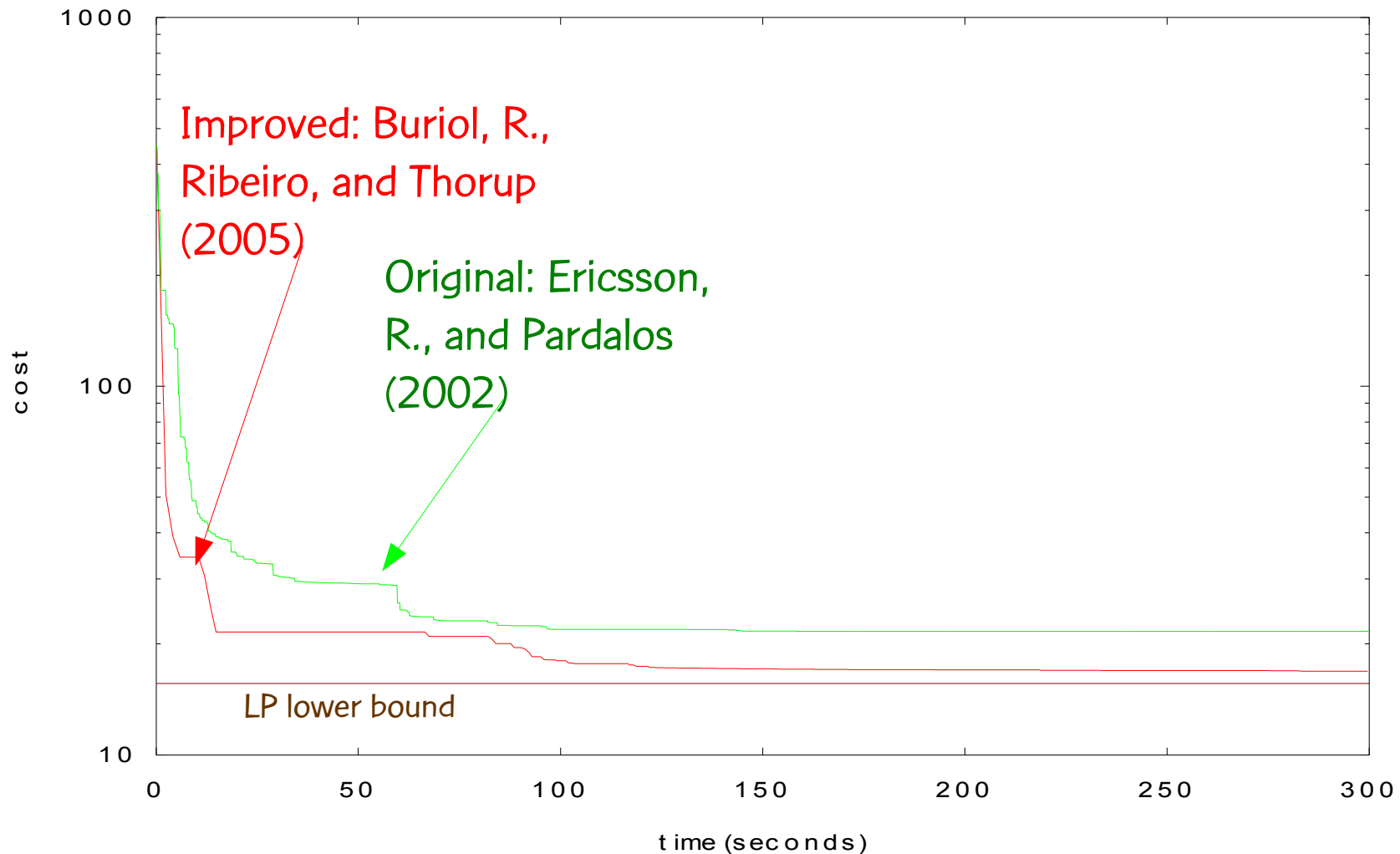
- Let  $A^*$  be the set of five arcs  $a \in A$  having largest  $\Phi_a$  values.
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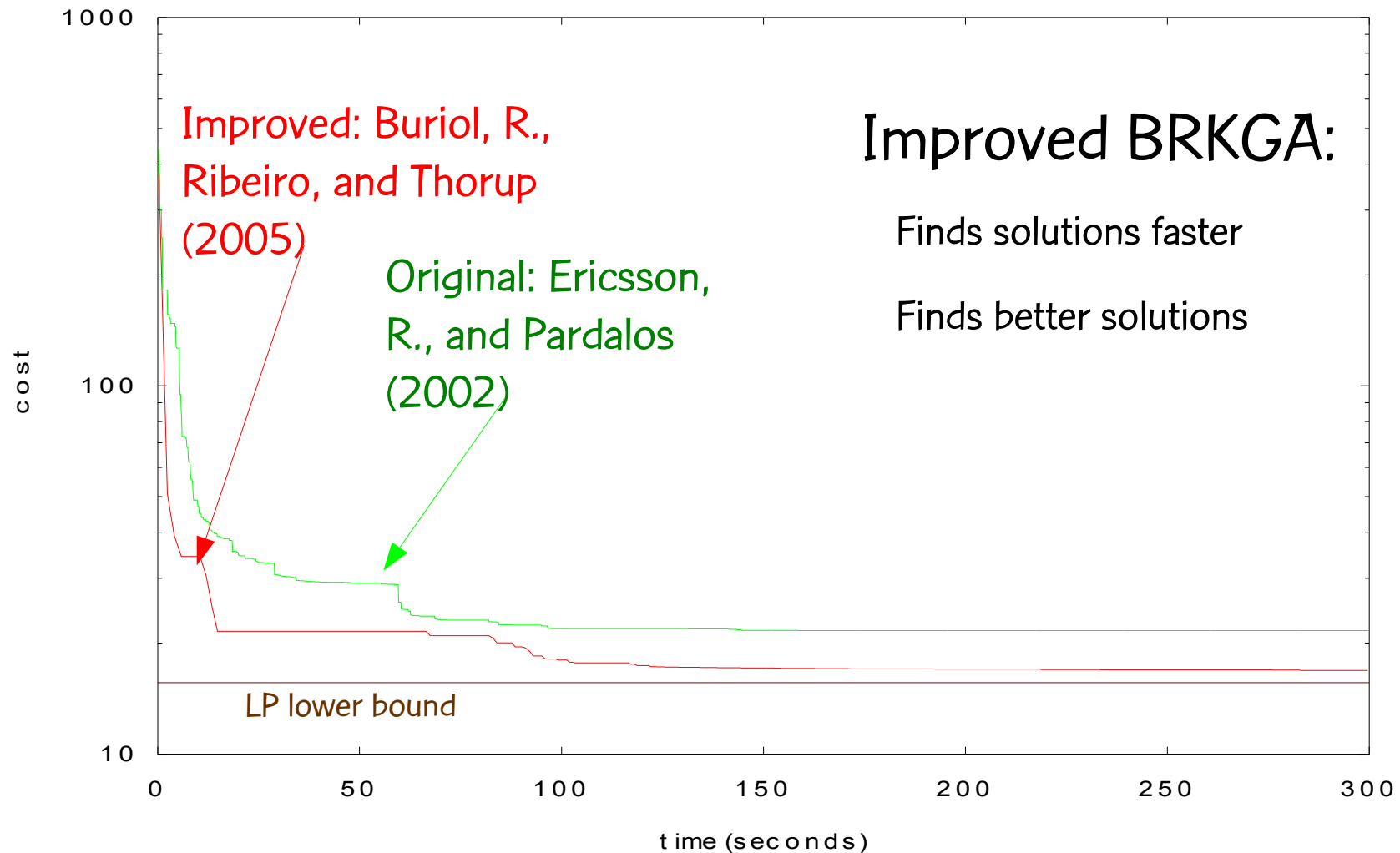
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  - If total cost  $\Phi$  is reduced, restart local search.

# Effect of decoder with fast local search





# Effect of decoder with fast local search



# Survivable IP network design

# Survivable IP network design



L.S. Buriol, M.G.C.R., and M. Thorup, “Survivable IP network design with OSPF routing,” *Networks*, vol. 49, pp. 51–64, 2007.

Tech report version:

<http://www2.research.att.com/~mgcr/doc/gamult.pdf>

# Survivable IP network design

Buriol, R., & Thorup (Networks, 2007)

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- Min total **design cost**  $= \sum_{a \in A} M(a) \times K(a)$ .

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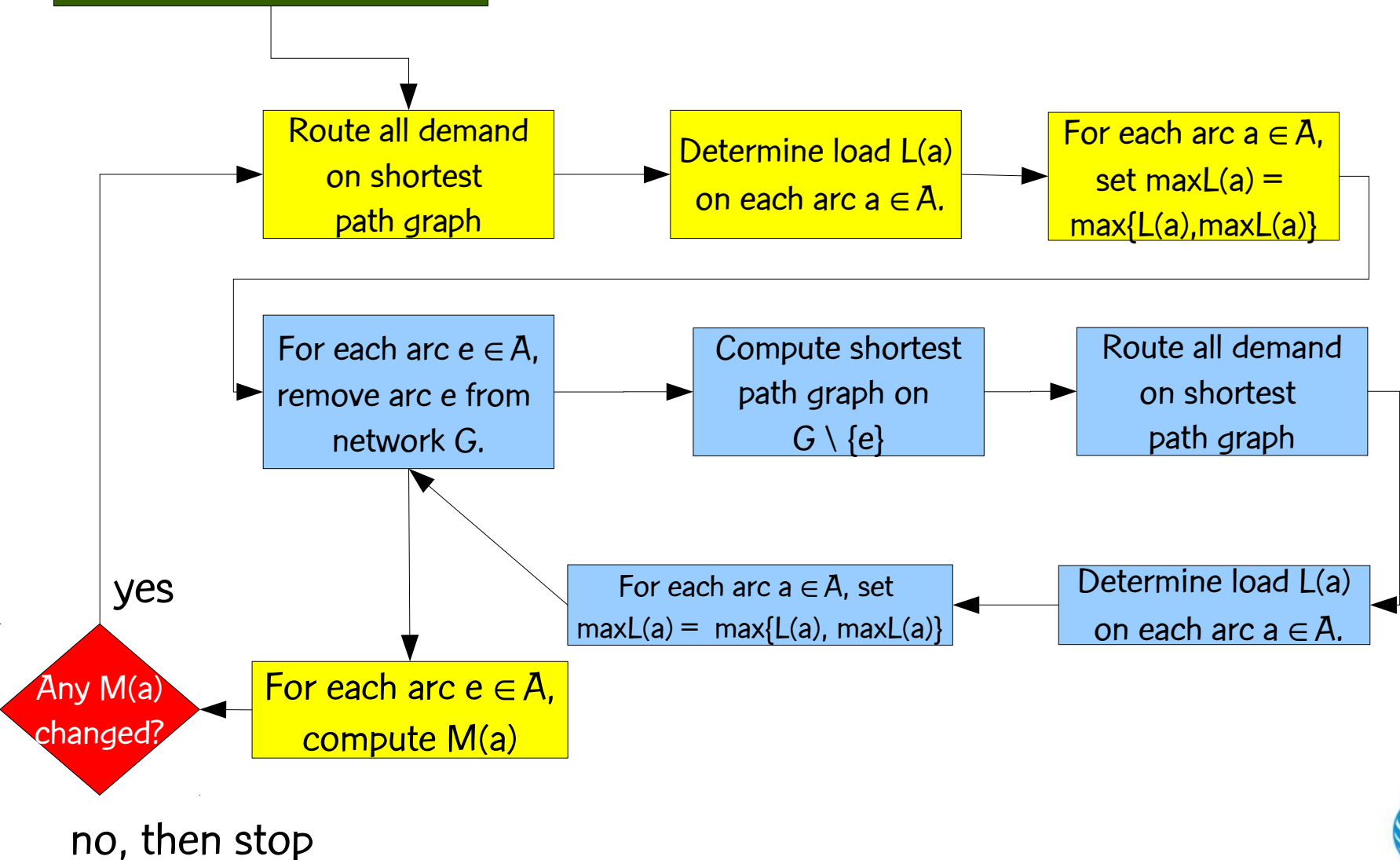
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- For each arc  $a \in A$ , determine the multiplicity  $M(a)$  using the maximum load for that arc over all failure modes.
- Network design  $\text{cost} = \sum_{a \in A} M(a) \times K(a)$

## Computing the “fitness” of a solution (single link failure case)

For each arc  $a \in \bar{A}$ , set  
 $M(a) = 1$ ;  $\max L(a) = -\infty$



# Composite-link design

- In Buriol, R., and Thorup (2006)
  - links were all of the same type,
  - only the link multiplicity had to be determined.
- Now consider composite links. Given a load  $L(a)$  on arc  $a$ , we can compose several different link types that sum up to the needed capacity  $c(a) \geq L(a)$ :
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# Composite-link design

D.V. Andrade, L.S. Buriol, M.G.C.R., and M. Thorup,  
“Survivable composite-link IP network design with OSPF  
routing,” The Eighth INFORMS Telecommunications  
Conference, Dallas, Texas, April 2006.

Tech report:

<http://www2.research.att.com/~mgcr/doc/composite.pdf>

# Composite-link design

- Link types =  $\{ 1, 2, \dots, T \}$
- Capacities =  $\{ c(1), c(2), \dots, c(T) \} : c(i) < c(i+1)$
- Prices / unit length =  $\{ p(1), p(2), \dots, p(T) \} : p(i) < p(i+1)$
- Assumptions:
  - $[p(T)/c(T)] < [p(T-1)/c(T-1)] < \dots < [p(1)/c(1)]$ , i.e. price per unit of capacity is smaller for links with greater capacity
  - $c(i) = \alpha \times c(i-1)$ , for  $\alpha \in \mathbb{N}$ ,  $\alpha > 1$ , i.e. capacities are multiples of each other by powers of  $\alpha$



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  - $c(i) = \alpha \times c(i-1)$ , for  $\alpha \in \mathbb{N}$ ,  $\alpha > 1$ , e.g.  
 $c(\text{OC192}) = 4 \times c(\text{OC48})$ ;  $c(\text{OC48}) = 4 \times c(\text{OC12})$ ;  
 $c(\text{OC12}) = 4 \times c(\text{OC3})$ ;

OC3	OC12	OC48	OC192	
155 Mb/s	622 Mb/s	2.5 Gb/s	10 Gb/s	$\alpha = 4$

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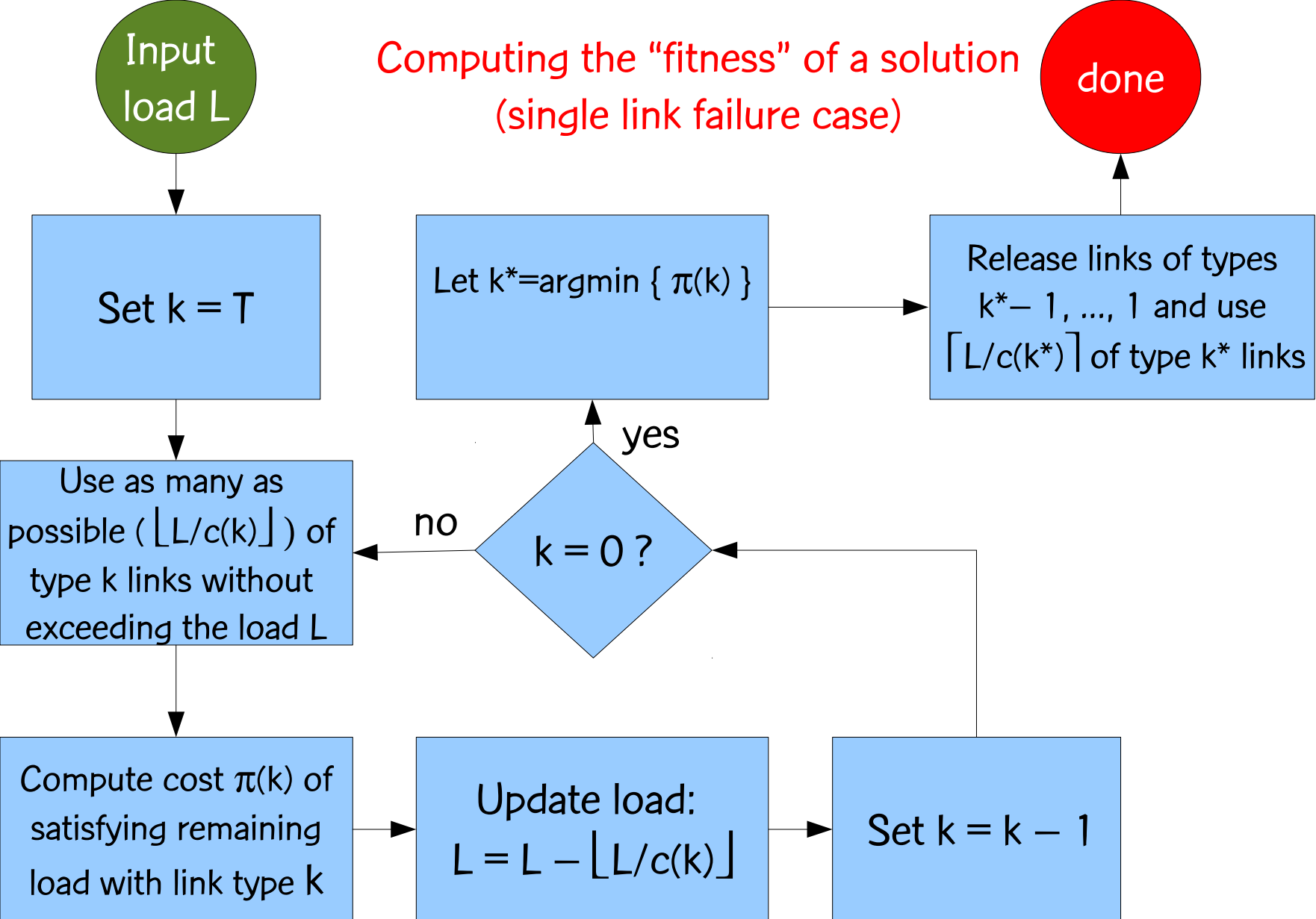
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- Network design cost =  $\sum_{i \in A} \sum_{t \text{ used in arc } i} M(t,i) \times p(t)$



Computing the “fitness” of a solution  
(single link failure case)



# Redundant content distribution

# Reference:

## ALENEX11

Workshop on  
Algorithm Engineering & Experiments

January 22, 2011

Holiday Inn San Francisco Golden Gateway  
San Francisco, California USA

L. Breslau, I. Diakonikolas, N. Duffield,  
Y. Gu, M. Hajiaghayi, D.S. Johnson,  
H. Karloff, M.G.C.R., and S. Sen,  
“Disjoint-path facility location: Theory and  
practice,” Proceedings of the Thirteenth  
Workshop on Algorithm Engineering and  
Experiments (ALENEX11), SIAM, San  
Francisco, pp. 60–74, January 22,  
2011

Tech report version:

<http://www2.research.att.com/~mgcr/doc/monitoring-alenex.pdf>

# Redundant content distribution (RCD)

- Suppose a number of users located at nodes in a network demand content.

