



# On the Dynamic Approximate Multicommodity Flow Problem

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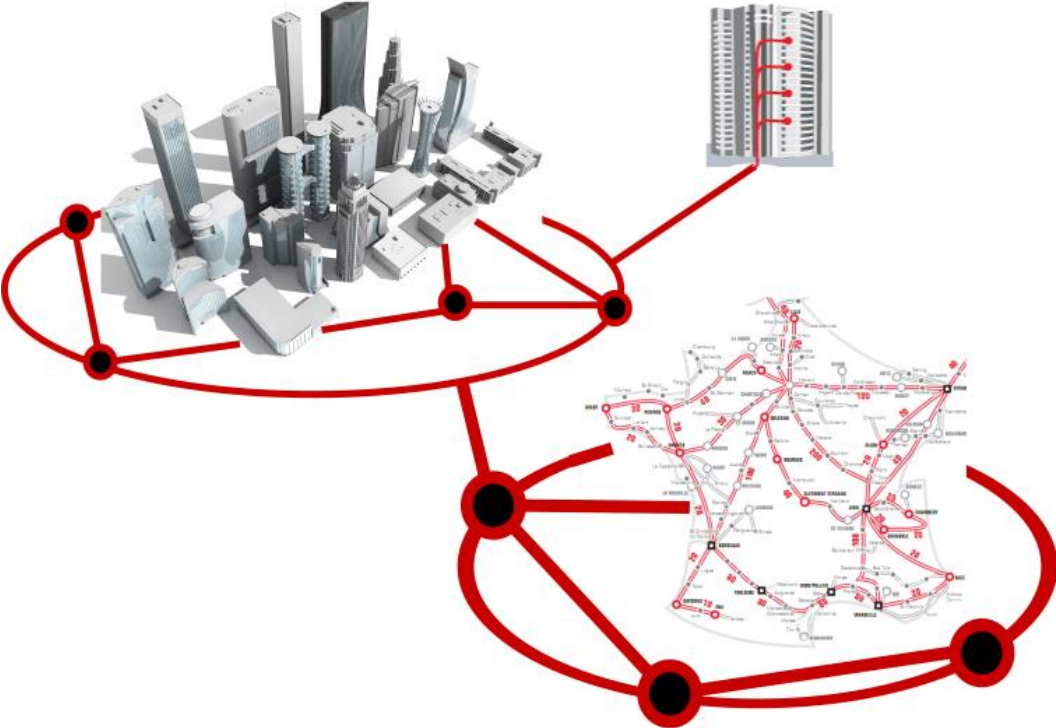


**CHIST-ERA STAR**

# 0

## Context

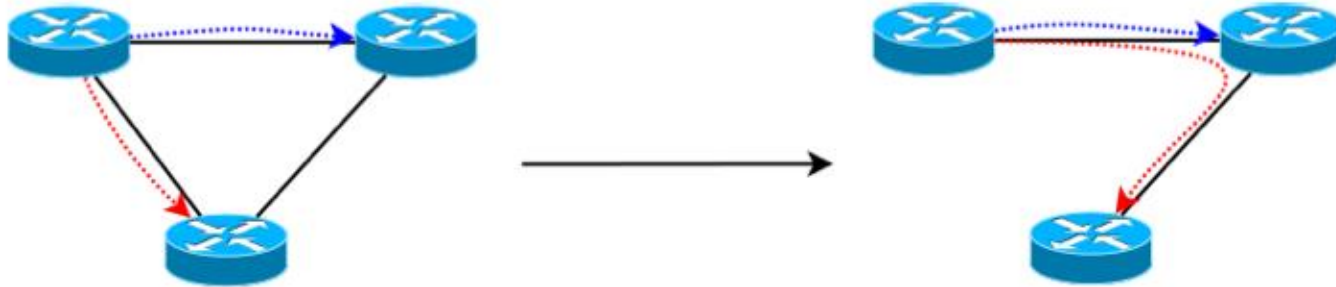
# Energy consumption of computer networks



