JOINT ROUTING AND PLACEMENT OF VIRTUAL NETWORK FUNCTIONS

Jorge Crichigno^{1,2}, D. Oliveira³, M. Pourvali³, N. Ghani³, D. Torres²

¹University of South Carolina, SC, USA

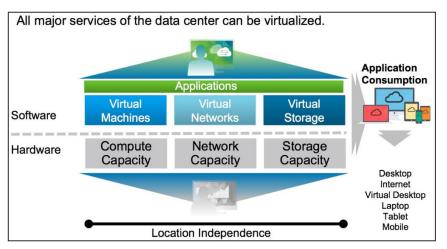
²Northern New Mexico College, NM, USA

³University of South Florida, FL, USA

Agenda

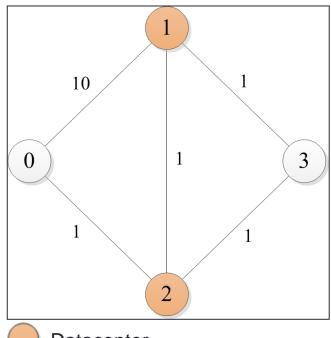
- Introduction
- Optimization model
- Numerical examples
- Concluding remarks

- Network Function Virtualization (NFV) is a technology that permits the implementation of Network Functions (NFs) on datacenters' commodity servers
- Network functions include
 - Firewall, access control lists
 - Routers, switches, NAT, DHCP



http://www.vmware.com/

- Consider the weighted network below
- A set of datacenter that implement particular functions
- There is a set of function $F = \{0, 1\}$
- A client request is interested in both functions to apply them to a flow from ingress switch 0 to egress switch 3
- A datacenter *d* implements $F_d \subseteq F$
- The cost and resources to implement a function are datacenter-dependent
- What should the path of the flow be, in order to minimize the routing and deployment costs?

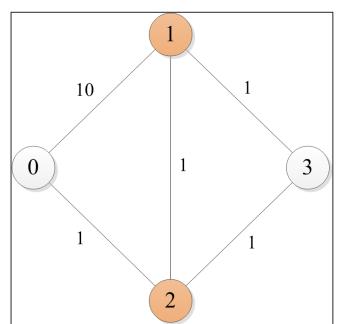


Datacenter

Example:

- Datacenter 1 implements functions 0 and 1 at costs 1 and 10
- Datacenter 2 implements functions 0 and 1 at costs 10 and 1

Function 0 at cost 1 Function 1 at cost 10

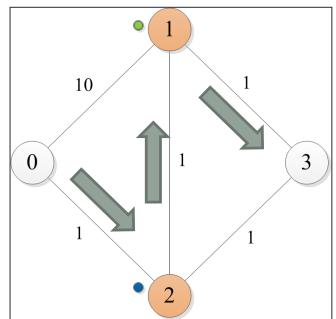


Function 0 at cost 10 Function 1 at cost 1

Example:

- Datacenter 1 implements functions 0 and 1 at costs 1 and 10
- Datacenter 2 implements functions 0 and 1 at costs 10 and 1
- The optimal solution places functions 0 and 1 at datacenters 1 and 2 respectively, and route the traffic through (0, 2), (2, 1), (1,3)

Function 0 at cost 1 Function 1 at cost 10



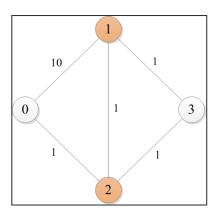
Function 0 at cost 10 Function 1 at cost 1

- Function 0 instance
- Function 1 instance

- The network is represented as a graph G = (V, E)
- Each link $(i, j) \in E$ has an associated cost c^{ij}
- The subset $D \subseteq V$ represents the set of datacenters
- A datacenter $d \in D$ implements a subset of functions $F_d \subseteq F$
- Each request $r \in R$ is characterized by a 3-tuple (src_r, dst_r, F_r)
- A datacenter has a set of resources $W = \{w_{d,1}, w_{d,2}, ..., w_{d,m}\}$
- To implement function $i \in F_d$, the datacenter uses $w^i_{d,1}, w^i_{d,2}, ..., w^i_{d,m}$
- The setup cost of an instance $i \in F_d$ is c_d^i
- Each instance $i \in F_d$ can serve up to λ_d^i requests
- Variable $x_{r,d}^i$ indicates whether datacenter d serves function $i \in F_r$ requested by $r \in R$
- Variable y_d^i indicates the number of instances of function i at d
- Variable l_r^{ij} indicates whether link $(i,j) \in E$ is used by flow $r \in R$