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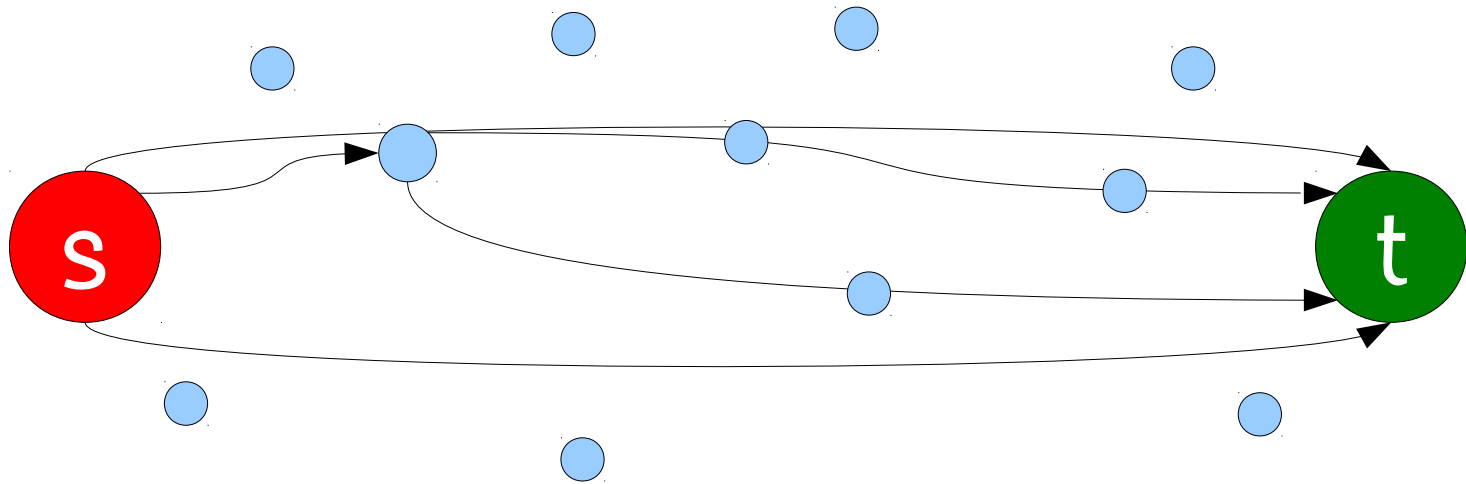
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- Problem: Locate minimum number of warehouses in network such all users get their content even in presence of edge failures.

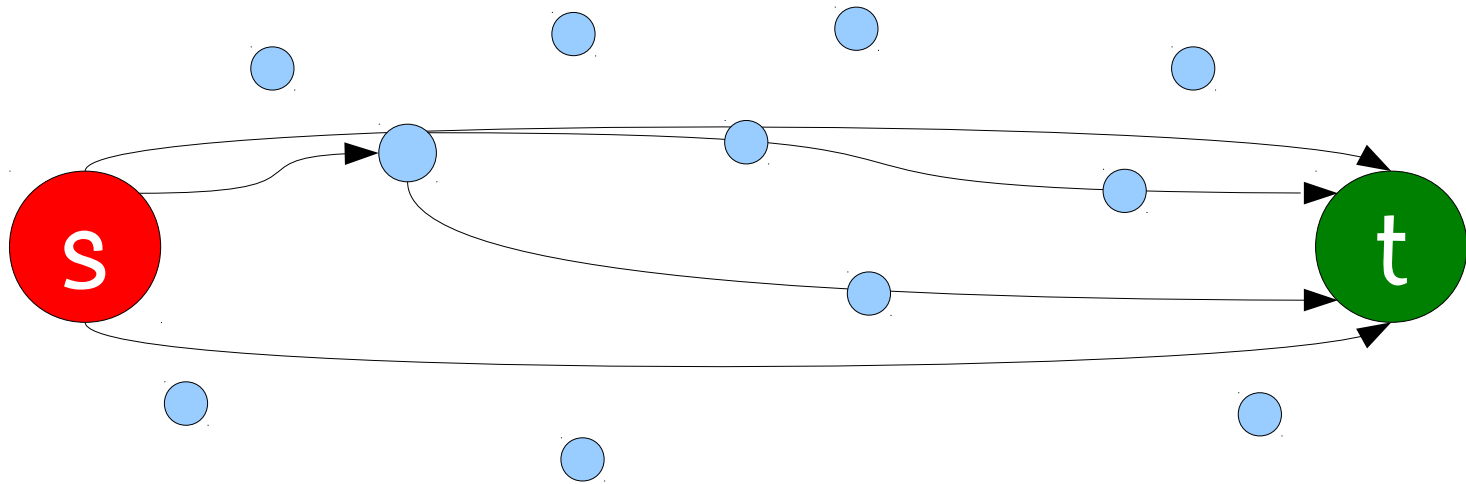
# Redundant content distribution



Traffic from node **s** to node **t** flows on paths defined by OSPF.

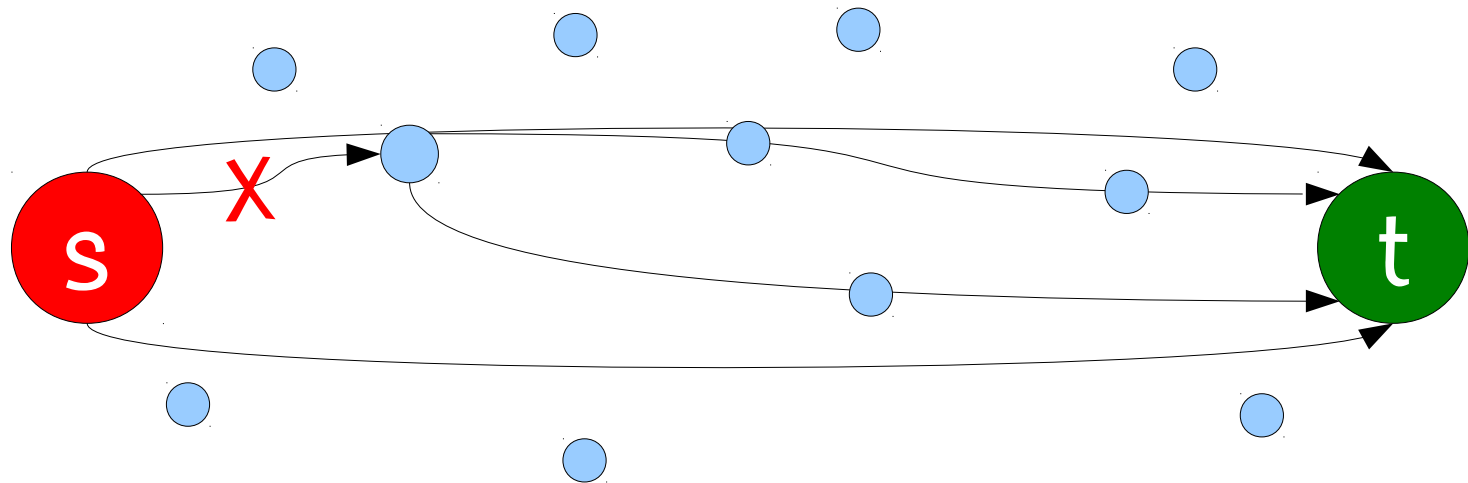


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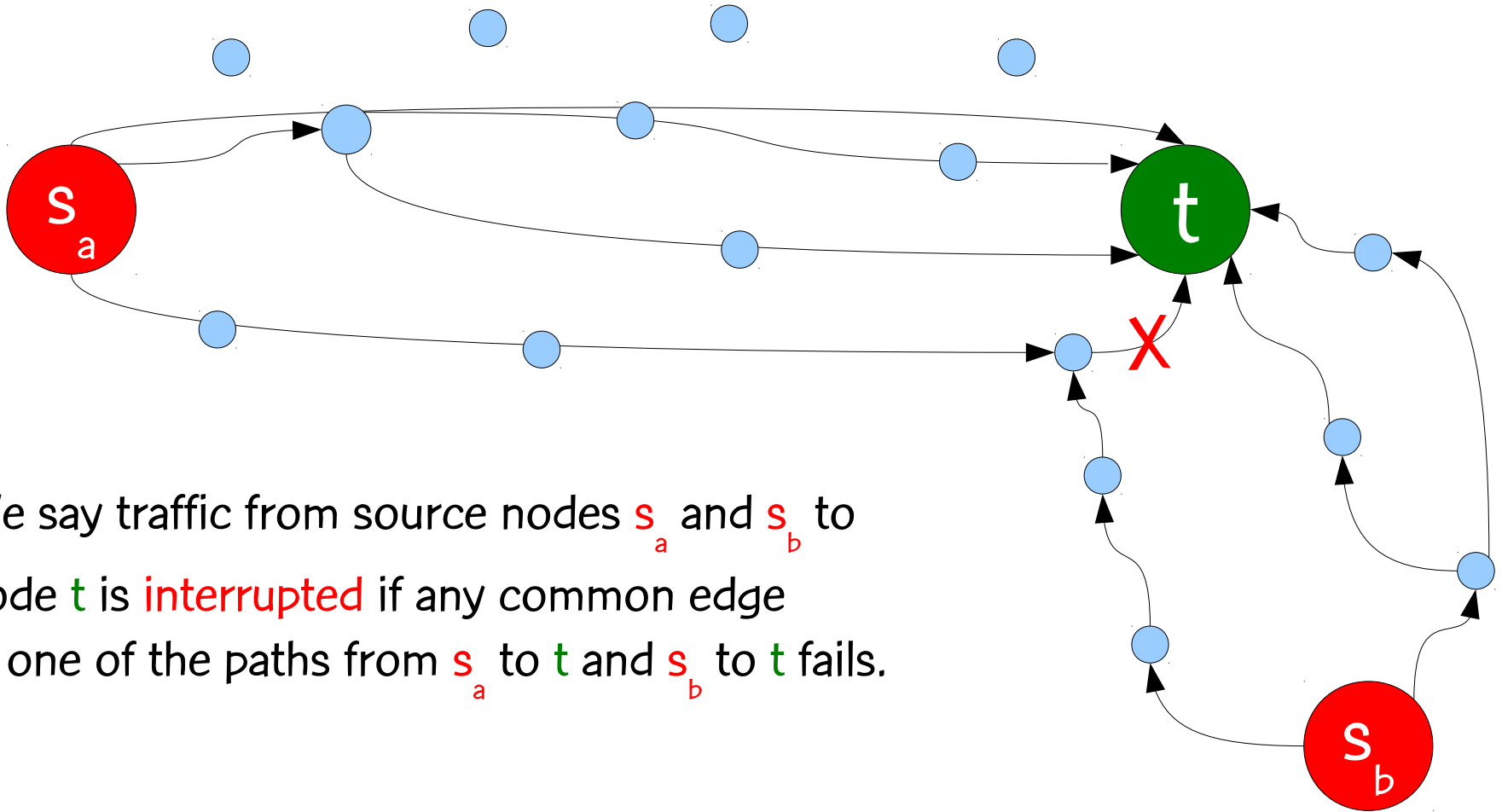


We don't know on which path a particular packet will flow.

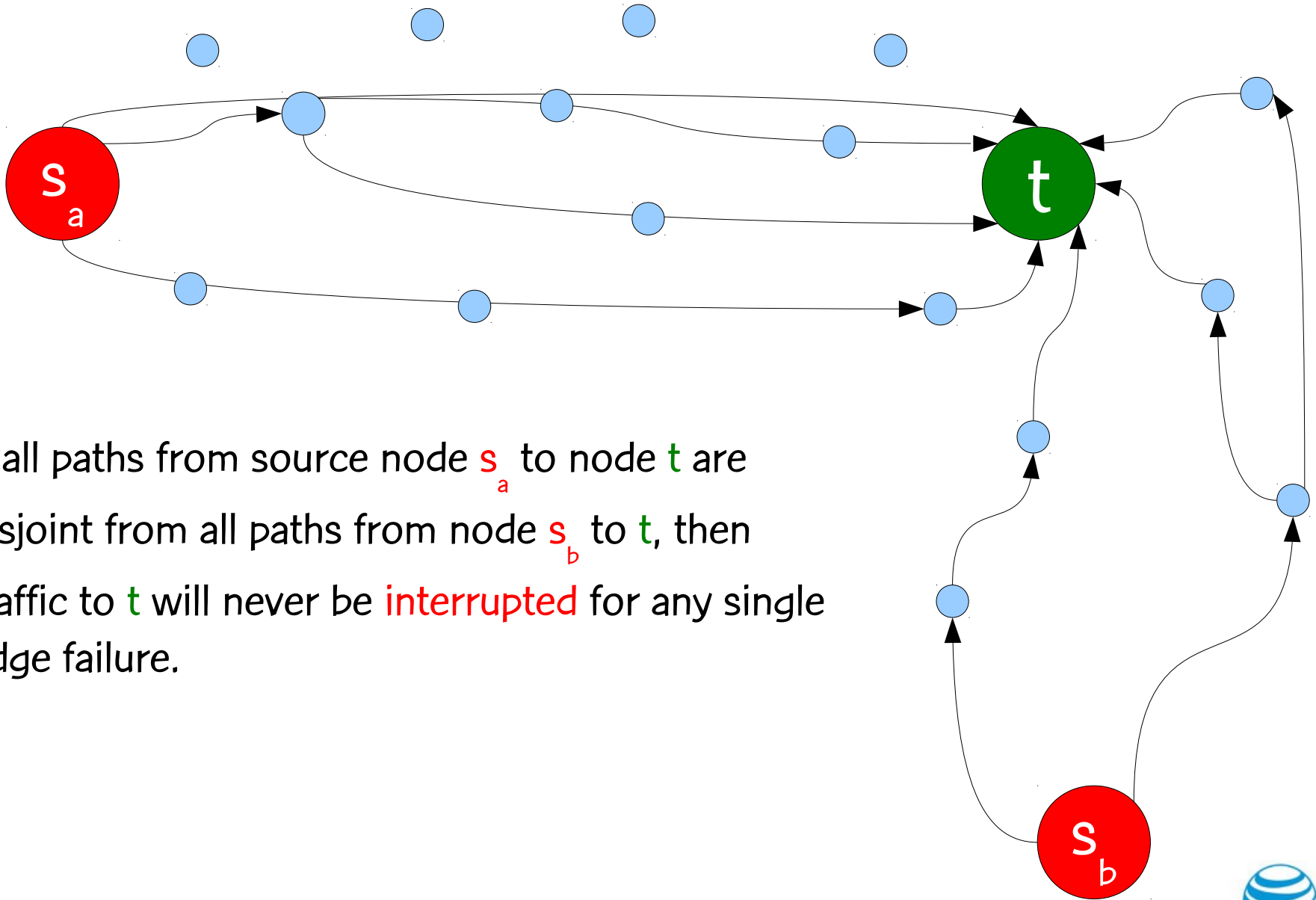
# Redundant content distribution



We say traffic from node **s** to node **t** is **interrupted** if any edge in one of the paths from **s** to **t** fails.



We say traffic from source nodes  $s_a$  and  $s_b$  to node  $t$  is **interrupted** if any common edge in one of the paths from  $s_a$  to  $t$  and  $s_b$  to  $t$  fails.



If all paths from source node  $s_a$  to node  $t$  are disjoint from all paths from node  $s_b$  to  $t$ , then traffic to  $t$  will never be interrupted for any single edge failure.

# Redundant content distribution

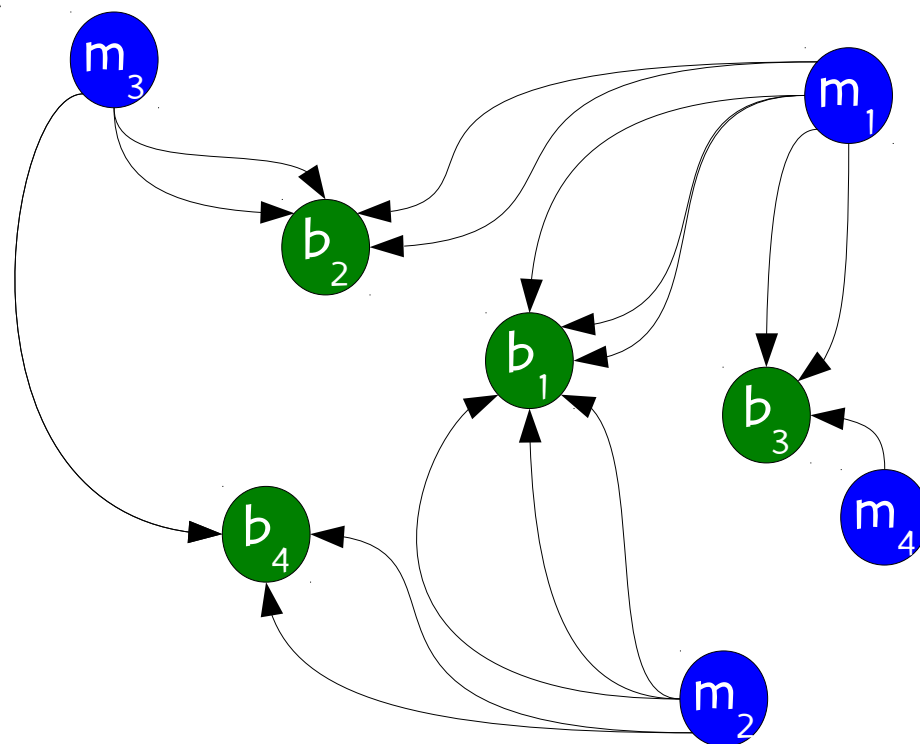
Suppose nodes  $b_1, b_2, \dots$  want some content (e.g. video).