gwatt.net, 2013



# Turn-off to save energy

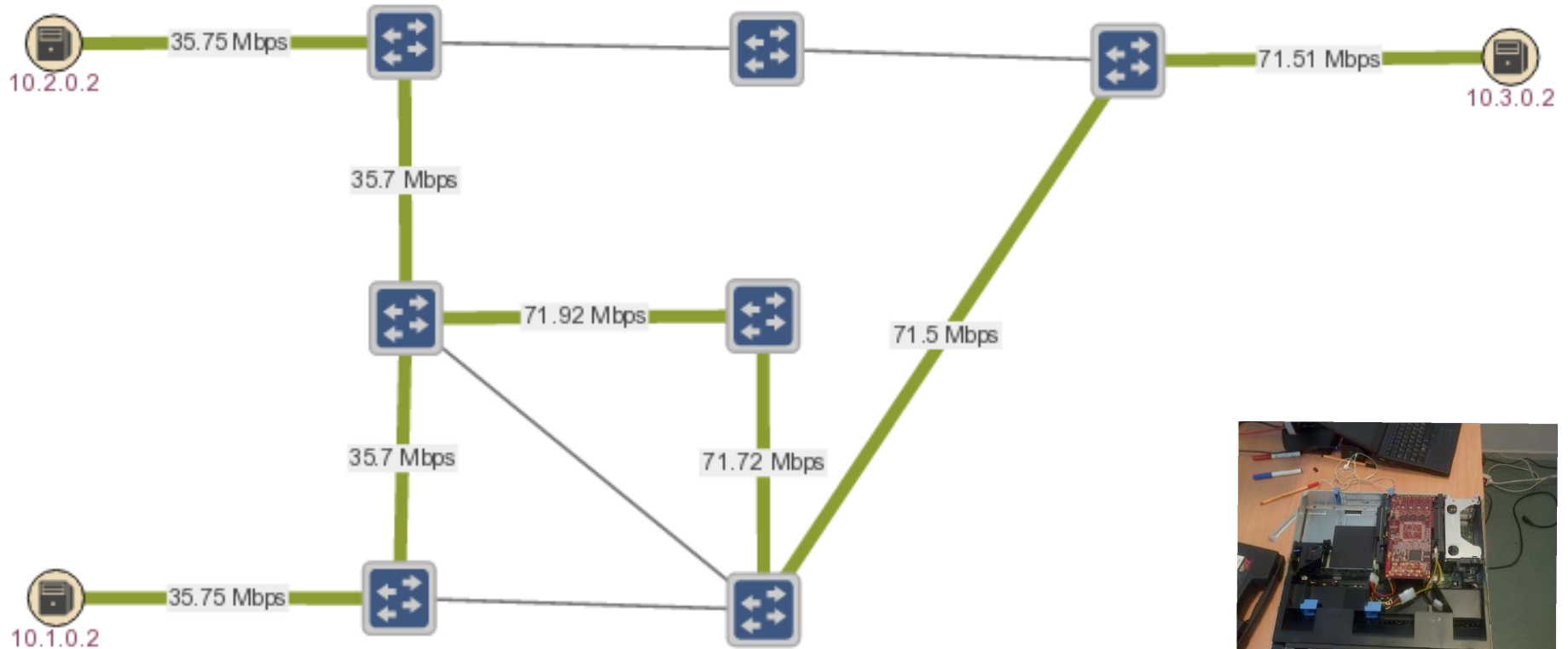


## *Power consumption [Watt]*

|               |          |
|---------------|----------|
| 1 Gbps port   | 7 W      |
| 2.5 Gbps port | 15 W     |
| 10 Gbps port  | 34 W     |
| 40 Gbps port  | 160 W    |
| 100 Gbps port | 360 W    |
| 400 Gbps port | (1236 W) |
| 1 Tbps port   | (2794 W) |

\*Van Heddeghem, Ward, Filip Idzikowski et al.. 2012. "Power Consumption Modeling in Optical Multilayer Networks." Photonic Network Communications 24 (2): 86–102

# Demo



# Outline

1. Single and multi-commodity flow problems
2. Approximations to the maximum concurrent flow problem
  1. Garg-Konemann framework
  2. Optimizations with dynamic graph algorithms
3. Future work