 The objective is the maximization of the number of satisfied network functions (NFs)

$$\operatorname{Max} F_1 = \sum_{r \in R} \sum_{i \in F_d} \sum_{d \in D} x_{r,d}^i = x_{0,1}^0 + x_{0,1}^1 + x_{0,2}^0 + x_{0,2}^1$$

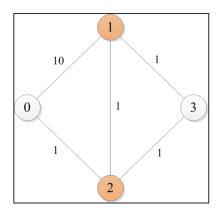


 The objective is the maximization of the number of satisfied network functions (NFs)

$$\operatorname{Max} F_1 = \sum_{r \in R} \sum_{i \in F_d} \sum_{d \in D} x_{r,d}^i = x_{0,1}^0 + x_{0,1}^1 + x_{0,2}^0 + x_{0,2}^1$$

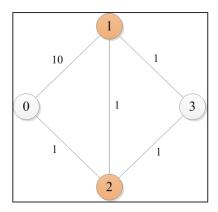
Minimization of the NF deployment cost

$$\operatorname{Max} - F_2 = \sum_{d \in D} \sum_{i \in F_d} c_d^i y_d^i = y_1^0 + 10y_1^1 + 10y_2^0 + 1y_2^1$$



Minimization of the routing cost

$$\begin{aligned} \text{Max - F3} &= \sum_{r \in R} \sum_{(i,j) \in E} c^{ij} l_r^{(i,j)} = \\ & 10 l_0^{(0,1)} + l_0^{(0,2)} + 10 l_0^{(1,0)} + l_0^{(1,2)} + l_0^{(1,3)} + l_0^{(2,0)} + l_0^{(2,1)} + l_0^{(2,3)} + l_0^{(3,1)} + l_0^{(3,2)} \end{aligned}$$

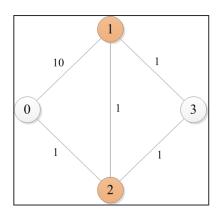


Requested functions 0 and 1 are only implemented in one datacenter

$$\sum_{d \in D} x_{r,d}^i \leq 1 \qquad \qquad \text{Function 0}$$

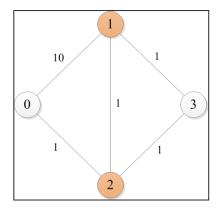
$$x_{0,1}^0 + x_{0,2}^0 \leq 1 \qquad \qquad \text{Function 0}$$

$$x_{0,1}^1 + x_{0,2}^1 \leq 1 \qquad \qquad \text{Function 1}$$



- The total amount of resources (memory, CPU, storage) is limited at each datacenter
- E.g., 15 and 20 storage units used by an instance of function 0 and 1 respectively at datacenter 1. Datacenter has 100 storage units

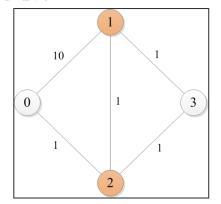
$$\sum_{i \in F} w_{d,j}^i y_d^i \leq w_{d,j} \qquad \qquad 15y_1^0 + 20y_1^1 \leq 100 \qquad \text{Datacenter 1,} \\ \text{storage resource}$$



There is a path from the ingress switch 0 to egress switch 3

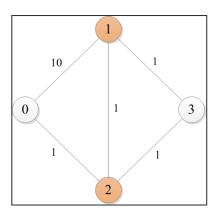
Node 0:
$$\left(l_0^{(0,1)} + l_0^{(0,2)}\right) - \left(l_0^{(1,0)} + l_0^{(2,0)}\right) \ = \ 1$$
 Node 1:
$$\left(l_0^{(1,0)} + l_0^{(1,2)} + l_0^{(1,3)}\right) - \left(l_0^{(0,1)} + l_0^{(2,1)} + l_0^{(3,1)}\right) \ = \ 0$$
 Node 2:
$$\left(l_0^{(2,0)} + l_0^{(2,1)} + l_0^{(2,3)}\right) - \left(l_0^{(0,2)} + l_0^{(1,2)} + l_0^{(3,2)}\right) \ = \ 0$$
 Node 3:
$$\left(l_0^{(3,1)} + l_0^{(3,2)}\right) - \left(l_0^{(1,3)} + l_0^{(2,3)}\right) \ = \ -1$$

$$\sum_{j:(i,j)\in E} l_{ij}^r \ - \ \sum_{j:(j,i)\in E} l_{ji}^r = \left\{ \begin{array}{l} -1; i = dst_r, \ src_r \neq dst_r \\ 1; i = src_r, \ src_r \neq dst_r \\ 0; \ \text{otherwise}. \end{array} \right.$$