We want the smallest set  $\mathbf{S}$  of servers such that:

for every  $b_i$  there exist  $m_1, m_2 \in \mathbf{S}$  both of which can provide content to  $b_i$

and all paths  $m_1 \rightarrow b$  are disjoint

with all paths  $m_2 \rightarrow b$



# Redundant content distribution

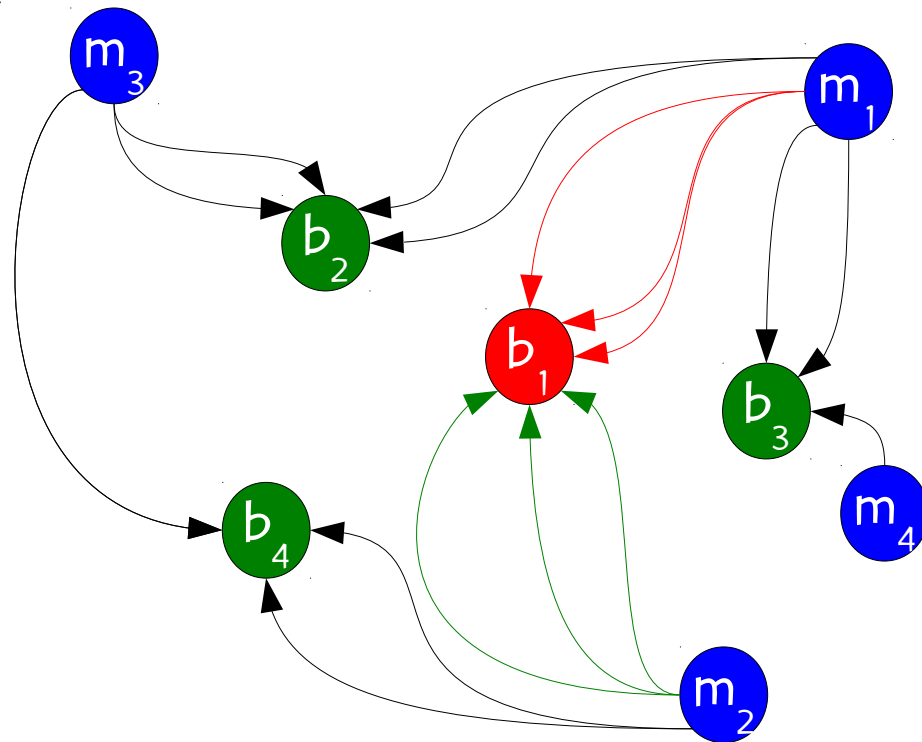
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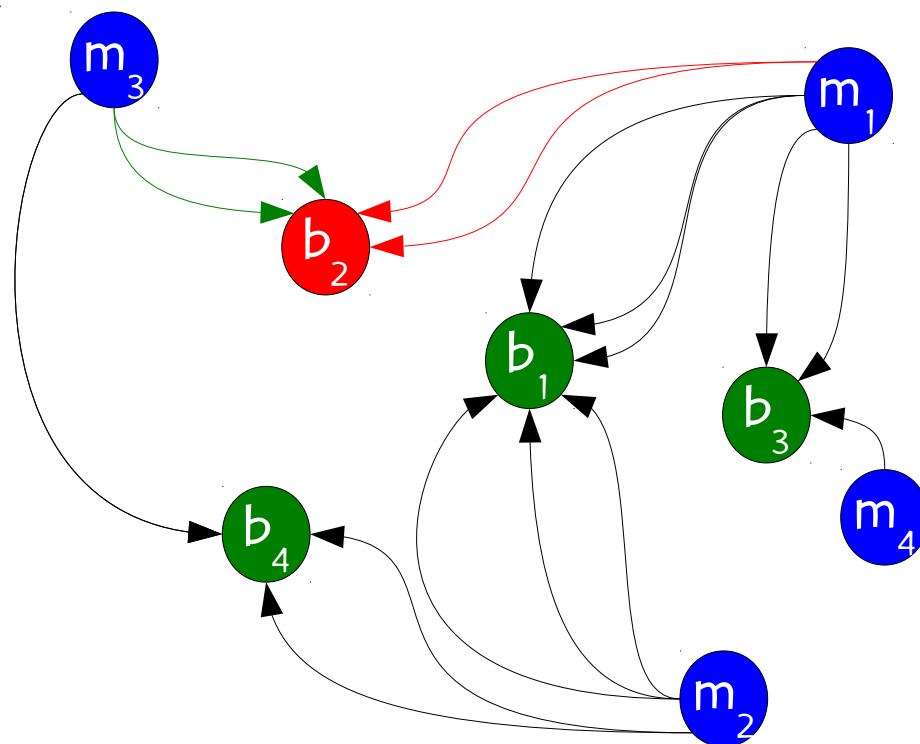
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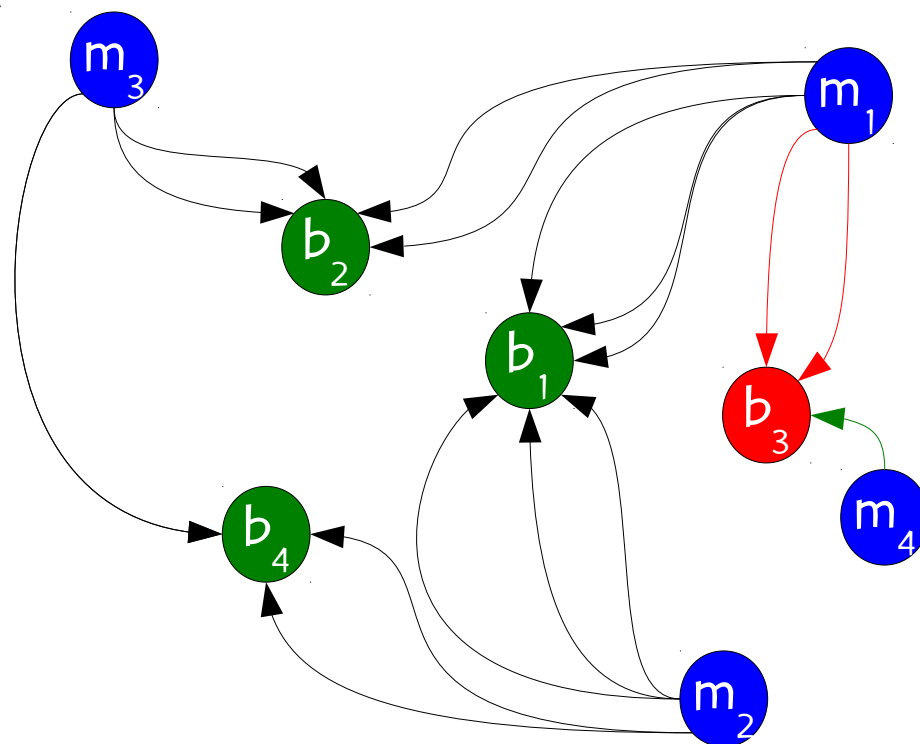
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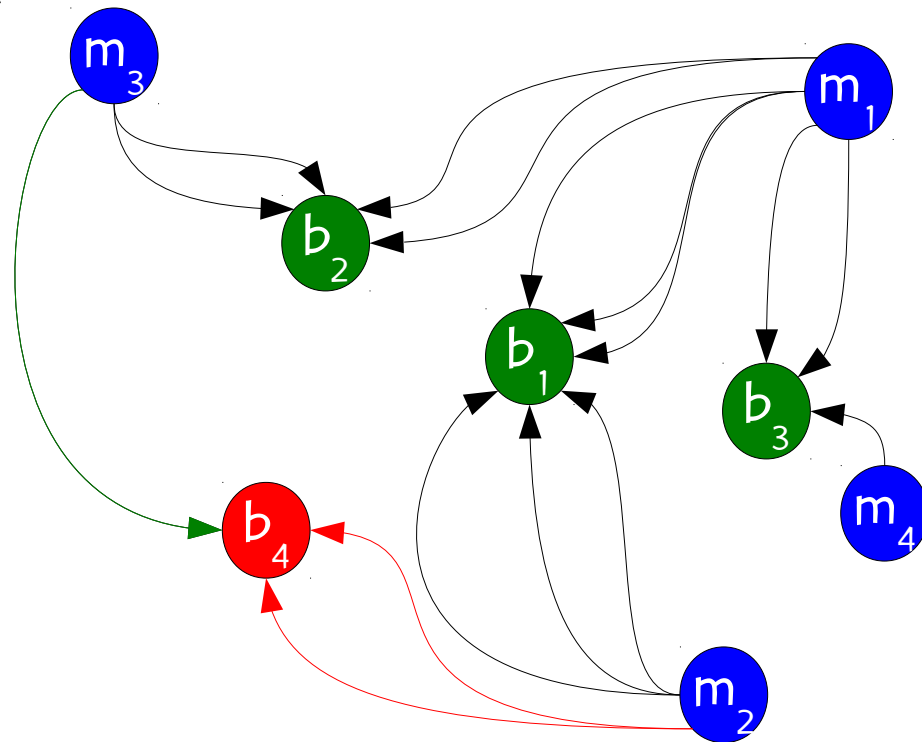
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# Redundant content distribution

- Given:
  - A directed network  $G = (V, E)$ ;
  - A set of nodes  $B \subseteq E$  where content-demanding users are located;
  - A set of nodes  $M \subseteq E$  where content warehouses can be located;
  - The set of all OSPF paths from  $m$  to  $b$ , for  $m \in M$  and  $b \in B$ .

# Redundant content distribution

- Compute:
  - The set of triples  $\{ m_1, m_2, b \}^i, i = 1, 2, \dots, T$ , such that all paths from  $m_1$  to  $b$  and from  $m_2$  to  $b$  are disjoint, where  $m_1, m_2 \in M$  and  $b \in B$ .
  - Note that if  $B \cap M \neq \emptyset$ , then some triples will be of the type  $\{ b, b, b \}$ , where  $b \in B \cap M$ , i.e. a data warehouse that is co-located with a user can provide content to the user by itself.

# Redundant content distribution

- Solve the covering by pairs problem:
  - Find a smallest-cardinality set  $M^* \subseteq M$  such that for all  $b \in B$ , there exists a triple  $\{m_1, m_2, b\}$  in the set of triples such that  $m_1, m_2 \in M^*$ .

# Greedy algorithm for covering by pairs

- initialize partial cover  $M^* = \{ \}$

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- initialize partial cover  $M^* = \{ \}$
- while  $M^*$  is not a cover do:
  - find  $m \in M \setminus M^*$  such that  $M^* \cup \{m\}$  covers a maximum number of additional user nodes (break ties by vertex index) and set  $M^* = M^* \cup \{m\}$

# Greedy algorithm for covering by pairs

- initialize partial cover  $M^* = \{ \}$
- while  $M^*$  is not a cover do:
  - find  $m \in M \setminus M^*$  such that  $M^* \cup \{m\}$  covers a maximum number of additional user nodes (break ties by vertex index) and set  $M^* = M^* \cup \{m\}$
  - if no  $m \in M \setminus M^*$  yields an increase in coverage, then choose a pair  $\{m_1, m_2\} \in M \setminus M^*$  that yields a maximum increase in coverage and set  $M^* = M^* \cup \{m_1\} \cup \{m_2\}$



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  - if no pair exists, then the problem is infeasible

# BRKGA for redundant content distribution

# BRKGA for the RCD problem

- Encoding:
  - A vector  $X$  of  $N$  keys randomly generated in the real interval  $(0,1]$ , where  $N = |M|$  is the number of potential data warehouse nodes. The  $i$ -th random key corresponds to the  $i$ -th potential data warehouse node.

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  - For  $i = 1, \dots, N$ : if  $X(i) > 1/2$ , add  $i$ -th data warehouse node to solution

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- If solution is feasible, i.e. all users are covered: **STOP**

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- For  $i = 1, \dots, N$ : if  $X(i) > 1/2$ , add  $i$ -th data warehouse node to solution
- If solution is feasible, i.e. all users are covered: **STOP**
- Else, apply greedy algorithm to cover uncovered user nodes.

# BRKGA for the RCD problem

- Size of population:  $N$  (number of monitoring nodes)
- Size of elite set: 15% of  $N$
- Size of mutant set: 10% of  $N$
- Biased coin probability: 70%
- Stop after  $N$  generations without improvement of best found solution



# Another application: Host placement for end-to-end monitoring

- Internet service provider (ISP) delivers virtual private network (VPN) service to customers.

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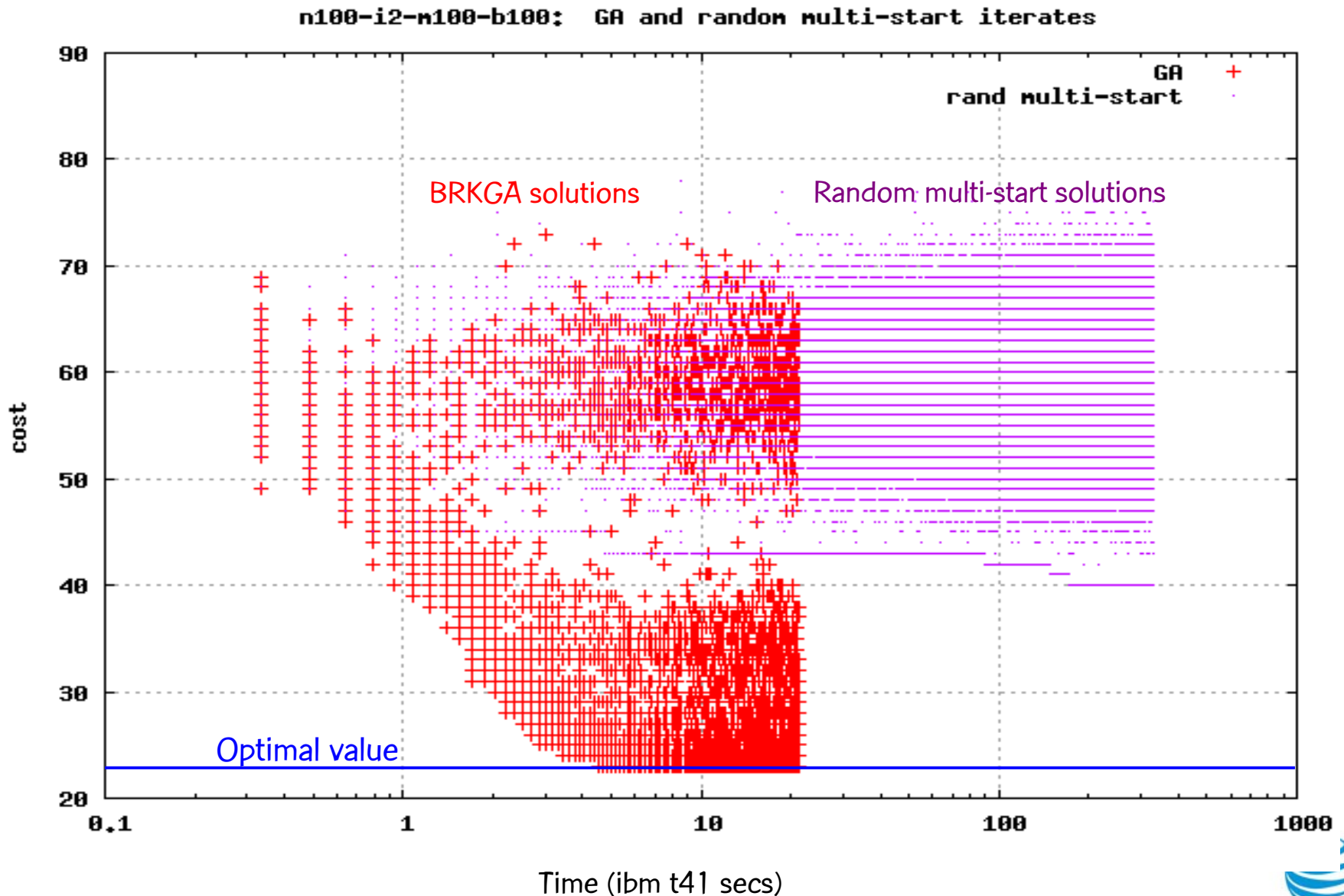
# Another application: Host placement for end-to-end monitoring

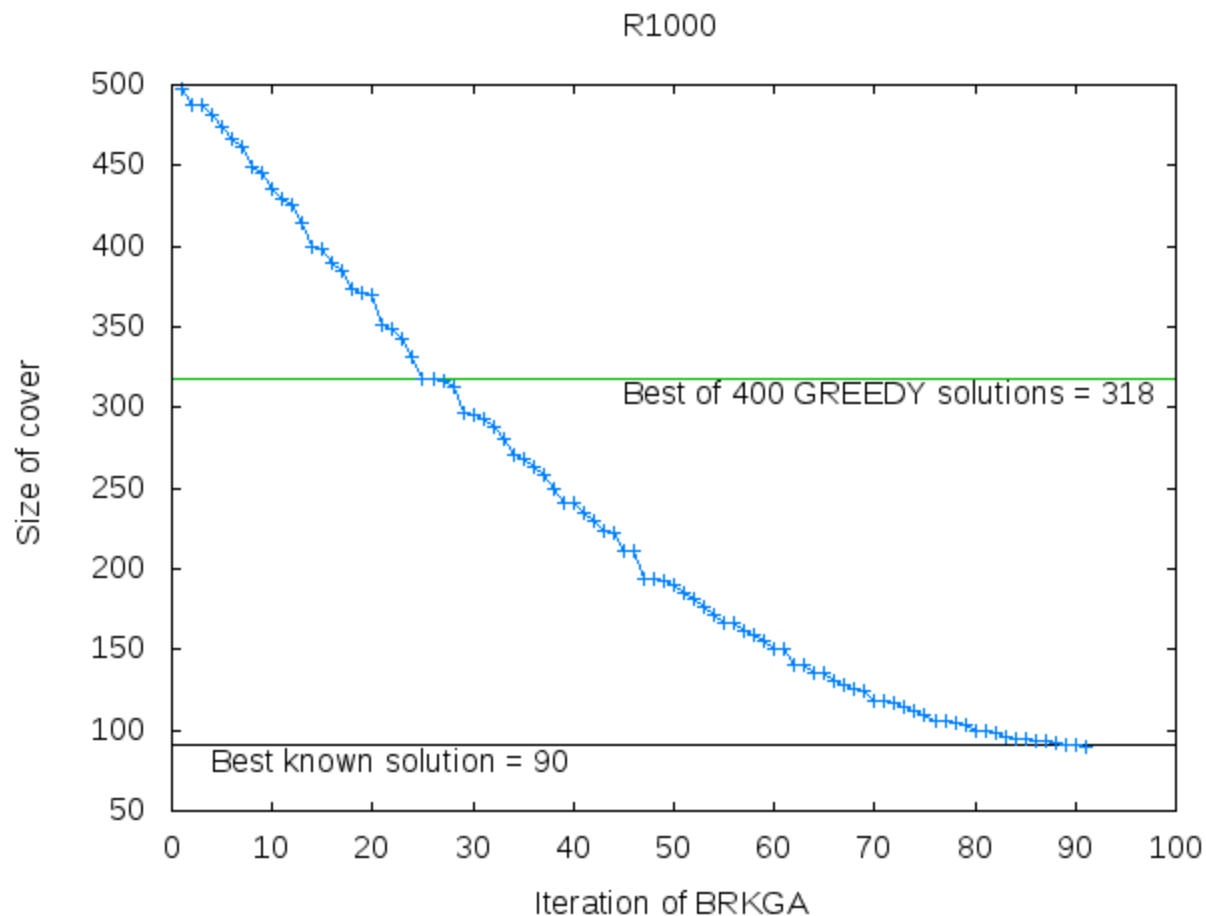
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# Another application: Host placement for end-to-end monitoring

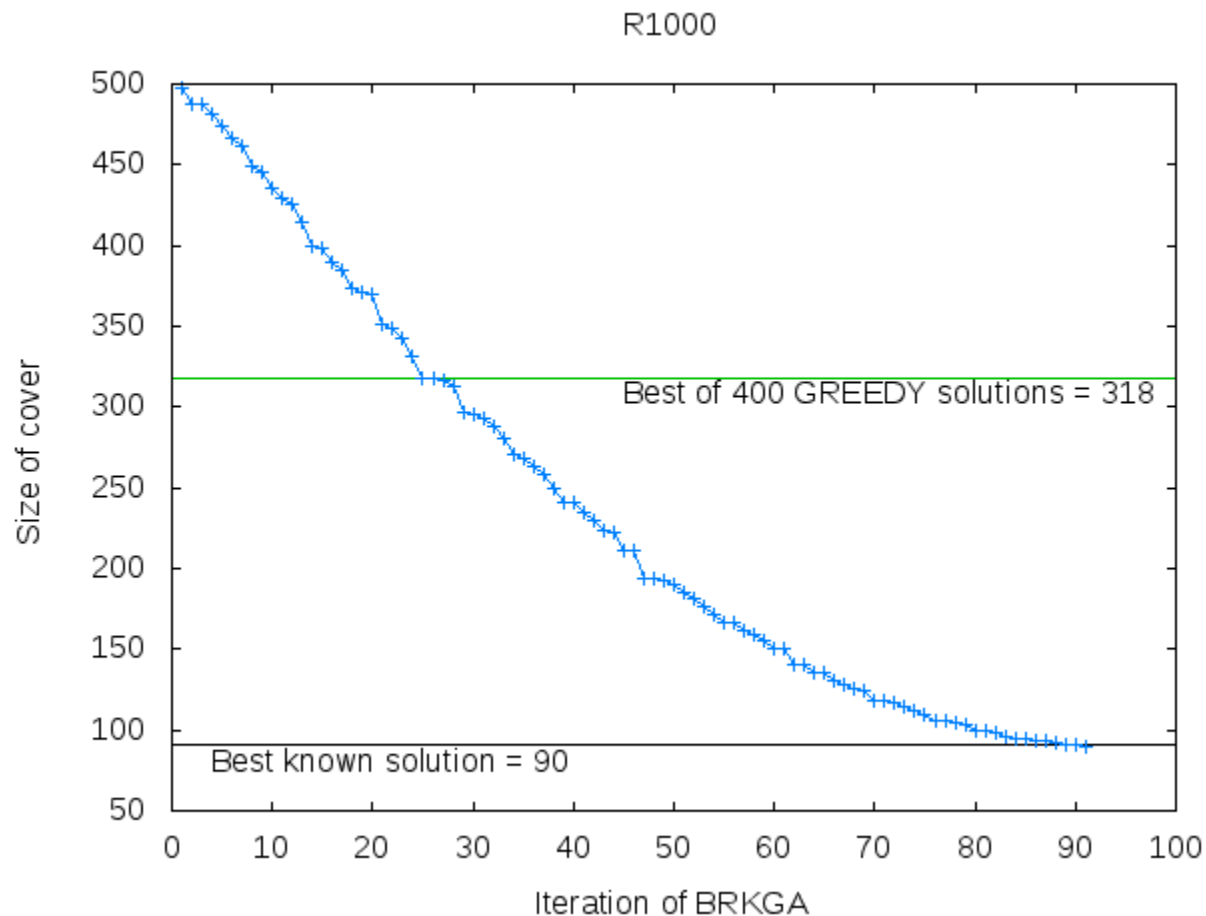
- Internet service provider (ISP) delivers virtual private network (VPN) service to customers.
- The ISP agrees to send traffic between locations specified by the customer and promises to provide certain level of service on the connections.
- A key service quality metric is packet loss rate.
- We want to minimize the number of monitoring equipment placed in the network to measure packet loss rate: This is a type of covering by pairs problem.

solution





Real-world instance derived from a proprietary Tier-1 Internet Service Provider (ISP) backbone network using OSPF for routing.