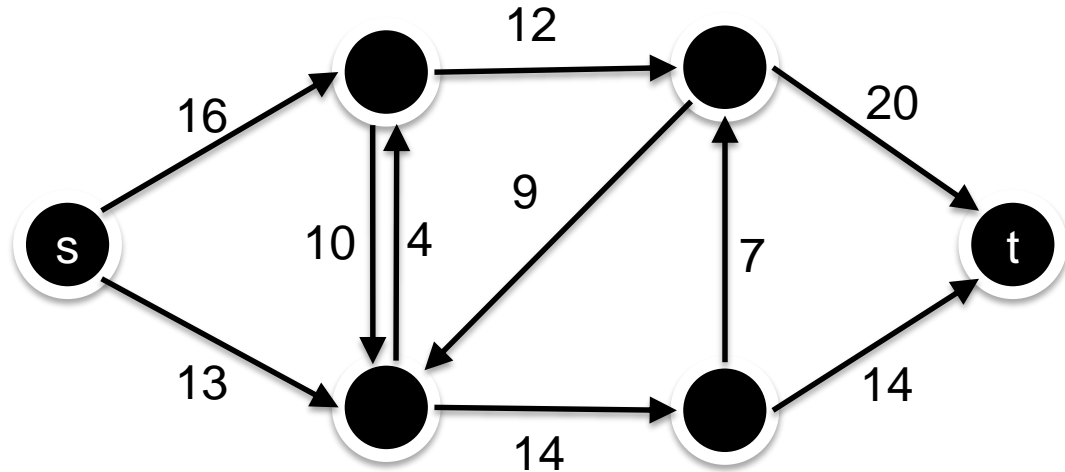
G. Karakostas, Faster approximation schemes for fractional multicommodity flow problems, ACM Trans. Algorithms 4 (2008) 1V17.

# 1

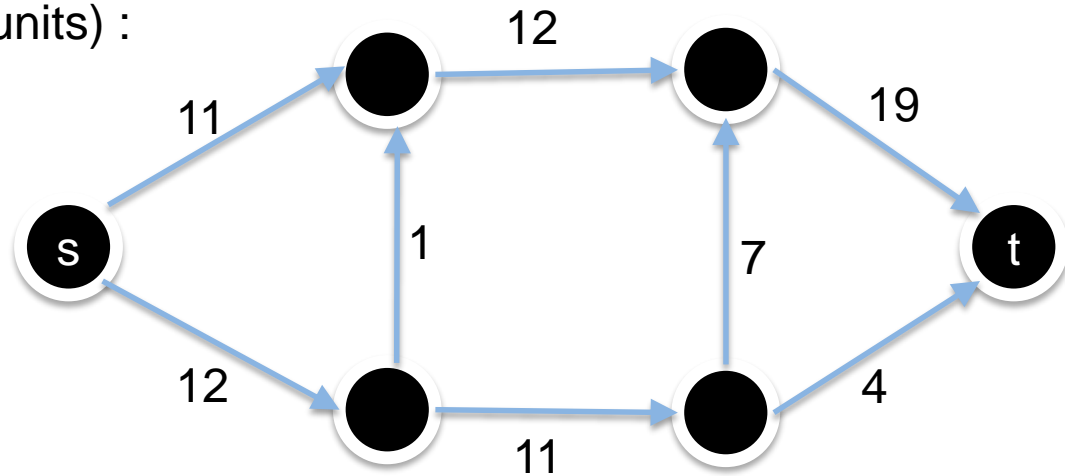
## Single and multi-commodity flow problems

# Maximum (single-commodity) flow

Capacities  $u(e), e \in E$ :



Maximum flow (of 23 total units) :