 If a function 0 is placed at datacenter 1, then the path from the ingress switch 0 to egress switch 3 must include datacenter 1

$$l_0^{(1,0)} + l_0^{(1,2)} + l_0^{(1,3)} \ge x_{0,1}^0$$

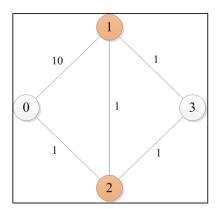


 If a function 0 is placed at datacenter 1, then the path from the ingress switch 0 to egress switch 3 must include datacenter 1

$$l_0^{(1,0)} + l_0^{(1,2)} + l_0^{(1,3)} \ge x_{0,1}^0$$

 If a function 1 is placed at datacenter 1, then the path from the ingress switch 0 to egress switch 3 must include datacenter 1

$$l_0^{(1,0)} + l_0^{(1,2)} + l_0^{(1,3)} \ge x_{0,1}^1$$



 If a function 0 is placed at datacenter 1, then the path from the ingress switch 0 to egress switch 3 must include datacenter 1

$$l_0^{(1,0)} + l_0^{(1,2)} + l_0^{(1,3)} \ge x_{0,1}^0$$

 If a function 1 is placed at datacenter 1, then the path from the ingress switch 0 to egress switch 3 must include datacenter 1

$$l_0^{(1,0)} + l_0^{(1,2)} + l_0^{(1,3)} \ge x_{0,1}^1$$

$$\sum_{(d,j)\in E} l_r^{dj} \ge x_{r,d}^i$$

• Variables $x_{r,d}^i$, y_d^i , l_{ij}^r are binary, integer, and real — NP hard

For large instances of the problem, finding the optimal solution is

not practical

$$\max F = w_1 \sum_{r \in R} \sum_{i \in F_r} \sum_{d \in D | i \in F_d} x_{r,d}^i - w_2 \sum_{d \in D} \sum_{i \in F_d} c_d^i y_d^i$$

$$-w_3 \sum_{r \in R} \sum_{(i,j) \in E} c^{ij} l_r^{ij} \qquad (1)$$

$$\sum_{d \in D} x_{r,d}^i \leq 1 \qquad r \in R, i \in F_r \qquad (2)$$

$$x_{r,d}^i \leq y_d^i \qquad r \in R, i \in F_r, d \in D | i \in F_d \qquad (3)$$

$$\sum_{i \in F_d} w_{d,j}^i y_d^i \leq w_{d,j} \qquad d \in D, r \in R, j \in \{1, 2, ..., |W_d|\} \qquad (4)$$

$$\sum_{r \in R} x_{r,d}^i \leq \lambda_d^i y_d^i \qquad d \in D, i \in F_d \qquad (5)$$

$$\sum_{j:(i,j) \in E} l_{ij}^r - \sum_{j:(j,i) \in E} l_{ji}^r = \begin{cases} -1; i = dst_r, src_r \neq dst_r \\ 1; i = src_r, src_r \neq dst_r \\ 0; \text{ otherwise.} \qquad i \in V, r \in R \end{cases} \qquad (6)$$

$$\sum_{(d,j) \in E} l_r^{dj} \geq x_{r,d}^i \qquad r \in R, i \in F_r, d \in D | i \in F_d \qquad (7)$$

$$x_{r,d}^i \in \{0,1\} \qquad r \in R, i \in F_r, d \in D | i \in F_d \qquad (8)$$

$$y_d^i \in Z^+ \qquad d \in D, i \in F_d \qquad (9)$$

$$l_r^{ij} \in \{0,1\} \qquad r \in R, (i,j) \in E \qquad (10)$$

Greedy Approach

Greedy approach based on Dijkstra algorithm

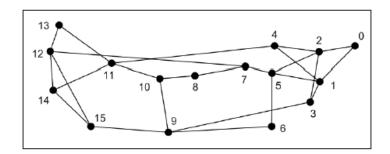
Algorithm 1 Greedy Routing and Placement of NFs