Size of network: about 1000 nodes, where almost all can store content and about 90% have content-demanding users. Over 45 million triples.

# Regenerator location problem



# Reference

A. Duarte, R. Martí, M.G.C.R., and R.M.A. Silva, "Improved heuristics for the regenerator location problem," *International Transactions in Operational Research*, to appear in 2014.

Tech report version:

<http://www.research.att.com/~mgcr/doc/gpr-regenloc.pdf>

# Signal regeneration

- Telecommunication systems use optical signals to transmit information

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- Telecommunication systems use optical signals to transmit information
- Strength of signal deteriorates and loses power as it gets farther from source
- Signal must be regenerated periodically to reach destination: Regenerators
- Regenerators are expensive: minimize the number of regenerators in the network

# Regenerator location problem (RLP)

- Given:
  - Graph  $G=(V,E)$ , where  $V$  are vertices,  $E$  are edges, where edge  $(i,j)$  has a real-valued length  $d(i,j) > 0$
  - $D$  is the maximum length that a signal can travel before it must be regenerated

# Regenerator location problem (RLP)

- Find:
  - Paths that connect all pairs of nodes in  $V \times V$
  - Nodes where it is necessary to locate single regenerators
- Minimize number of deployed regenerators

# Regenerator location problem (RLP)

- Path between  $\{s,t\} \in E$ 
  - $\{ (s,v[1]), (v[1],v[2]), \dots, (v[k],t) \}$  is formed by one or more path segments
- Path segment is sequence of consecutive edges
  - $\{ (v[i],v[i+1]), (v[i+1],v[i+2]), \dots, (v[q-1],v[q]) \}$  in the path satisfying the condition

$$d(v[i],v[i+1]) + d(v[i+1],v[i+2]) + \dots + d(v[q-1],v[q]) \leq D$$

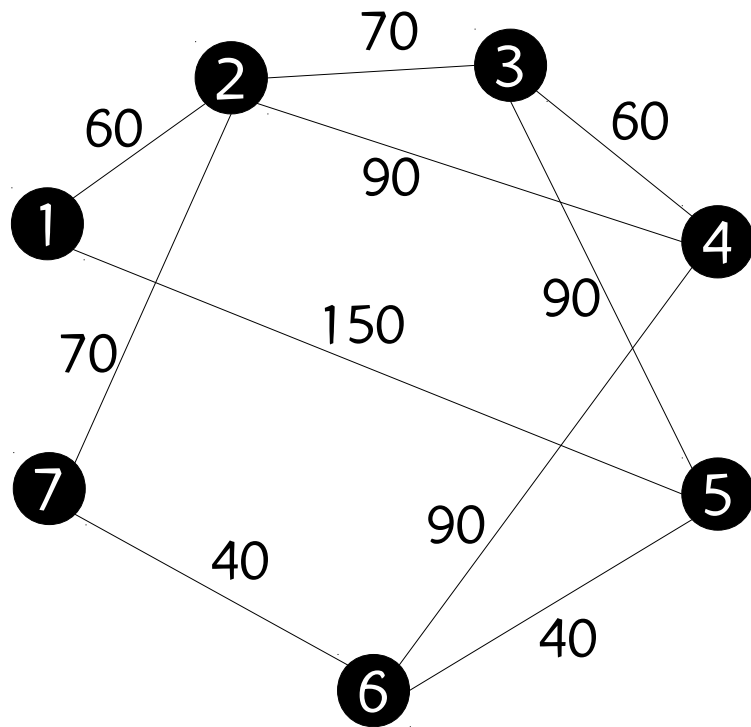


# Regenerator location problem (RLP)

- If total length of path is no more than  $D$ , then path consists of a single path segment
- Otherwise, it consists of one or more segments
  - Regenerators will be located in the internal nodes of the path

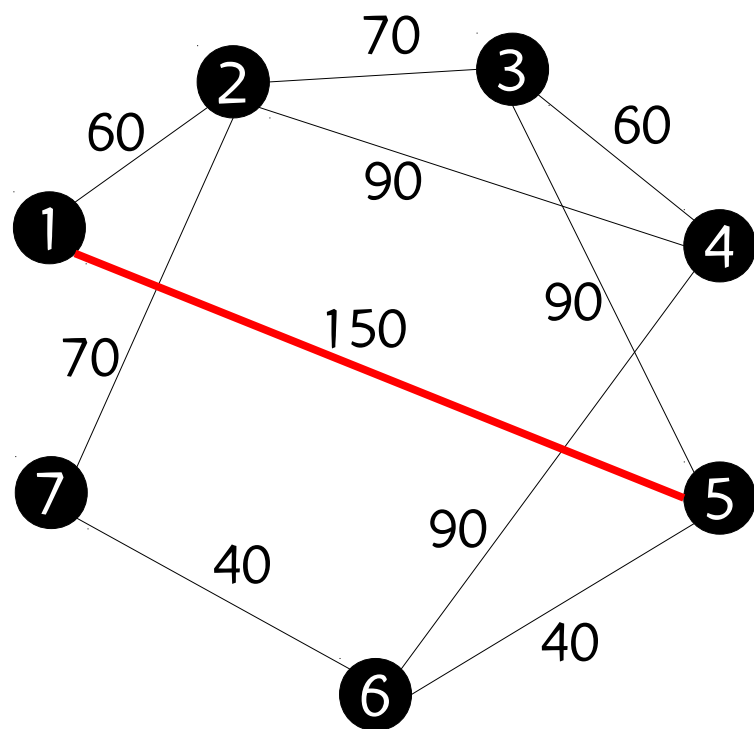
# Regenerator location problem (RLP)

7-node graph with  $D = 100$



$D = 100$

# Regenerator location problem (RLP)

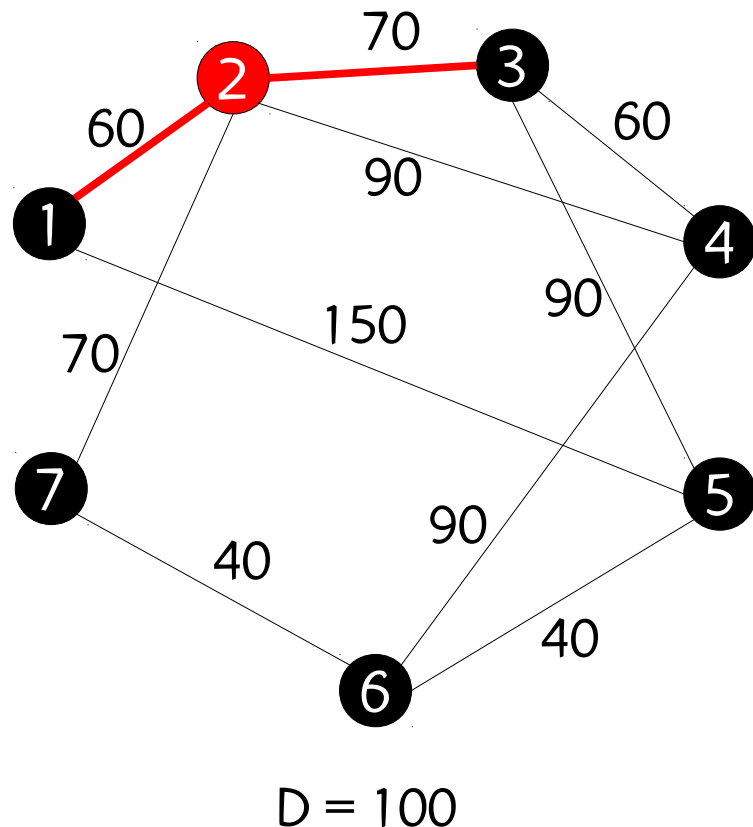


$D = 100$

(1) Note that:

- $D(1,5) = 150 > 100 = D$
- Edge (1,5) cannot be part of any path

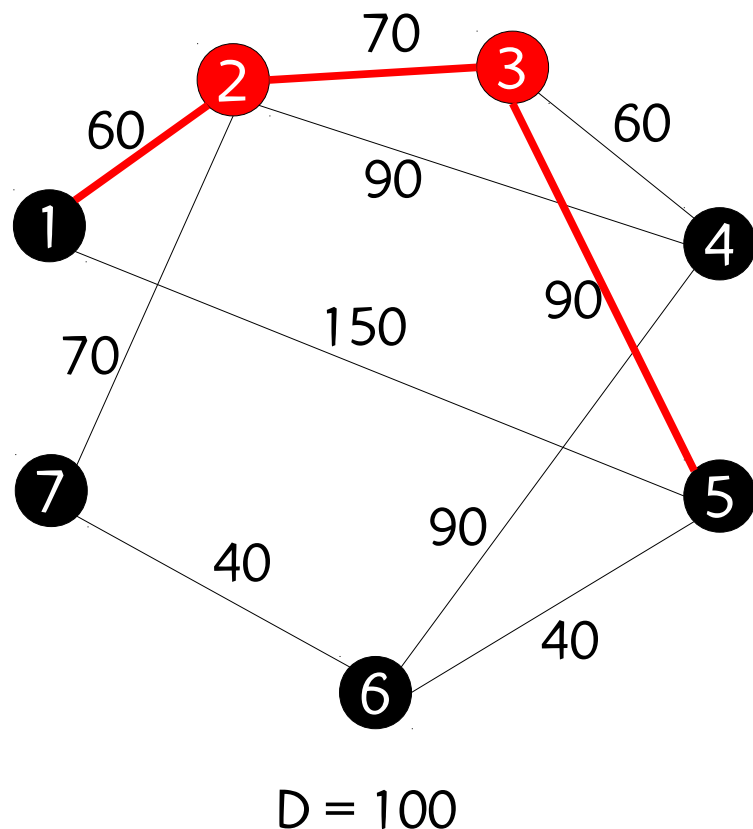
# Regenerator location problem (RLP)



(2) Note that:

- Shortest path from 1 to 3 is  $\{ (1,2), (2,3) \}$  with total length  
 $60 + 70 = 130 > 100 = D$
- Must be decomposed into two path segments  $\{ (1,2) \}$  and  $\{ (2,3) \}$  with a regenerator in node 2

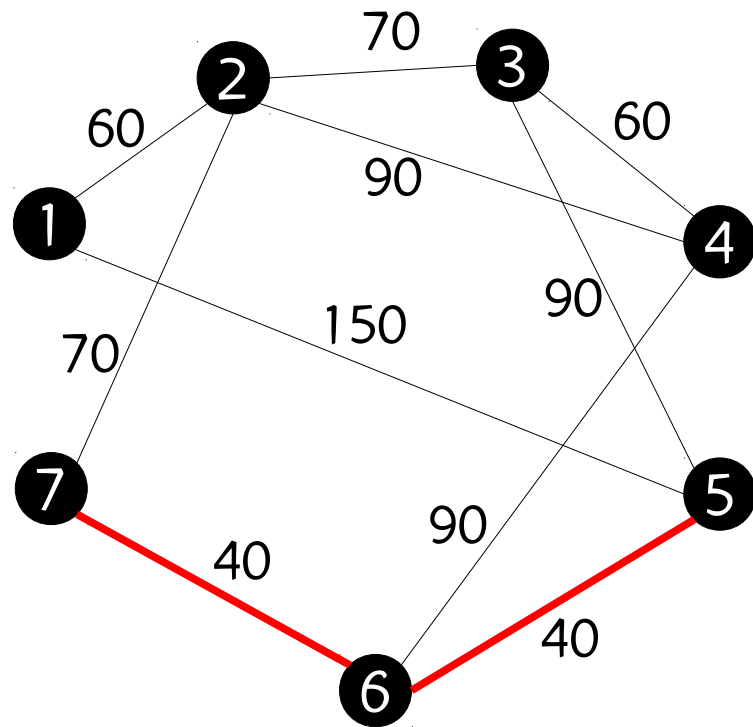
# Regenerator location problem (RLP)



(3) Note that:

- Shortest feasible path from 1 to 5 is  $\{ (1,2), (2,3), (3,5) \}$  with total length  $60 + 70 + 90 = 220 > 100 = D$
- Must be decomposed into three path segments  $\{ (1,2) \}$ ,  $\{ (2,3) \}$ , and  $\{ (3,5) \}$  with regenerators in nodes 2 and 3

# Regenerator location problem (RLP)

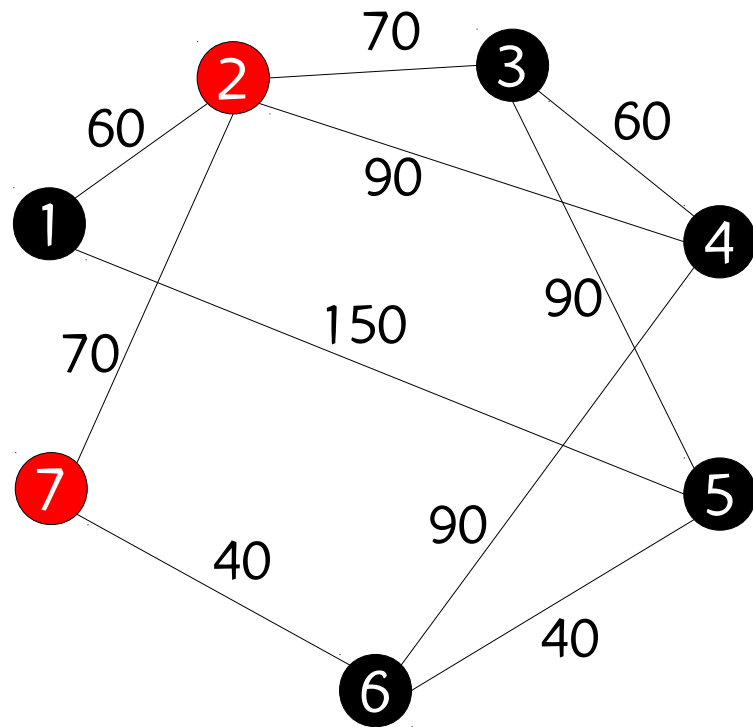


$D = 100$

(4) Note that:

- Shortest feasible path from 5 to 7 is  $\{ (5,6), (6,7) \}$  with total length  $40 + 40 = 80 \leq 100 = D$
- No regenerator is needed to connect nodes 5 and 7

# Regenerator location problem (RLP)

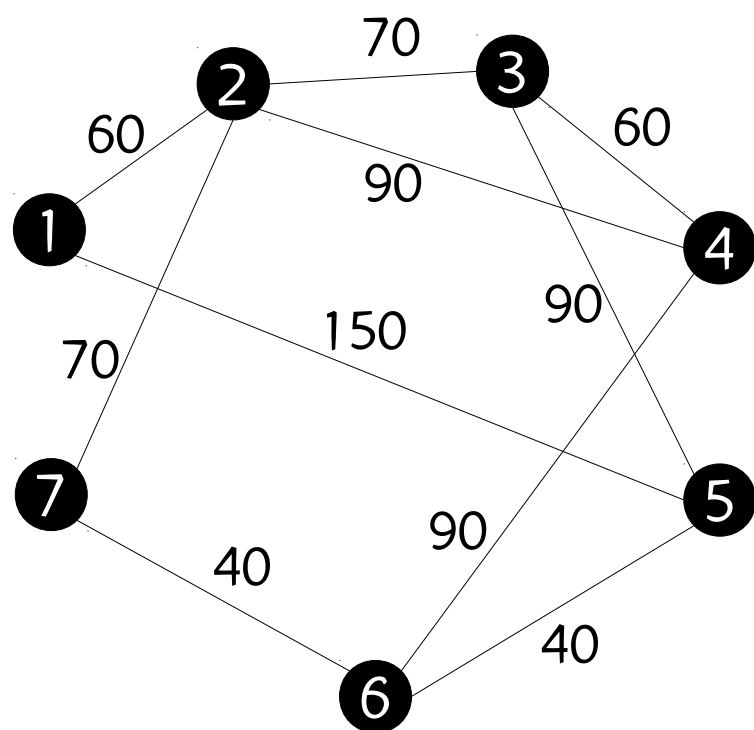


$D = 100$

(5) Note that:

- Placing regenerator in nodes 2 and 7 allows for communication between all pairs of nodes in the graph