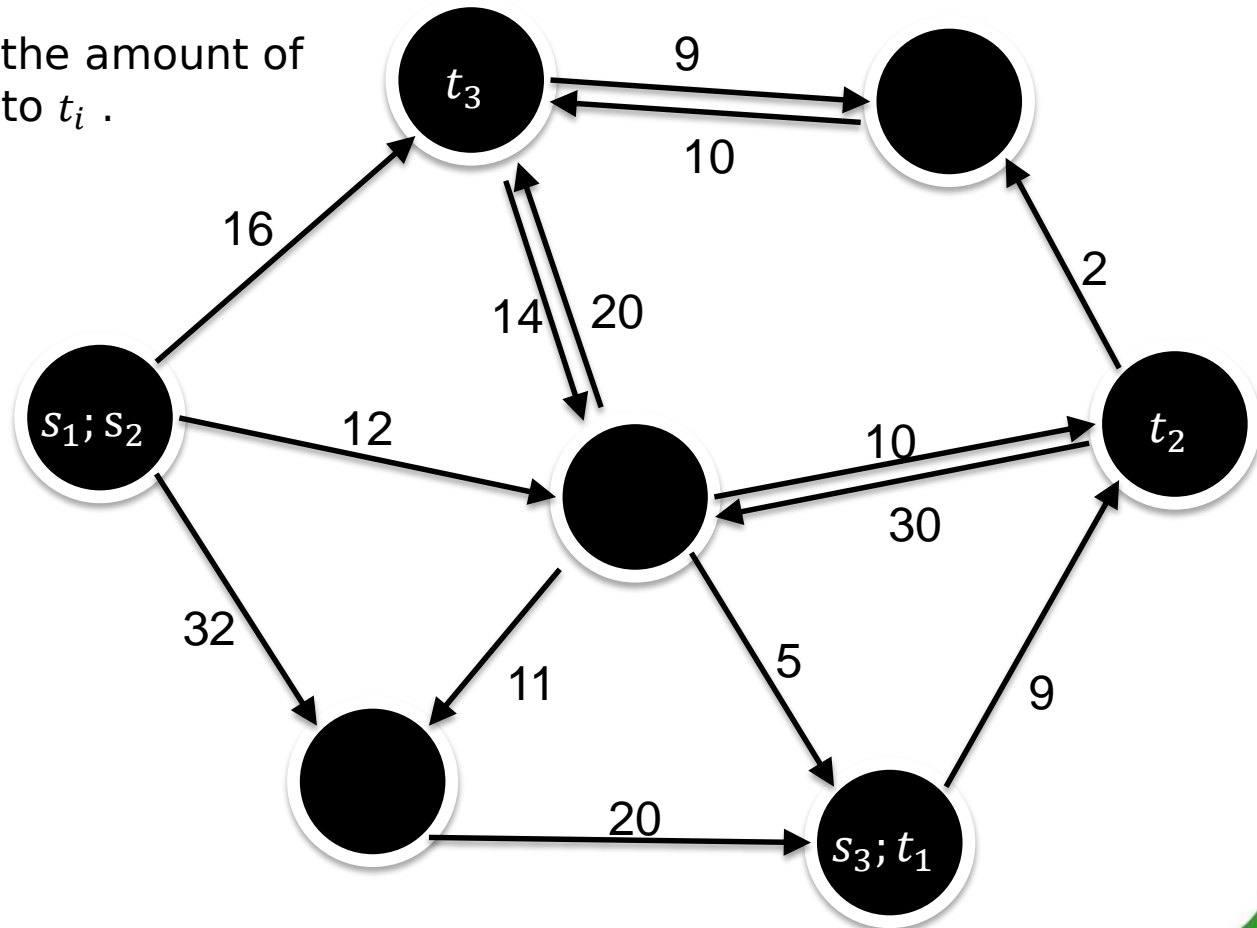




# Maximum multi-commodity flow

$k$  commodities  $K_1, \dots, K_k$  defined by  $K_i = (s_i, t_i)$

maximize  $\sum_i f_i$ , where  $f_i$  is the amount of commodity routed from  $s_i$  to  $t_i$ .