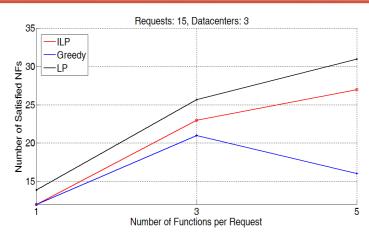
```
1. INPUT: G(V, E), c^{ij} \forall (i, j) \in E, R, F, D
 2. OUTPUT: x_{r,d}^i, y_d^i, l_r^{ij} values
 3. set x_{r,d}^i = 0, y_d^i = 0, l_r^{ij} = 0 for all r \in R, i \in F_r, d \in D, (i,j) \in E
 4. for all r \in R do
       D(r) = \{\}
       k = 1
       for all i \in F_r do
           d_k = datacenter that implements i at minimum cost and has enough resources
           to serve an additional request
           update resources of d_k
           update ydi
10.
           set x_{r,d_k}^i = 1
11.
           D(r) = D(r) \cup d_k
12.
13.
           k = k + 1
       end for
15. end for
16. for all r \in R do
        src = src_r
        C(r) = \{src\}
       for k = 1 to |D(r)| do
20.
           dst = d_k
21.
           if d_k \ni C(r) then
               SP = Dijkstra(src, dst)
               set l_n^{ij} = 1 for all link (i, j) \in SP
24.
               C(r) = C(r) \cup d_k
25.
               C(r) = C(r) \cup j, for all datacenter j \in SP, j \in D(r)
26.
           end if
27.
           src = dst
       end for
       dst = dst_r
       SP = Dijkstra(src, dst)
       set l_*^{ij} = 1 for all (i, j) \in SP
32. end for
33. return x_{r,d}^i, y_d^i, l_r^{ij}
```

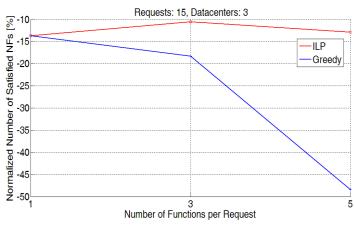
Placement of network functions, one request at a time

Routing of flows through datacenters implementing the functions, one request at a time

- The number of types of resources at a datacenter was set to three (e.g., RAM, storage, CPU)
- The amount of resources of a type at a datacenter is uniform in [.33, 300]
- There are five network functions; each datacenter implements three functions
- The amount of resources of a type needed for an instance of a function is uniform in [0,100]
- The cost of instantiate a function is uniform in [0, 100]
- Datacenters were randomly located in the topology below