# Communication graph (Chen et al., 2010)

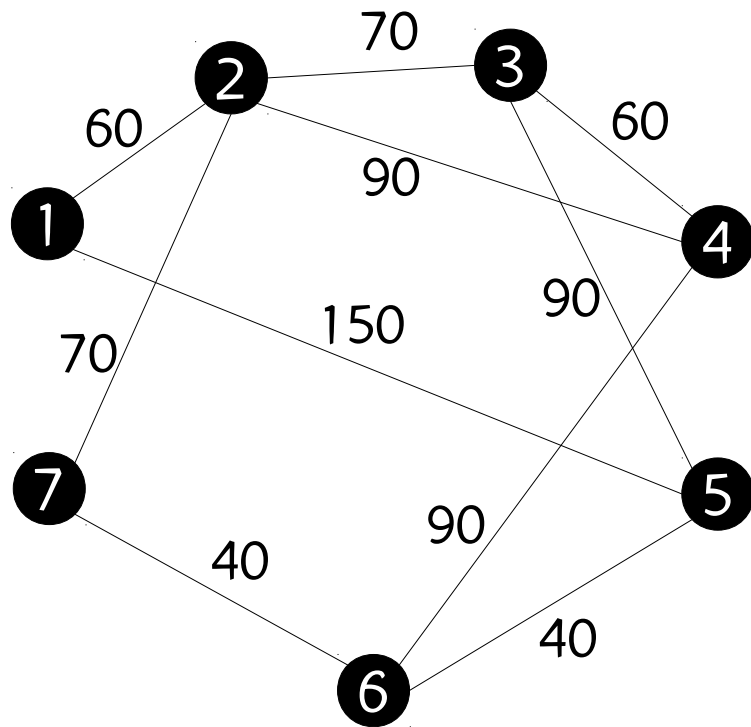


$G =$   
 $(V, E)$

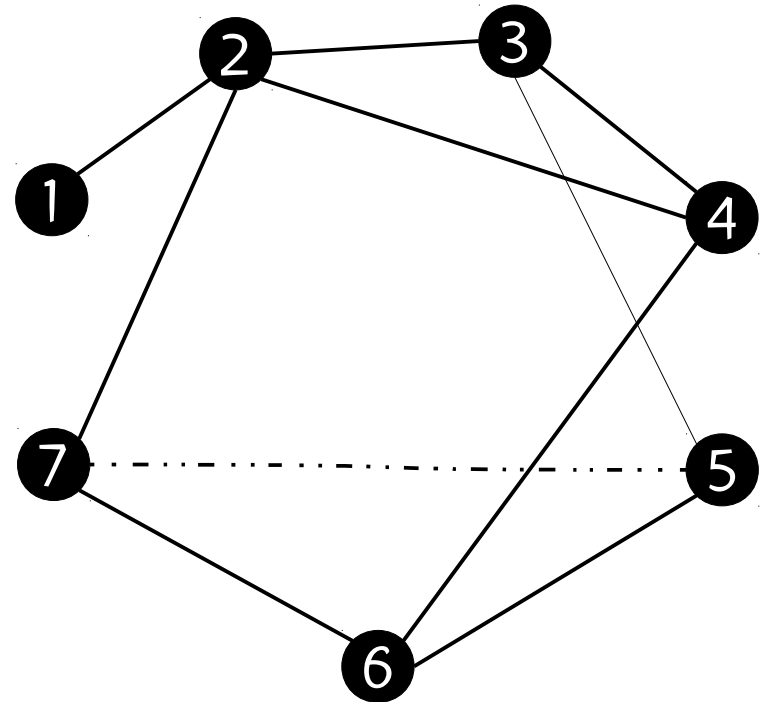
- Given weighted graph  $G$ 
  - Delete all edges having length greater than  $D$
  - For all non-adjacent nodes, add an edge between them of length equal to the corresponding shortest path in  $G$  if it is less than  $D$
  - Disregard all length info



# Communication graph (Chen et al., 2010)



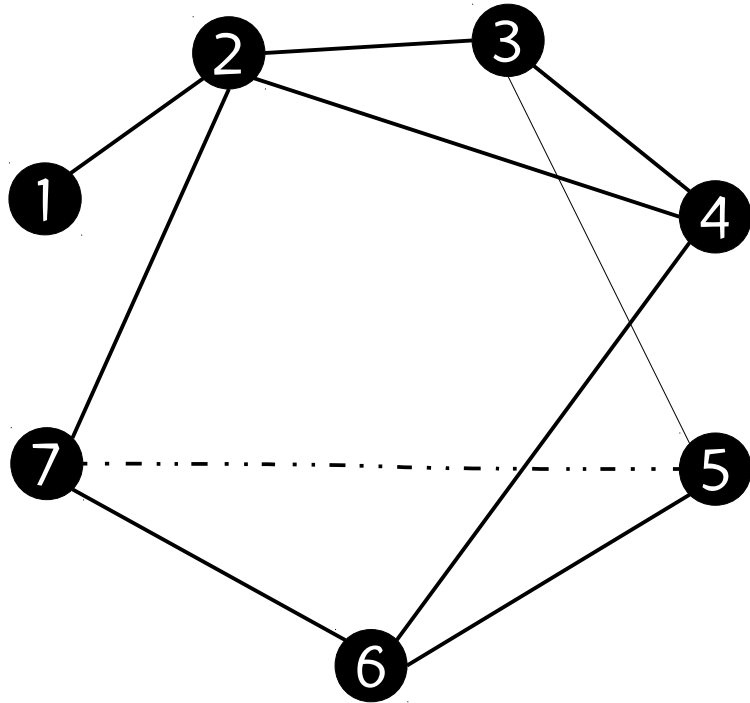
$G =$   
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$M = (V, E')$

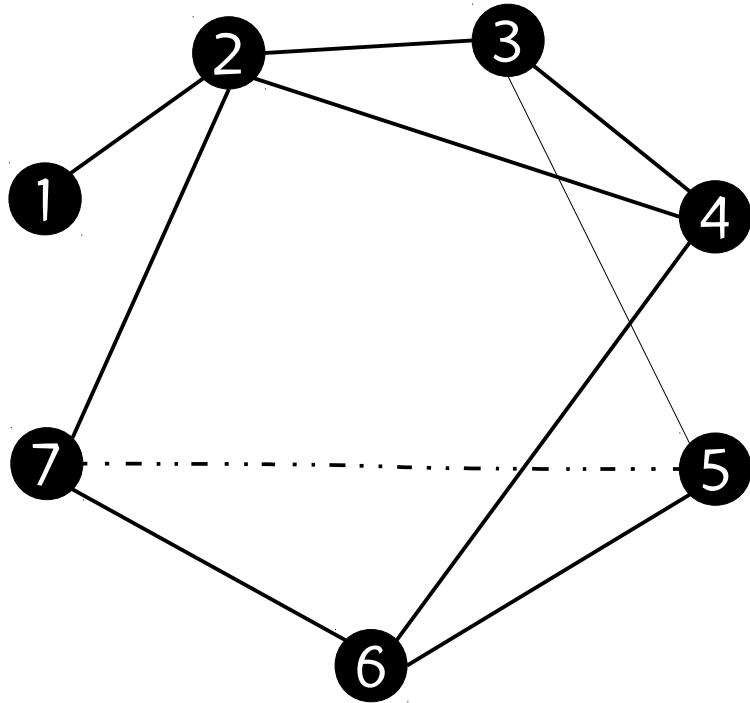
# Communication graph (Chen et al., 2010)

- If  $M$  is complete, then there is no need for regenerators



$M = (V, E')$

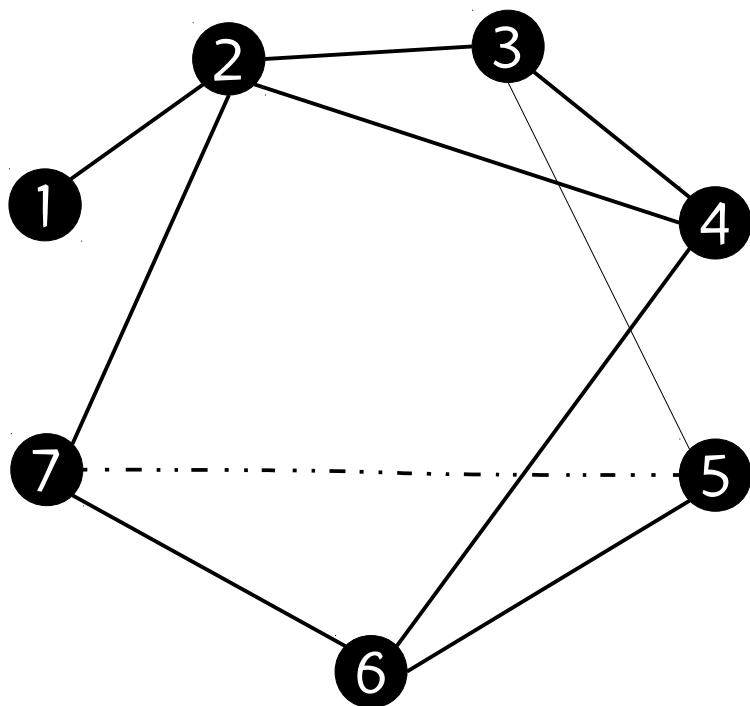
# Communication graph (Chen et al., 2010)



$M = (V, E')$

- If  $M$  is complete, then there is no need for regenerators
- If  $M$  is not connected, then the problem is infeasible

# Communication graph (Chen et al., 2010)

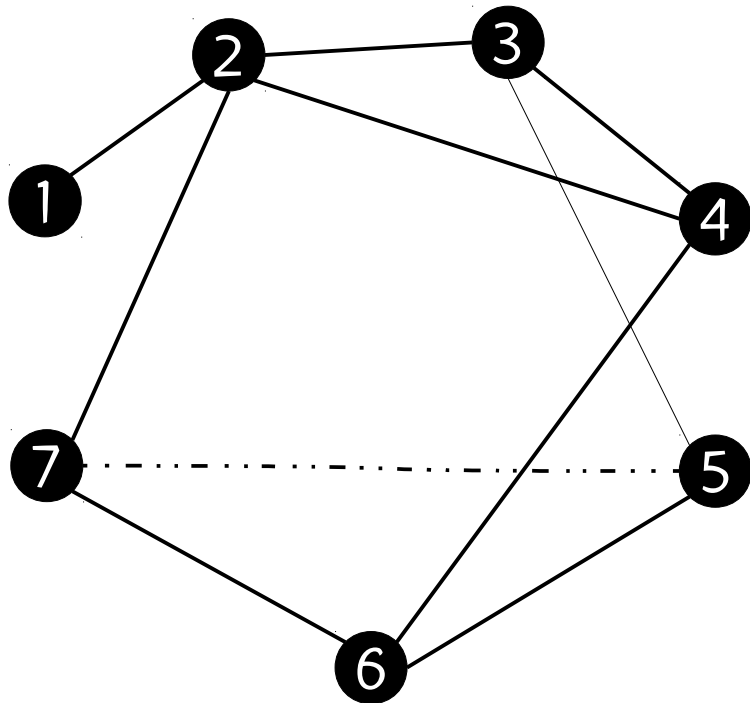


$M = (V, E')$

- If  $M$  is complete, then there is no need for regenerators
- If  $M$  is not connected, then the problem is infeasible
- Otherwise, one or more regenerators are needed

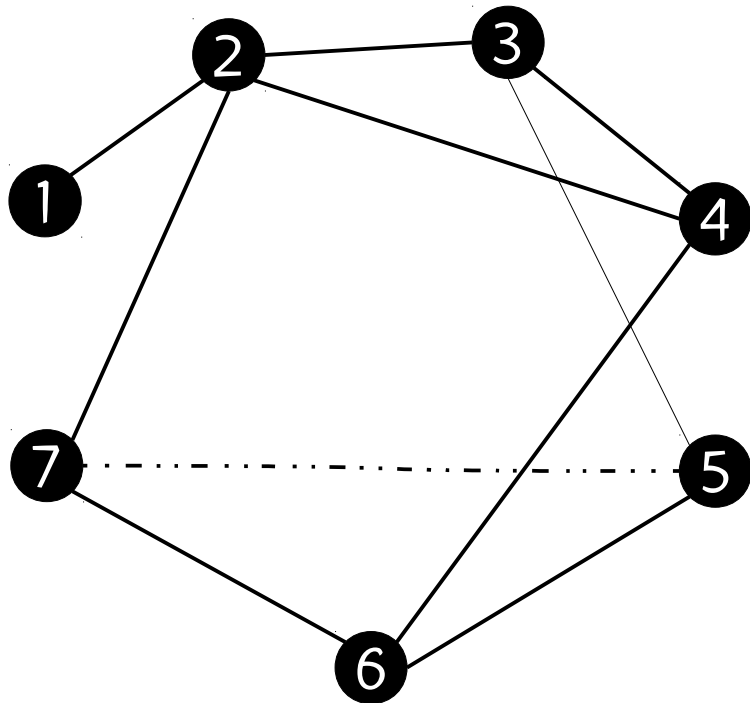
# Greedy algorithm (Chen et al., 2010)

- Works on communication graph  $M$



$M = (V, E')$

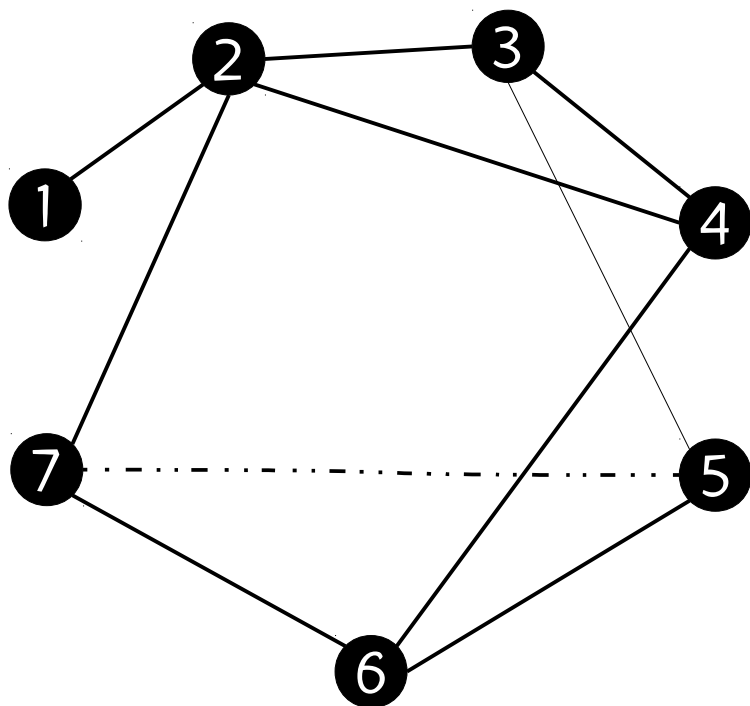
# Greedy algorithm (Chen et al., 2010)



$M = (V, E')$

- Works on communication graph  $M$
- Input: set of nodes not directly connected (NDC) in  $M$  and builds a set  $R$  of regenerator nodes

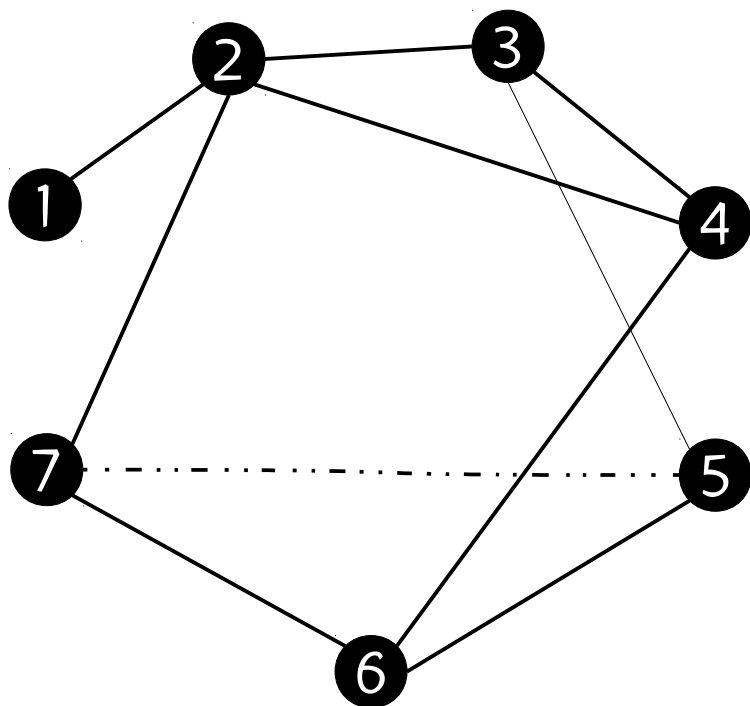
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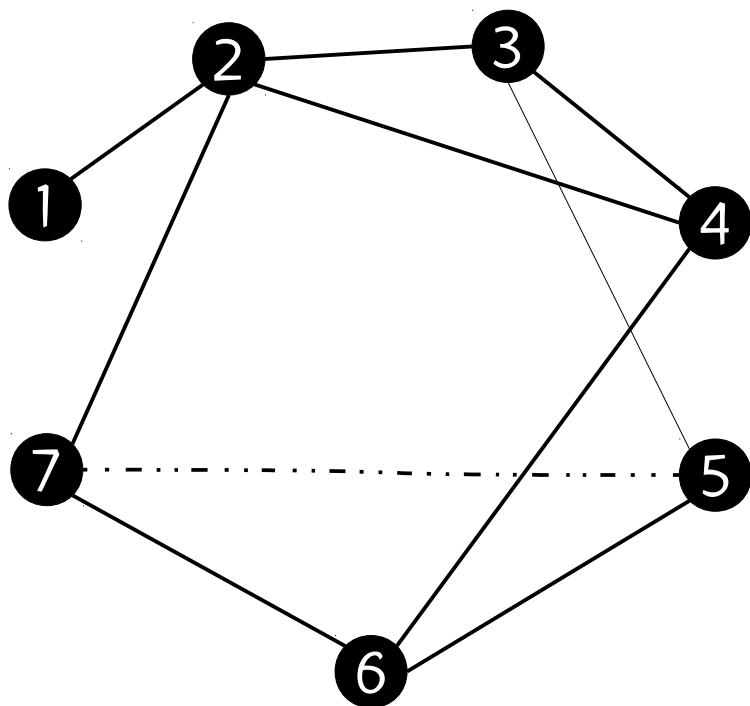


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- At each step the procedure determines a node  $u^*$  whose inclusion in  $R$  enables the connection of the largest number  $g(u^*)$  of yet unconnected pairs  $X(u^*)$  in  $M$
- Node  $u^*$  is added to  $R$  and  $M$  is updated