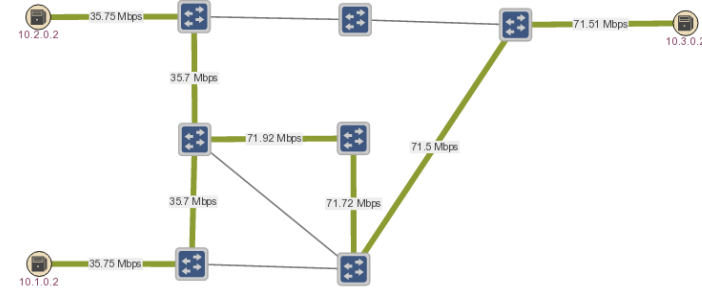


# 2

## Approximations to the maximum concurrent flow problem

# Maximum concurrent flow

Multi-commodity flow +  $k$  positive demands  $d_1, \dots, d_k$



Find the maximum constant  $\lambda$ , such that:

$\forall i, \lambda d_i$  units of commodity  $K_i$  are routed between  $s_i$  and  $t_i$

Garg-Konemann framework:

approximation based on the dual of the linear definition

minimize  $D(l) = \sum_e (l(e)u(e))$

# Garg-Konemann framework

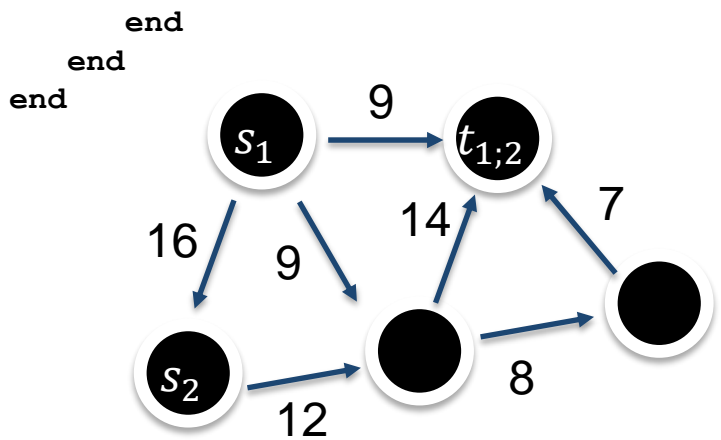
**Input:** Graph  $G = (V, E)$   
 capacities  $u(e)$ ,  
 commodity pairs  $\{(s_i, t_i)\}_i$  with demands  $d_i > 0$ ,  
 accuracy parameter  $\epsilon > 0$   
**Output:** (Infeasible) flow  $f$

Initialize  $f \leftarrow \emptyset$ ,  $l(e) \leftarrow \frac{\gamma}{u(e)}$  for all arcs  $e \in E$ , where  $\gamma = \left(\frac{m}{1-\epsilon}\right)^{-\frac{1}{\epsilon}}$

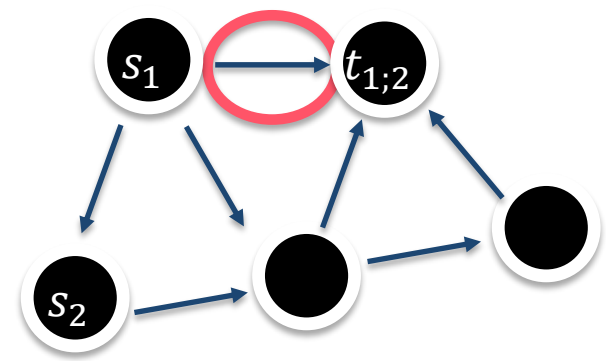
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while  $D(l) < 1$  do
  for  $i := 1, \dots, k$  do
     $d'_i \leftarrow d_i$ 
    while  $D(l) < 1$  and  $d'_i > 0$  do
      Find the shortest path  $p$  in  $P_i$ 
      Find the bottleneck capacity  $u$  of  $p$ 
       $d'_i \leftarrow d'_i - u$ 
      Augment the flow  $f$  by routing  $u$  units of flow along the path  $p$ 
      foreach arc  $e$  in  $p$  do  $l(e) \leftarrow l(e) \cdot \left(1 + \frac{\epsilon \cdot u}{u(e)}\right)$ 
    end
  end
end
    