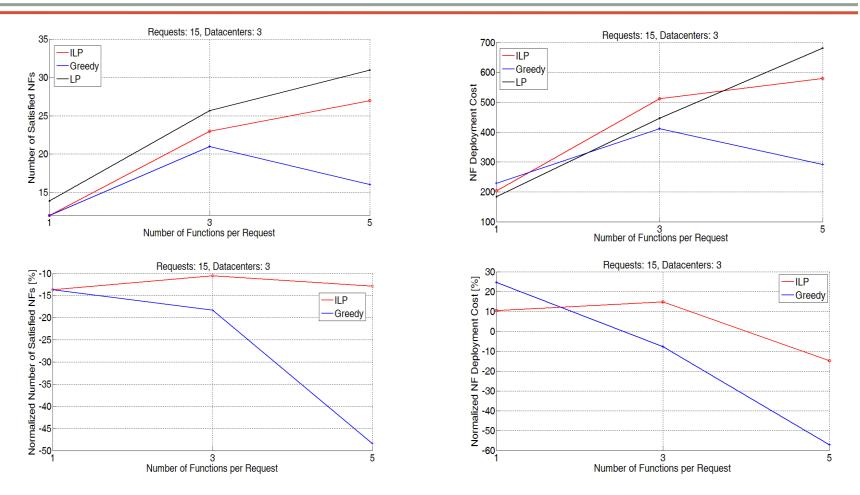




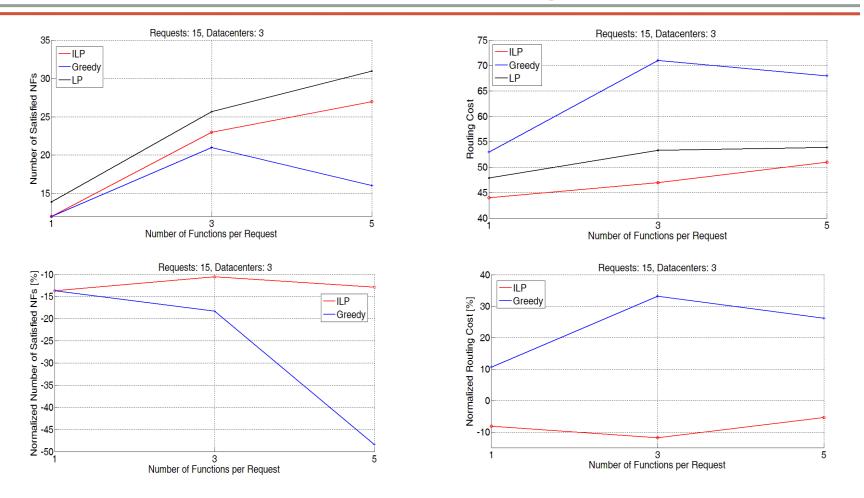
$$Gap = \frac{ov_{LP} - ov_{alg}}{ov_{LP}}$$

where ov_{LP} is the optimal value obtained with the LP scheme, and ov_{alg} is the optimal value obtained with the ILP or greedy heuristic.

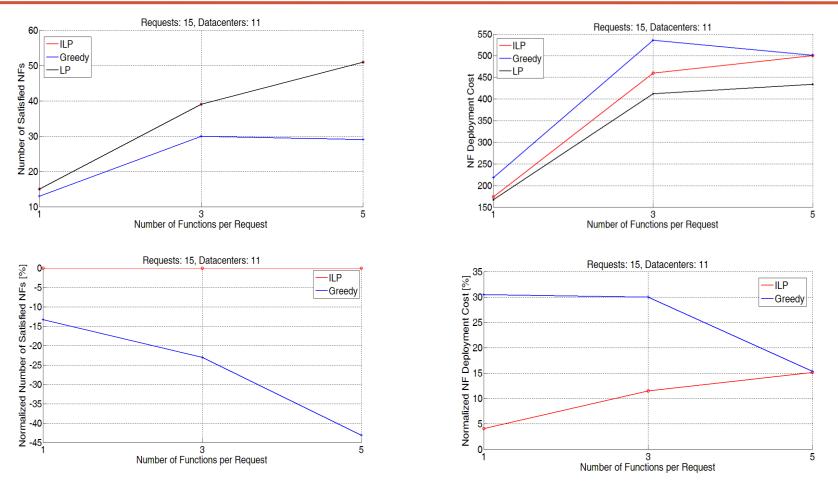
- When there is a small number of datacenters (3) and multiple requests (15), ILP has a comparable performance to that of LP; deployment cost increases with the number of function per request
- The gap of the heuristic increases with the number of function per requests; finding the optimal solution requires the evaluation of a large number of combinations



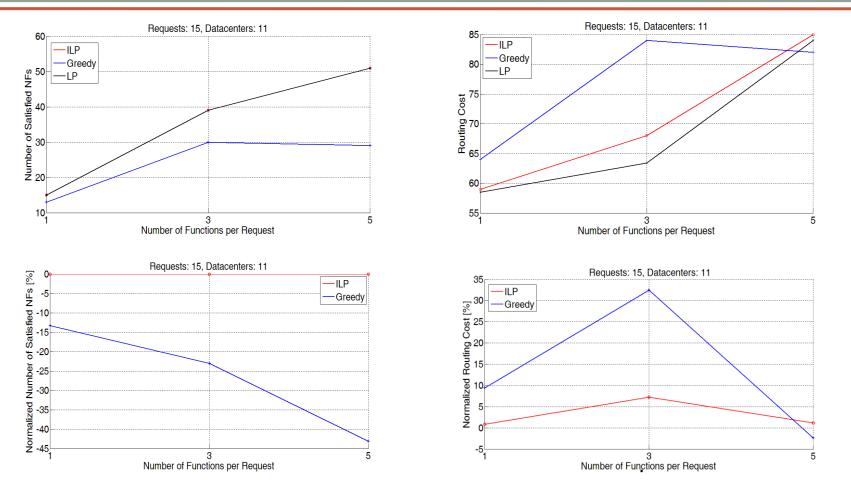
Deployment cost increases with the number of functions per request



 For LP and ILP, the increase in routing cost is mostly flat; i.e., when the number of datacenters is small, routing is 'less important', because the implementation of functions are concentrated in few datacenters



- When there is a large number of datacenters (11) and multiple requests (15), ILP continues to have a comparable performance to LP
- Deployment cost increases substantially when the number of functions per request increases from 1 to 3. However, the increase in cost is minimal when the number of functions per request increases from 3 to 5; i.e., a single instance serves multiple requests without an increase of deployment of functions