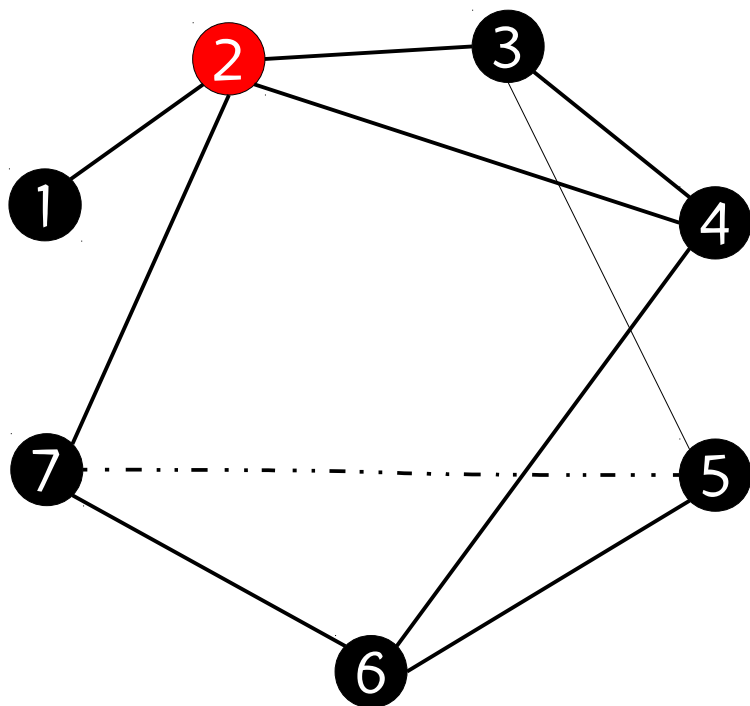
# Greedy algorithm (Chen et al., 2010)



$M = (V, E')$

u	$X(u)$	$g(u)$
1	$\emptyset$	0
2	$\{ (1,3), (1,4), (1,7), (3,7), (4,7) \}$	5
3	$\{ (4,5), (2,5) \}$	2
4	$\{ (2,6), (3,6) \}$	2
5	$\{ (3,7), (3,6) \}$	2
6	$\{ (4,7), (4,5) \}$	2
7	$\{ (2,6), (2,5) \}$	2

# Greedy algorithm (Chen et al., 2010)

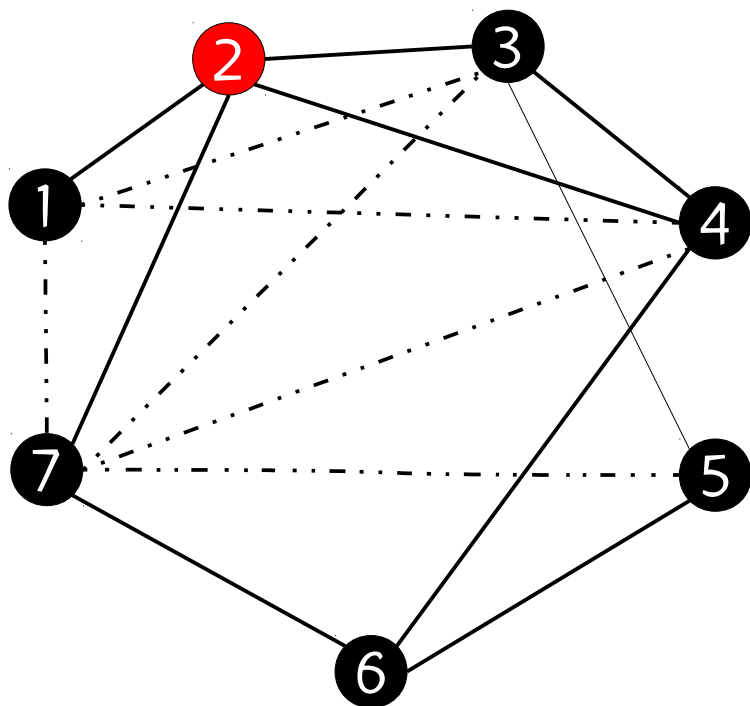


$M = (V, E')$

u	$X(u)$	$g(u)$
1	$\emptyset$	0
<b>2</b>	<b><math>\{ (1,3), (1,4), (1,7), (3,7), (4,7) \}</math></b>	<b>5</b>
3	$\{ (4,5), (2,5) \}$	2
4	$\{ (2,6), (3,6) \}$	2
5	$\{ (3,7), (3,6) \}$	2
6	$\{ (4,7), (4,5) \}$	2
7	$\{ (2,6), (2,5) \}$	2

Add regenerator to node 2

# Greedy algorithm (Chen et al., 2010)

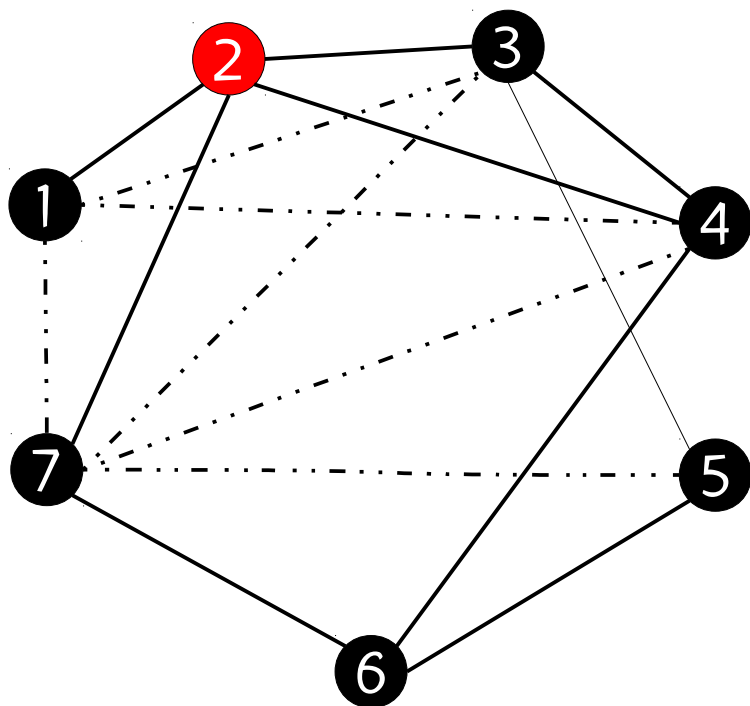


$M = (V, E')$

u	$X(u)$	$g(u)$
1	$\emptyset$	0
<b>2</b>	<b><math>\{ (1,3), (1,4), (1,7), (3,7), (4,7) \}</math></b>	<b>5</b>
3	$\{ (4,5), (2,5) \}$	2
4	$\{ (2,6), (3,6) \}$	2
5	$\{ (3,7), (3,6) \}$	2
6	$\{ (4,7), (4,5) \}$	2
7	$\{ (2,6), (2,5) \}$	2

Update  $M$  to account for regenerator in node 2

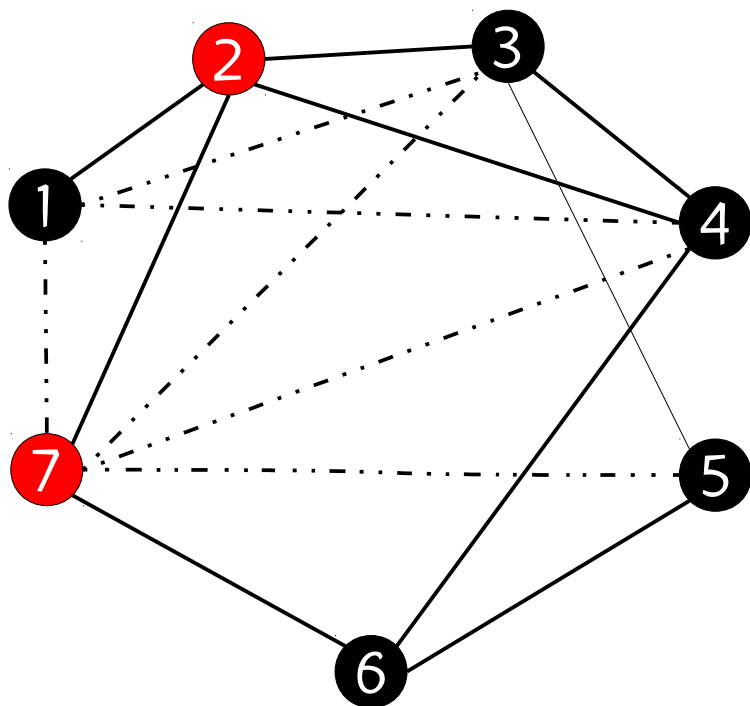
# Greedy algorithm (Chen et al., 2010)



$M = (V, E')$

u	$X(u)$	$g(u)$
1	$\emptyset$	0
-	-	-
3	$\{ (1,5), (2,5), (4,5) \}$	3
4	$\{ (1,6), (2,6), (3,6) \}$	3
5	$\{ (3,6) \}$	1
6	$\{ (4,5) \}$	1
7	$\{ (1,5), (1,6), (2,5), (2,6), (3,6), (4,5) \}$	6

# Greedy algorithm (Chen et al., 2010)

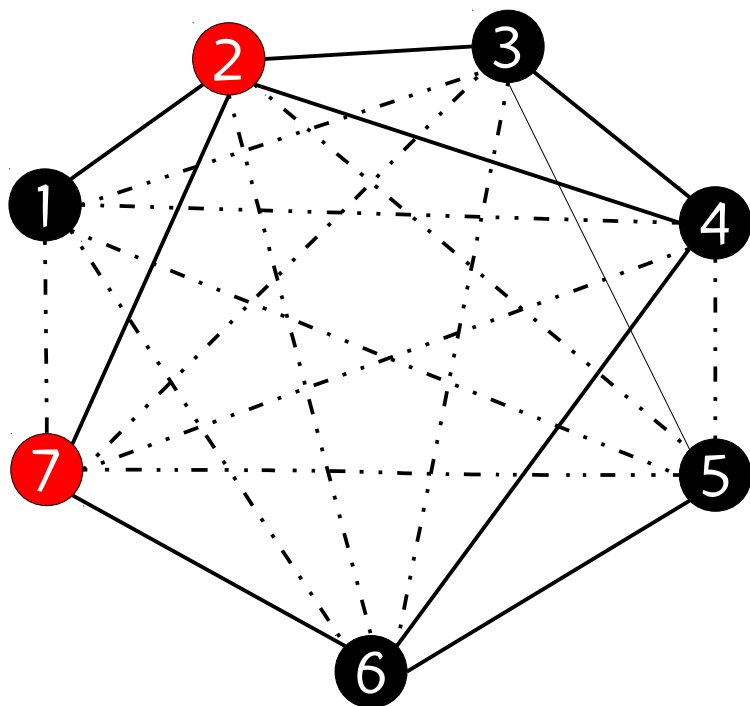


$M = (V, E')$

u	$X(u)$	$g(u)$
1	$\emptyset$	0
-	-	-
3	$\{(1,5), (2,5), (4,5)\}$	3
4	$\{(1,6), (2,6), (3,6)\}$	3
5	$\{(3,6)\}$	1
6	$\{(4,5)\}$	1
<b>7</b>	<b><math>\{(1,5), (1,6), (2,5), (2,6), (3,6), (4,5)\}</math></b>	<b>6</b>

Add regenerator to node 7

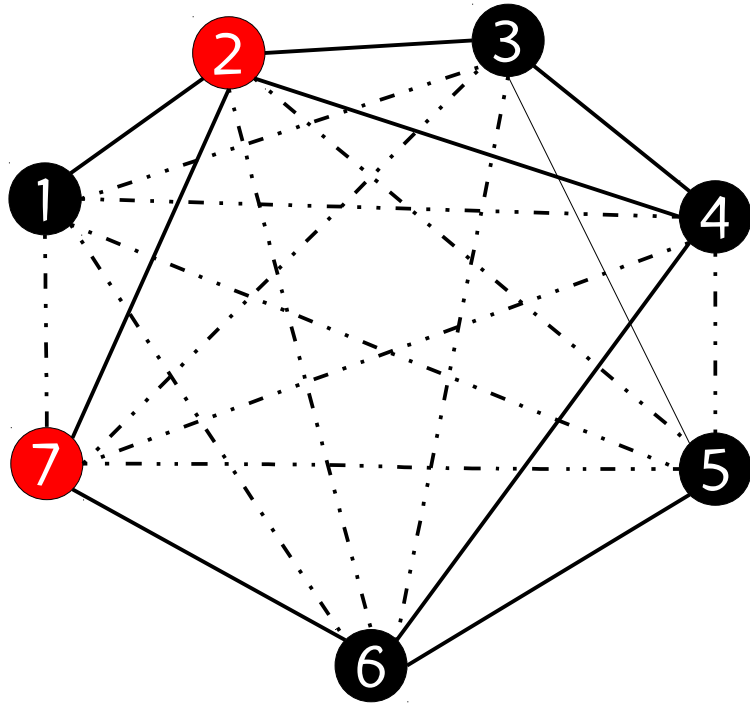
# Greedy algorithm (Chen et al., 2010)



$M = (V, E')$

u	$X(u)$	$g(u)$
1	$\emptyset$	0
-	-	-
3	$\{ (1,5), (2,5), (4,5) \}$	3
4	$\{ (1,6), (2,6), (3,6) \}$	3
5	$\{ (3,6) \}$	1
6	$\{ (4,5) \}$	1
<b>7</b>	<b><math>\{ (1,5), (1,6), (2,5), (2,6), (3,6), (4,5) \}</math></b>	<b>6</b>

# Greedy algorithm (Chen et al., 2010)



$M = (V, E')$

Since  $M$  is complete, all pairs can communicate and solution  $R = \{2, 7\}$

# BRKGA for the regenerator location problem



# Encoding

Solutions are encoded as vectors  $Y$  of  $n = |V|$  random keys, each in the real interval  $[0,1)$

Random key  $Y[i]$  corresponds to node  $i \in V$

# Decoding

Takes as input a communication graph  $M = (V, E')$   
and a vector of random keys  $Y$

Outputs a set of regenerator nodes  $R \subseteq V$

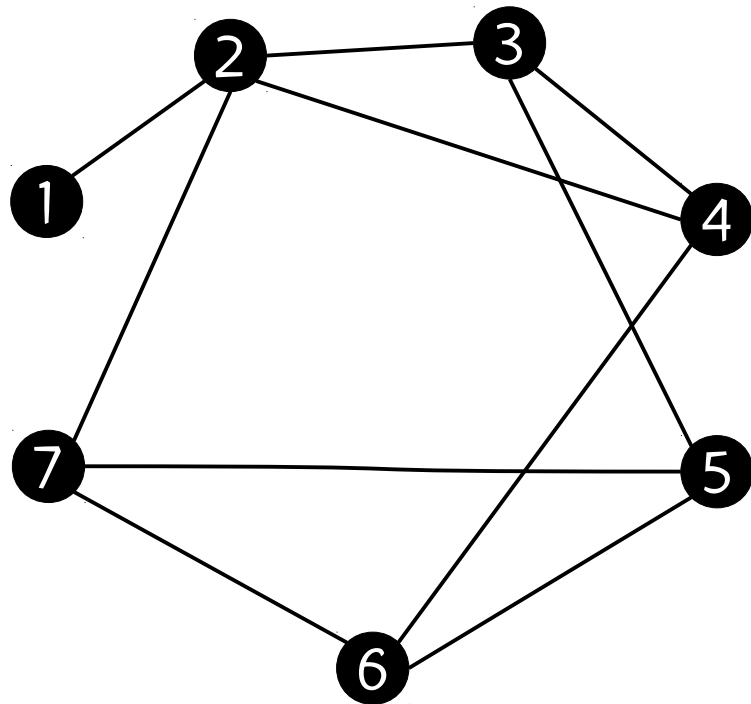
# Decoding

Takes as input a communication graph  $M = (V, E')$   
and a vector of random keys  $Y$

Outputs a set of regenerator nodes  $R \subseteq V$

Sorting  $Y$  implies an ordering of  $V$

# Decoding



$M = (V, E')$

Scan  $V$  in order implied by  $Y$

while some pair in  $V \times V$  cannot communicate ( $M = (V, E')$  is not complete):

Add next vertex  $v$  in order into  $R$

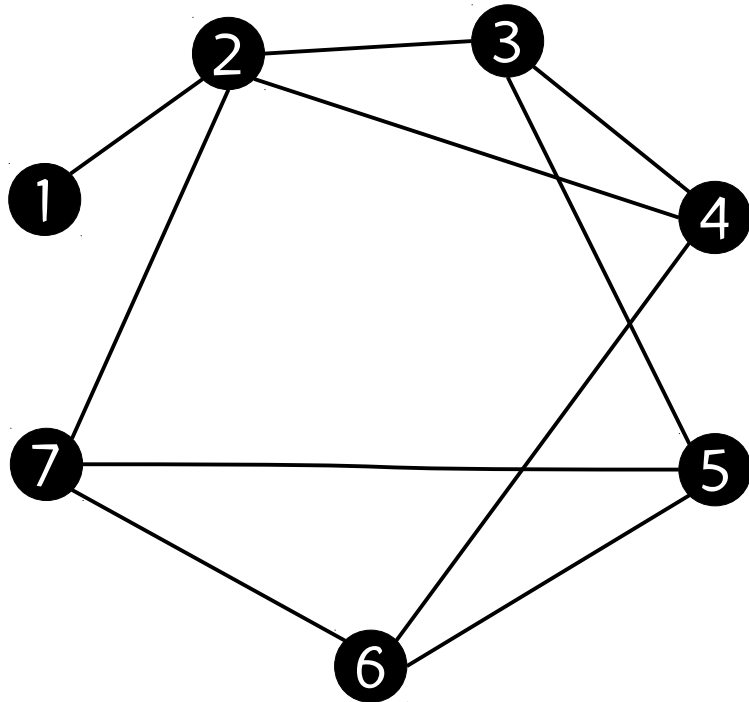
Compute set  $X$  of pairs that do not communicate that would if  $v$  becomes a regenerator

Add  $X$  to  $E'$

end while

return  $R$

# Decoding



$X = (0.1, 0.2, 0.3, 0.4, 0.5, 0.6, 0.7)$

Scan  $V$  in order implied by  $Y$

while some pair in  $V \times V$  cannot communicate ( $M = (V, E')$  is not complete):

Add next vertex  $v$  in order into  $R$

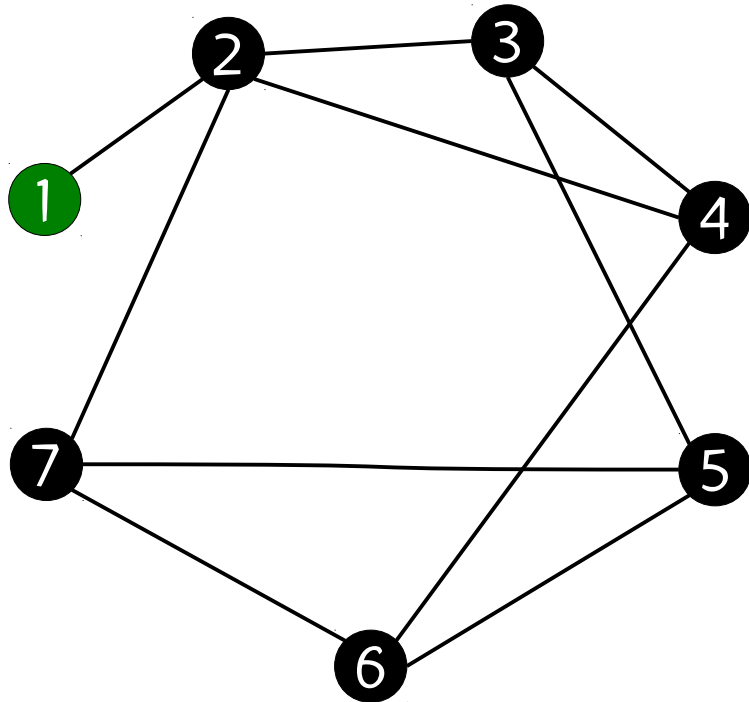
Compute set  $X$  of pairs that do not communicate that would if  $v$  becomes a regenerator

