

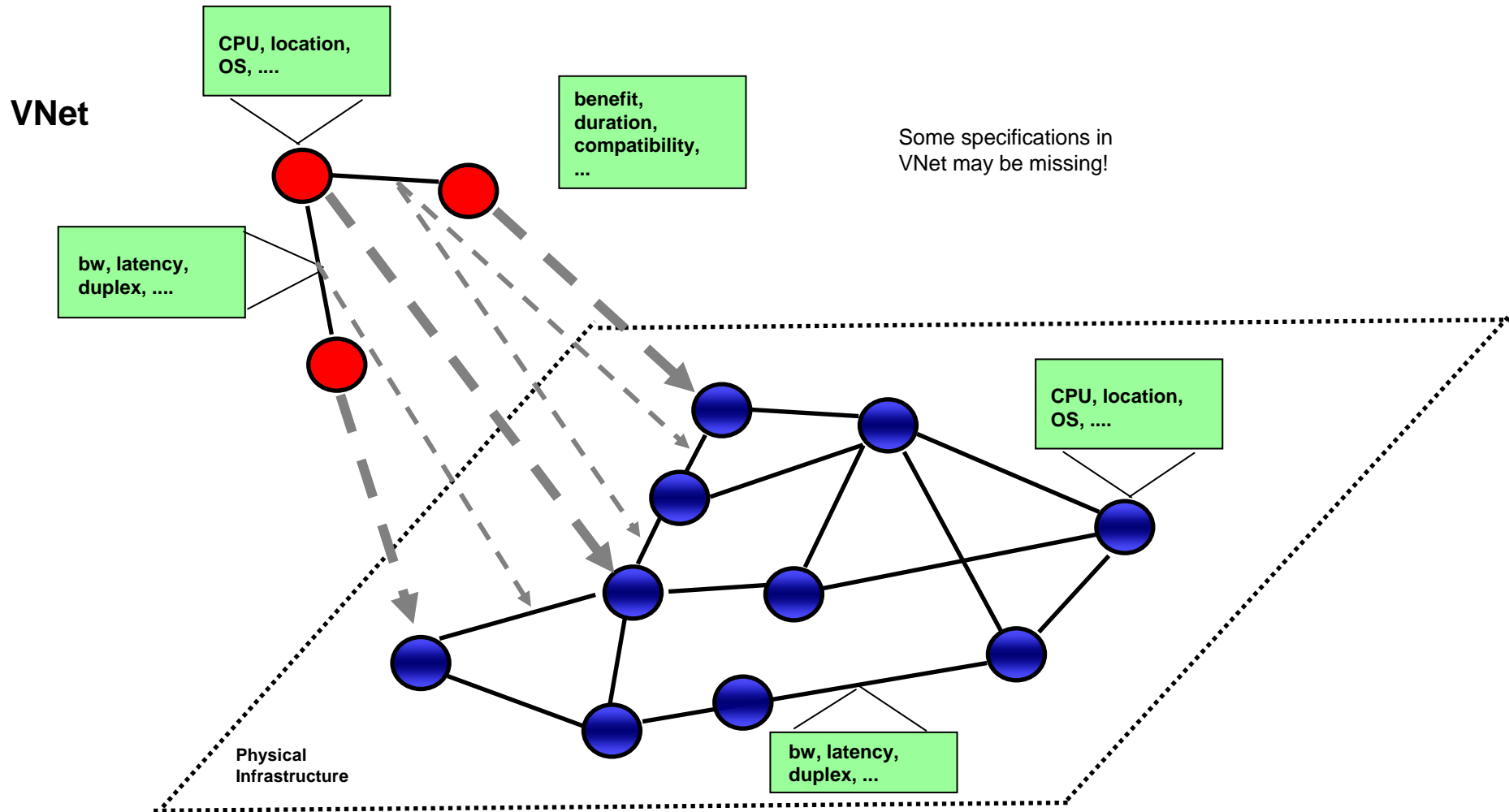
Competitive and Deterministic Embeddings of Virtual Networks



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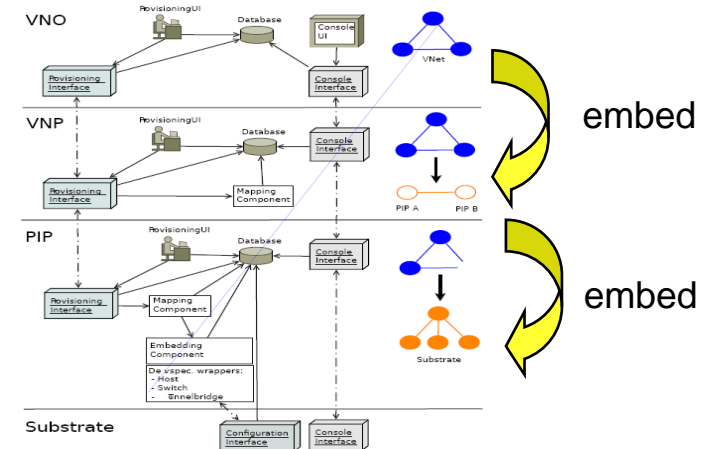
The Virtual Network Embedding Problem



Context: VNet Prototype Architecture at T-Labs (& DoCoMo)