```

$P_1 = \{\text{all paths from } s_1 \text{ to } t_1\}$   
 $( * u \leftarrow \min\{d'_i, \min_{e \in p} u(e)\} *)$   $u \leftarrow d'_1 = d_1 = 3$   
**For  $\epsilon = 0.05$ ,  $l(e) = l(e) \cdot 1.016$**



$d_1 = 3$   
 $d_2 = 2$



# Garg-Konemann framework

## key observations

The shown algorithm finishes after at most  $t := 1 + \frac{\lambda}{\epsilon} \log_{1+\epsilon} \frac{m}{1-\epsilon}$  phases

The obtained solution must be scaled down by  $\log_{1+\epsilon} \frac{1}{\gamma}$

If  $\lambda > 1$ , the scaled down flow has a value of at least  $(1 - 3\epsilon) \lambda$

m: number of arcs

# Optimization with dynamic graph algorithms

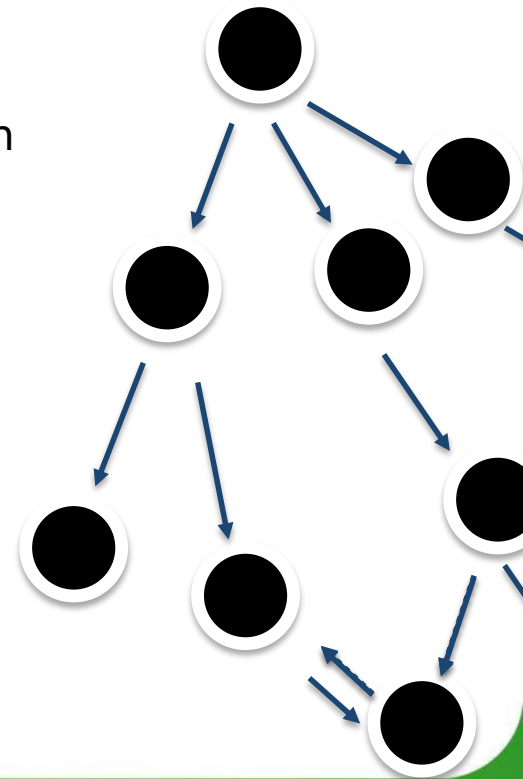
## Solution 1

☹ Shortest path computation takes time

Use dynamic all-pairs shortest paths algorithms :  
Partially update the structure when arc weight changes

☹ Still costly updates

☹ Worst case update cost equal to re-computing from scratch



\*Camil Demetrescu and Giuseppe F. Italiano. Experimental analysis of dynamic all pairs shortest path algorithms. ACM Trans. Algorithms,2(4):578–601, October 2006.

# Optimization with dynamic graph algorithms

## Solution 2

Construct a data structure allowing a fixed  $O(1)$  cost per increase of the length of any arc.

based on probabilistic graph sparsification

probabilistic result : probability of  $n^{-5}$  to have a sup-optimal result

Final complexity :  $O((m + k)n\epsilon^{-2} \log M)$

# 3

## Future work

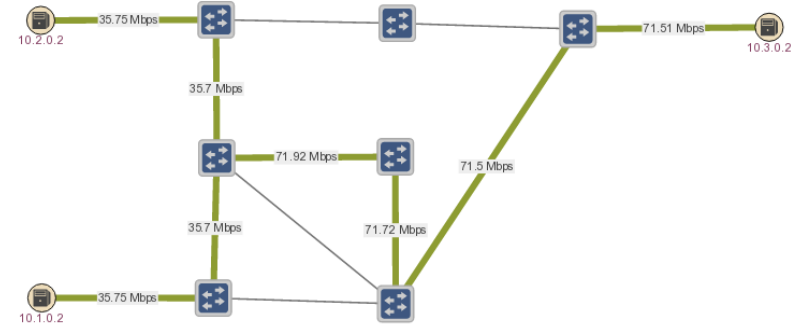


# Ideas for future work

To better save energy in computer networks:

Networks are not static, demands permanently change

A fast reaction to network changes is needed



- ☺ Update  $\lambda$  without re-computing the solution from scratch
- ☺ Probabilistic and approximate result is enough if it is fast
- ☺ Computational time limit and getting  $\epsilon$  as output
- ☺ Fast non-fractional commodity placement ? ☹

Thank you

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