Add  $X$  to  $E'$

end while

return  $R$

# Decoding



$X = (0.1, 0.2, 0.3, 0.4, 0.5, 0.6, 0.7)$

↑  
 $i = 1$

Scan  $V$  in order implied by  $Y$

while some pair in  $V \times V$  cannot communicate ( $M = (V, E')$  is not complete):

Add next vertex  $v$  in order into  $R$

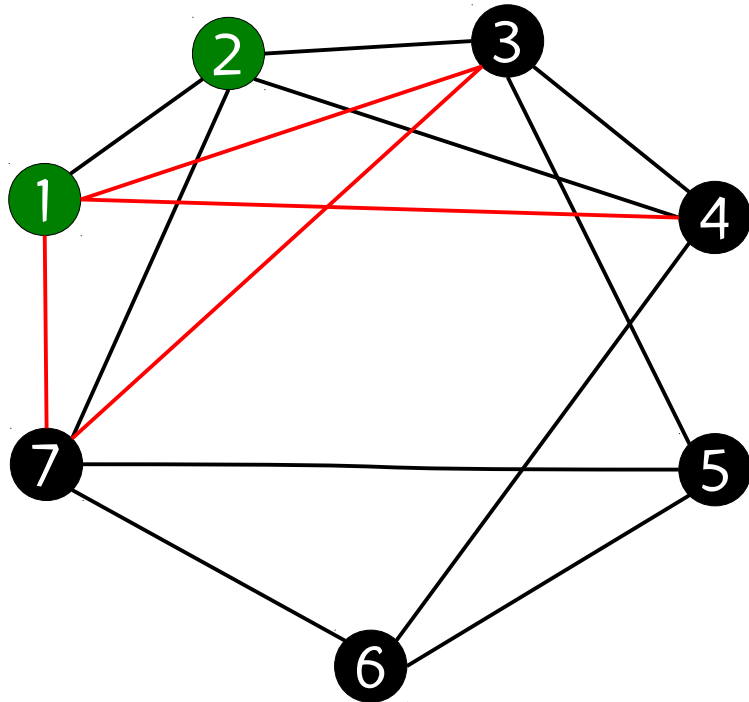
Compute set  $X$  of pairs that do not communicate that would if  $v$  becomes a regenerator

Add  $X$  to  $E'$

end while

return  $R$

# Decoding



$X = (0.1, 0.2, 0.3, 0.4, 0.5, 0.6, 0.7)$

↑  
 $i = 2$

Scan  $V$  in order implied by  $Y$

while some pair in  $V \times V$  cannot communicate ( $M = (V, E')$  is not complete):

Add next vertex  $v$  in order into  $R$

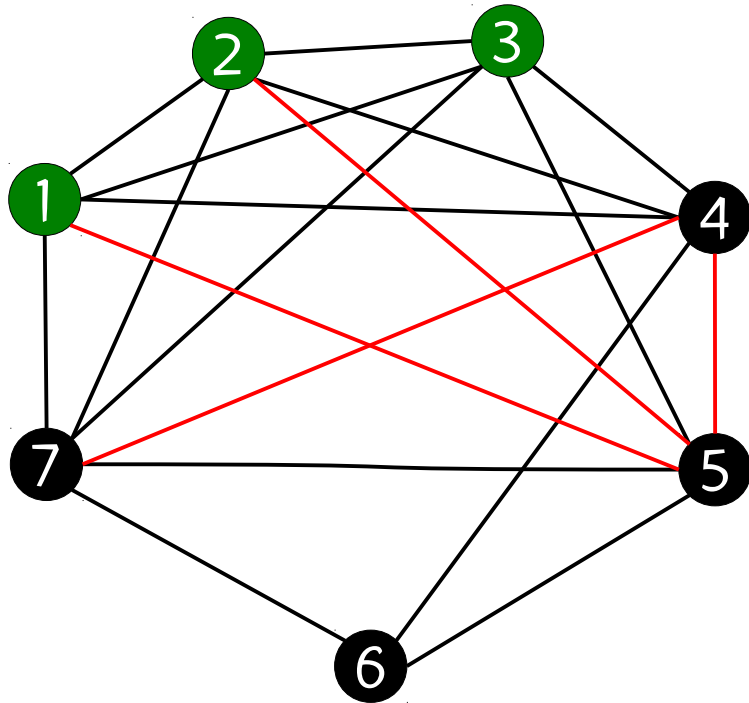
Compute set  $X$  of pairs that do not communicate that would if  $v$  becomes a regenerator

Add  $X$  to  $E'$

end while

return  $R$

# Decoding



$X = (0.1, 0.2, 0.3, 0.4, 0.5, 0.6, 0.7)$

↑  
 $i = 3$

Scan  $V$  in order implied by  $Y$

while some pair in  $V \times V$  cannot  
communicate ( $M = (V, E')$  is not  
complete):

Add next vertex  $v$  in order into  $R$

Compute set  $X$  of pairs that do  
not communicate that would if  
 $v$  becomes a regenerator

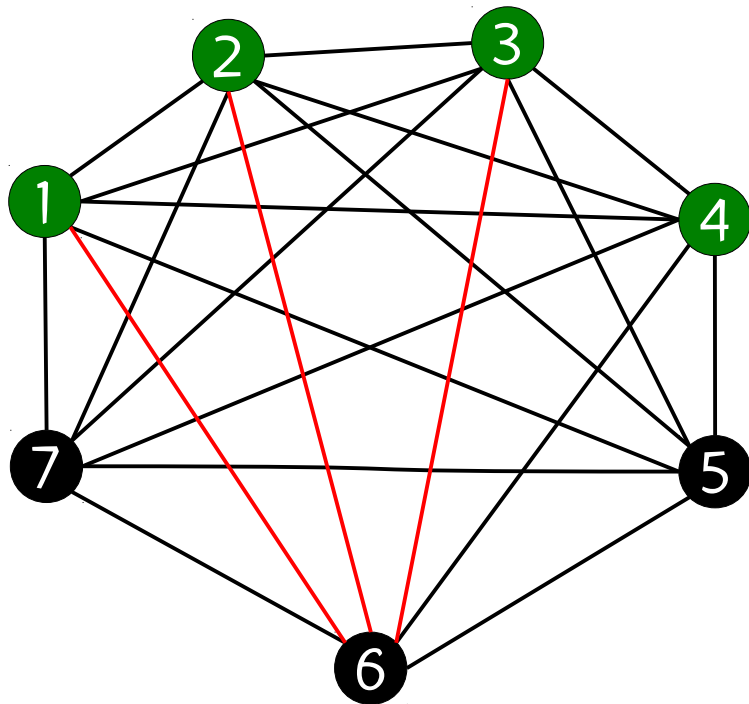
Add  $X$  to  $E'$

end while

return  $R$



# Decoding



$X = (0.1, 0.2, 0.3, 0.4, 0.5, 0.6, 0.7)$

↑  
 $i = 4$

Scan  $V$  in order implied by  $Y$

while some pair in  $V \times V$  cannot communicate ( $M = (V, E')$  is not complete):

Add next vertex  $v$  in order into  $R$

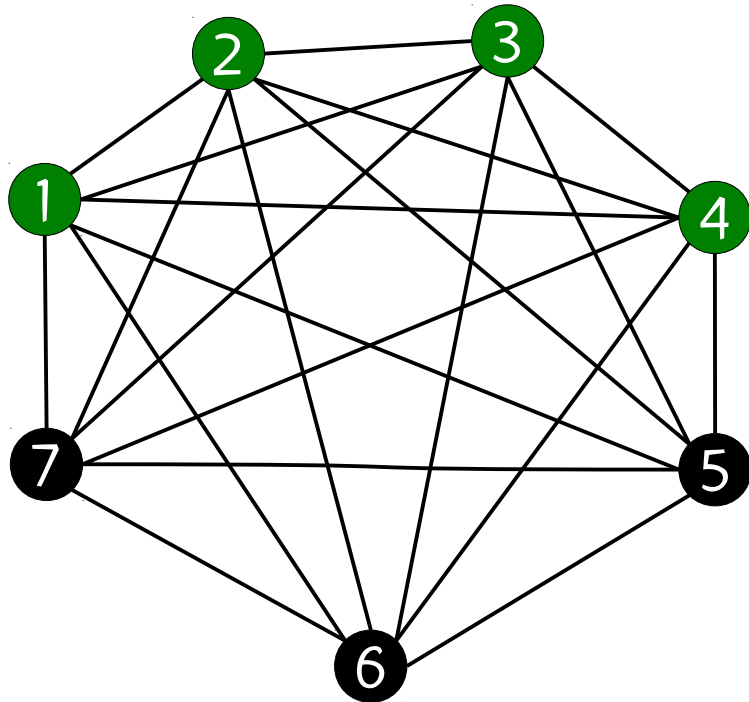
Compute set  $X$  of pairs that do not communicate that would if  $v$  becomes a regenerator

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end while

return  $R$

# Decoding



M is complete!

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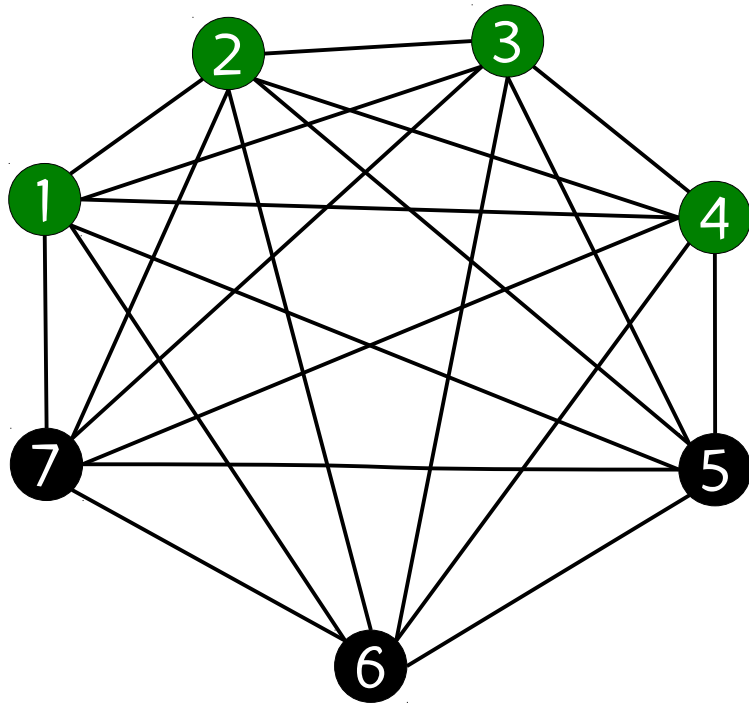
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local search

# Routing and wavelength assignment in optical networks

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- Objective: Route a set of connections (called lightpaths) and assign a wavelength to each of them.

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- If no wavelength converters are available, the same wavelength must be assigned along the entire route (wavelength continuity constraint).



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  - NP-hard (Erlebach and Jansen, 2001)

# Routing and wavelength assignment (RWA)

## Connections

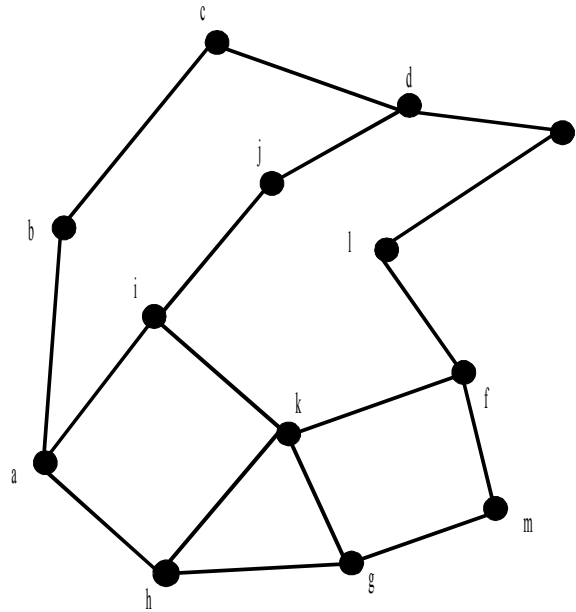
$c \leftrightarrow m$

$d \leftrightarrow b$

$e \leftrightarrow h$

$a \leftrightarrow e$

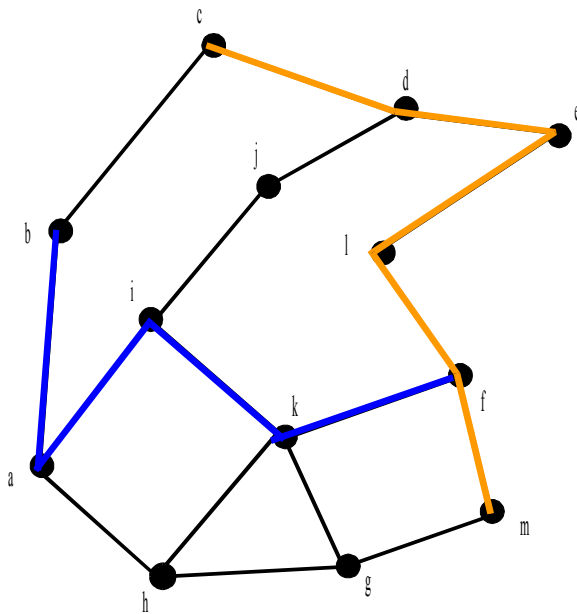
$b \leftrightarrow f$



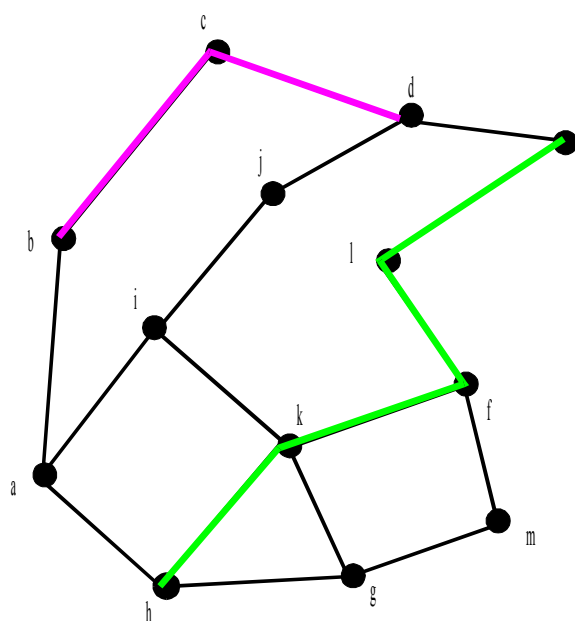


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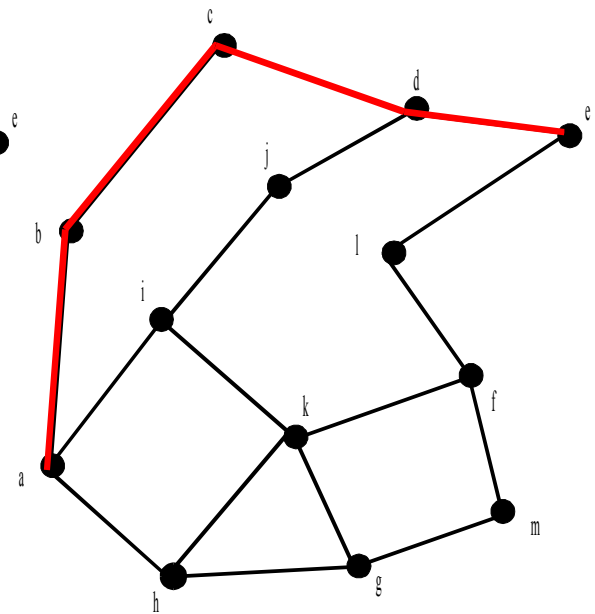
Connections:  $(a \leftrightarrow e)$   $(b \leftrightarrow f)$   $(c \leftrightarrow m)$   $(d \leftrightarrow b)$   $(e \leftrightarrow h)$



wavelength 1



wavelength 2



wavelength 3

# Heuristic of N. Skorin-Kapov (EJOR, 2007)

- Associates the min-RWA with the bin packing problem.
  - Wavelengths are associated with bins.
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  - **Best Fit Decreasing (BFD): state of the art heuristic for RWA**

# Efficient implementation of BFD-RWA



T.F. Noronha, M.G.C.R., and C.C. Ribeiro,

“Efficient implementations of heuristics for routing and wavelength assignment,” in “Experimental Algorithms,”

7th International Workshop (WEA 2008), C.C. McGeoch (Ed.), LNCS, vol. 5038, pp. 169-180, Springer, 2008.

Tech report version:

[http://www.research.att.com/~mgcr/doc/impl\\_rwa\\_heur.pdf](http://www.research.att.com/~mgcr/doc/impl_rwa_heur.pdf)

# BFD-RWA

N. Skorin-Kapov (2007); Noronha, R., and Ribeiro (2008)

- Input:
  - A directed graph  $G$  representing the network topology.
  - A set  $T$  of connection requests.
  - The value  $d$  of the maximum number of arcs in each route. It is set to be the maximum of the square root of the number of links in the network and the diameter of  $G$ .
- Starts with only one copy of  $G$  (called  $G_1$ ).
- Connections are selected according to non-increasing order of the lengths of their shortest paths in  $G_i$ . Ties are broken at random.
- The connection is assigned wavelength  $i$ , and the arcs along path are deleted from  $G_i$ .
- If no existing bin can accommodate the connection with fewer than  $d$  arcs, a new bin is created.

# BRKGA for RWA: GA-RWA



T.F. Noronha, M.G.C.R., and C.C. Ribeiro, “**A biased random-key genetic algorithm for routing and wavelength assignment**,” J. of Global Optimization, vol. 50, pp. 503–518, 2011.