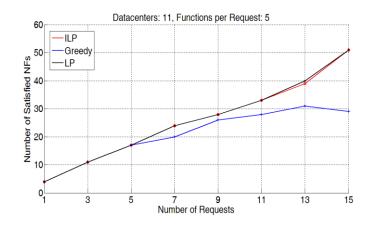


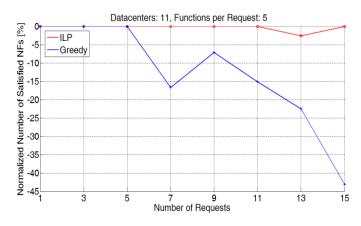
 For LP and ILP, the routing cost increases with the number of function per requests; i.e., when the number of datacenters is large, routing is 'more important', because the implementation of functions are dispersed in many datacenters

Concluding Remarks

- We are currently working on an optimization scheme for the joint routing and placement of virtual network functions (NFs) problem
- The proposed ILP maximizes the number of satisfied NFs while at the same time minimizes the deployment and routing costs
- A heuristics and ILP are currently being tested
- The implementation of the proposed schemes in small testbeds using ONOS SDN is being implemented

THANK YOU

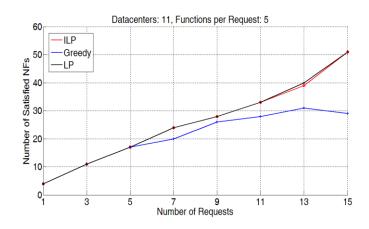


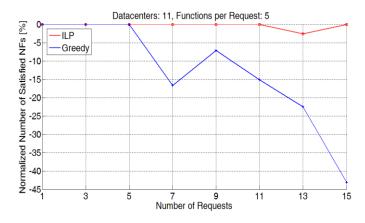


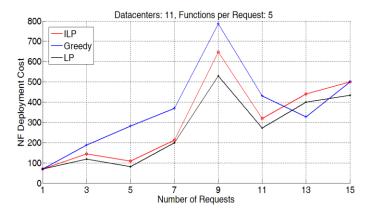
$$Gap = \frac{ov_{LP} - ov_{alg}}{ov_{LP}}$$

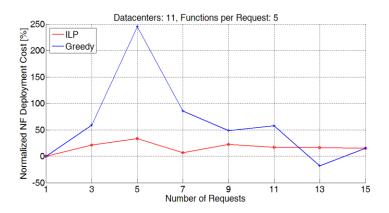
where ov_{LP} is the optimal value obtained with the LP scheme, and ov_{alg} is the optimal value obtained with the ILP or greedy heuristic.

- ILP and LP performances similar; ~2% gap
- As the number of request increases, the heuristic gap substantially increases; finding the optimal solution requires the evaluation of a large number of combinations

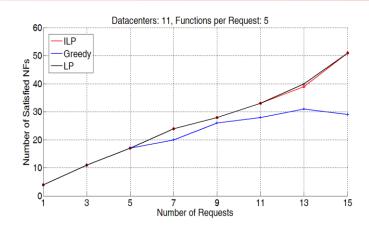


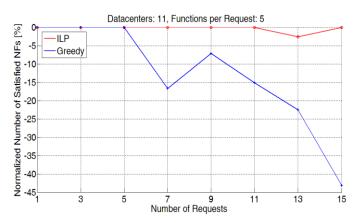


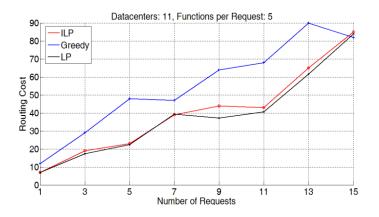




Deployment cost increases with the number of requests; ILP performance is comparable to that of LP –
'small' performance gap









- Routing cost increases with the number of requests; ILP performance is comparable to that of LP –'small' performance gap
- While the gap of the routing cost of the greedy approach decreases with the number of requests, the number of satisfied requests is mostly flat

About Linear Programming

- Many of the problems for which we want algorithms are optimization tasks
- Optimization tasks seek a solution that (1) satisfies certain constraints and (2) is the best, with respect to a criterion
- Linear programming describes a broad class of optimization tasks in which both the constraints and the optimization criterion are linear functions

About Reductions

- Sometimes a computational task is sufficiently general that any subroutine for it can also be used to solve a variety of other tasks, which at glance might seem unrelated
- Once we have an algorithm for a problem, we can use it to solve other problems