Tech report version:

<http://www.research.att.com/~mgcr/doc/garwa-full.pdf>

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Noronha, R., and Ribeiro (2011)

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  - 1) Sort the connection in set  $T$  in non-increasing order of  $c(i) = SP(i) \times 10 + X[i]$ , for each connection  $i \in T$ .
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# BRKGA for RWA: GA-RWA

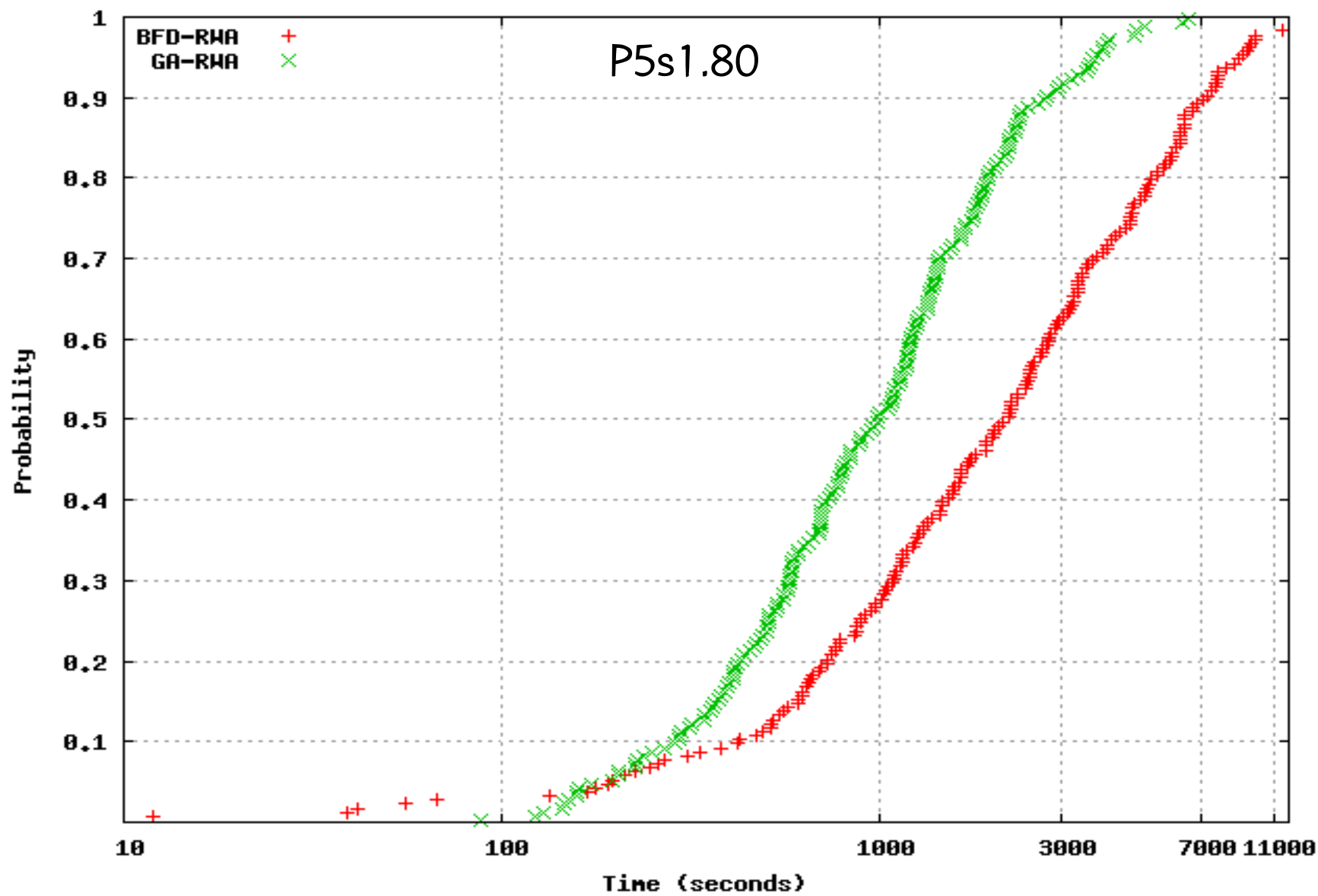
Noronha, R., and Ribeiro (2011)

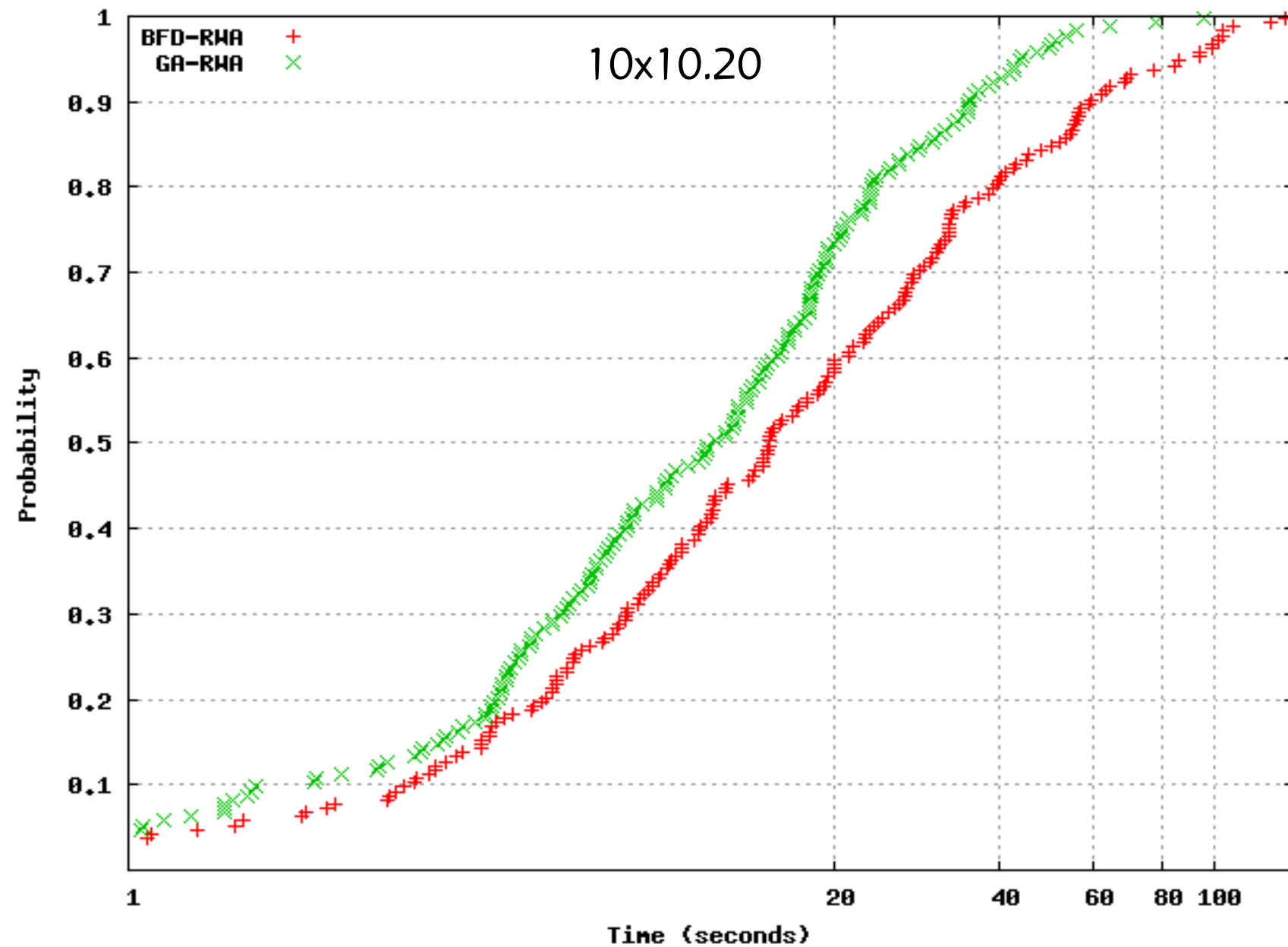
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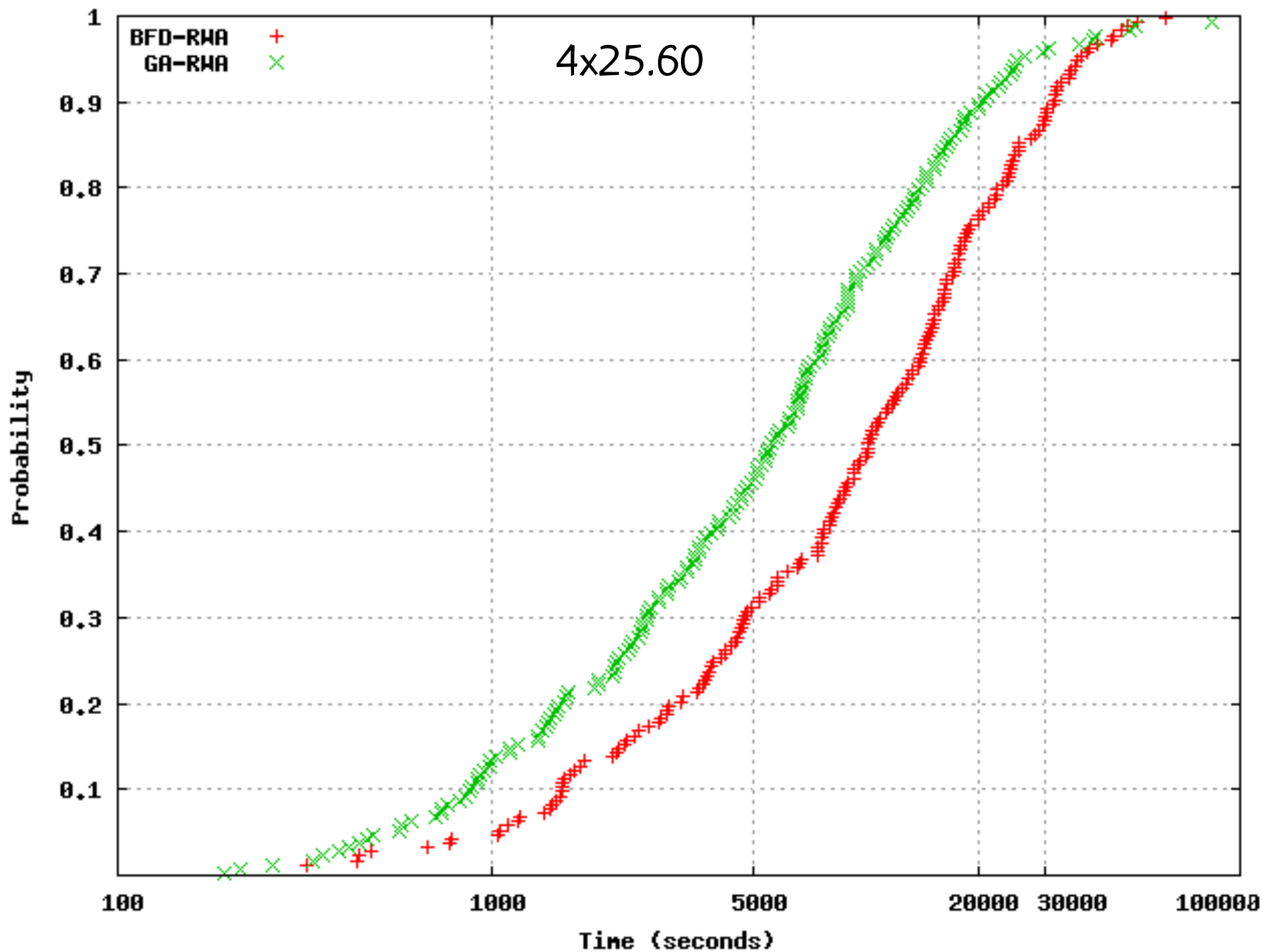
Since there are many ties connection pairs with The same  $SP(i)$  value, in the original algorithm of Skorin-Kapov, ties are broken at random. In the BRKGA, the algorithm “**learns**” how to break ties.

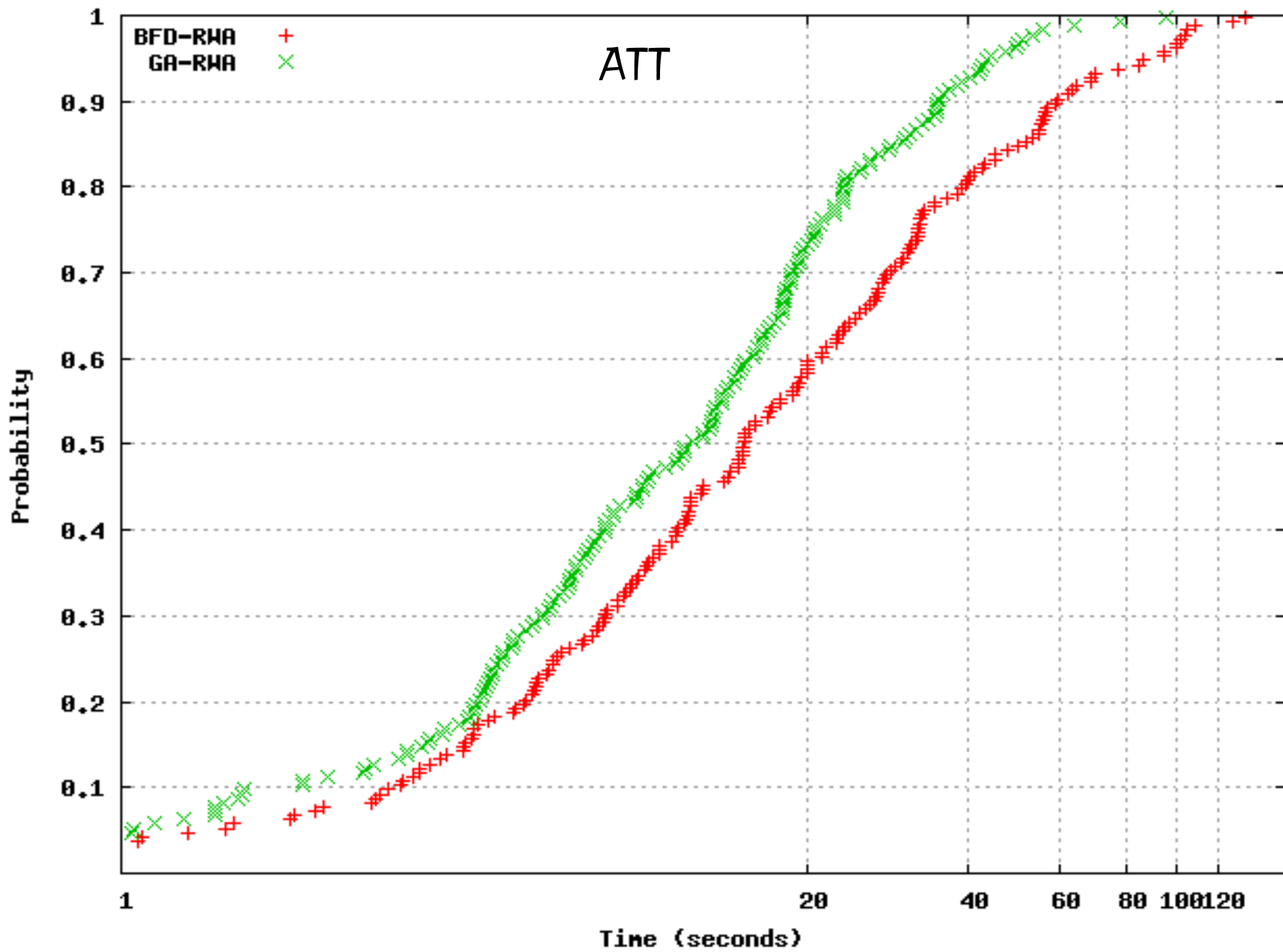
# Experiments

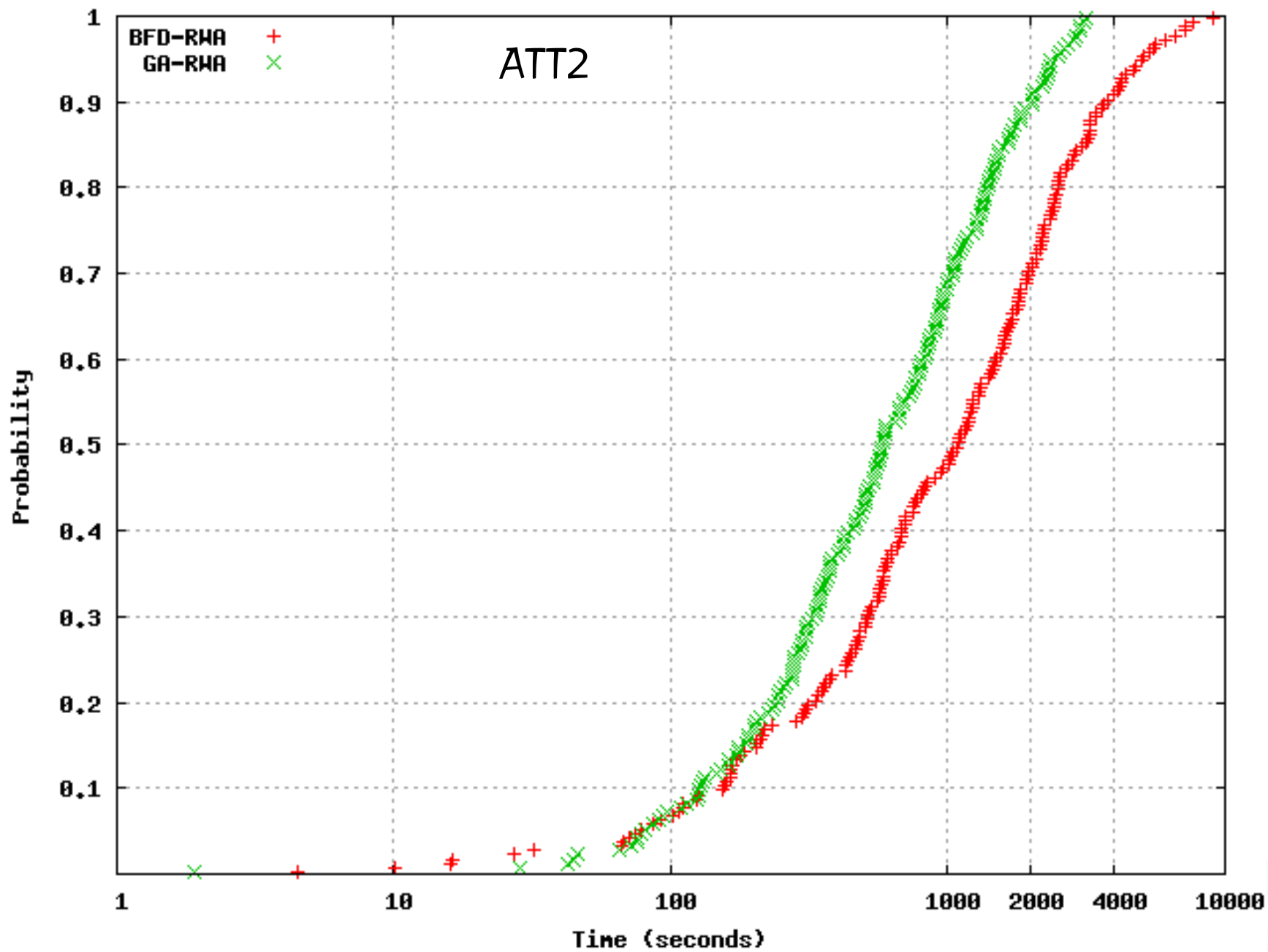
- Compare multi-start version of Skorin-Kapov's heuristic (MS-RWA) with GA-RWA.
- Make 200 independent runs of each heuristic of each heuristic on five instances, stopping when target solution was found (target was set to be best solution found by MS-RWA after 10,000 multi-start iterations).
- Plot CDF (runtime distribution) for each heuristic.













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- BRKGA heuristics are highly parallelizable. Calls to decoder are independent.

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- We have had only a small glimpse at BRKGA applications to problems arising in telecommunications.
- The BRKGAs described in this talk are all state-of-the-art heuristics for these applications
- We are currently working on a number of other applications in telecommunications, including the degree-constrained and the capacitated spanning tree problems and a metropolitan network design problem.



# Thanks!

These slides and all of the papers cited in this talk can be downloaded from my homepage:

<http://www.research.att.com/~mgcr>