Streaming server moves closer to (mobile) users,
yields better QoS (less important VNets can be
migrated away): <http://www.youtube.com/watch?v=IIJce0F1zHQ>



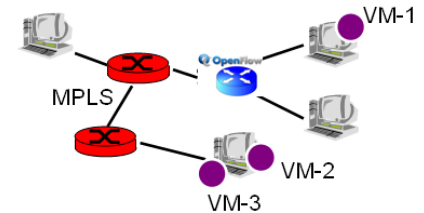
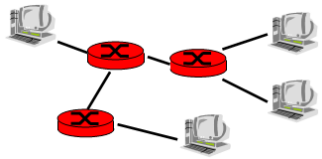
Prototype

- Layer 2 virtual networks w/ QoS (based on VLANs)
- Testbed TU Berlin + DoCoMo Eurolab Munich (multi-provider)
- Hierarchical provider model: economic roles communicate requirements by contract by contract...
- ... virtual networks get mapped in multiple stages!
- So far: Physical Infrastructure Provider and Virtual Network Operator
- Seamless migration (VNet configuration stays the same!), embedding

It's a virtual world...: What are virtual networks?

Virtualization

Attractive design principle: virtualization **abstracts** heterogeneous resources and allows for **resource sharing**



Node virtualization

- **success** over last years: revamped server business!
- **Xen, VMWare...**
- today's **clouds** hardly offer any physical resources

Link virtualization

- past and current **trend**
- **router virtualization** (Cisco, Juniper, ...)
- **MPLS / GMPLS** (too?) ubiquitous
- same holds for **VPNs**
- hype of SDN (split) architectures

Idea: Combine and generalize the two to entire virtual networks!

The next step: offer entire virtual networks!

Virtual Networks (VNets)

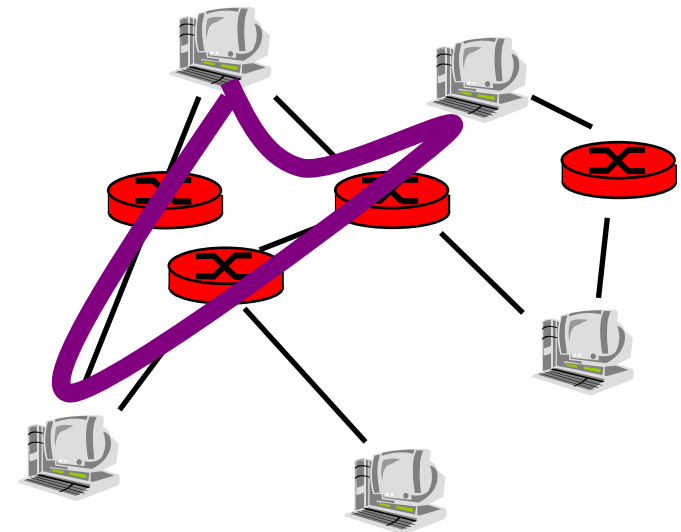
Basic unit: **set of links (graph)** rather than single links! VNets **co-habit** same physical network (substrate), but have different characteristics, **protocol stack**, ... (beyond IP). E.g., VNets may provide **QoS guarantees** on links (unlike VPNs)!

Use Cases:

- Traveling agents set up **video conference**
- **Start-up company** / academic institution experiments with new transmission protocols
- Data centers: performance guarantees despite dynamic demand to meet **deadlines** (improves service, saves bandwidth and energy, e.g., Octopus @ SIGCOMM 2011)
- Wireless access
- Migration for network maintenance or resilience

Advantages:

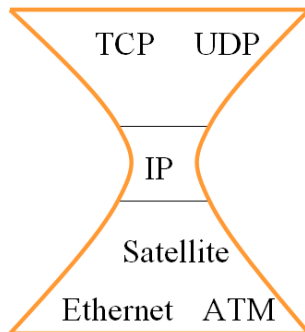
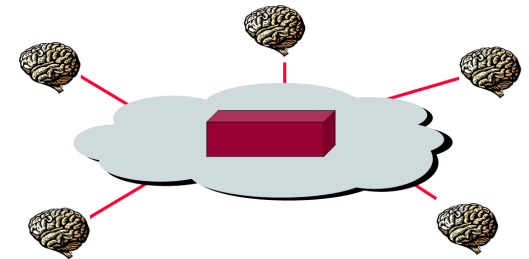
- **On demand** and **short notice**: as long as needed
- QoS **guarantees** and **isolation**
- No need to buy/build infrastructure, ...



VNets can help to innovate the Internet!

Ossification in today's Internet:

- **innovation** is only possible at lower and higher layers...
- ... but we cannot experiment with different network cores: Layer 3 is („**ossification**“)...
- different applications need **different technologies**: *bulk data transfers* vs *social networking* vs *gaming* vs *live streaming*... (distributions news vs social networking?)



IP hourglass

Network virtualization can help!

- experiment with different technologies
- **service-tailored**, but co-habit same substrate network
- enabling technologies: **OpenFlow**, VLANs, ...

Virtu Prototype Architecture: Challenges

Anja Feldmann, Gregor Schaffrath, Stefan Schmid (T-Labs/TU Berlin)

Algorithms

Implementation

Economics

(e.g., paper by Pan Hui
in this session)



A joint project with



Deutsche
Telekom

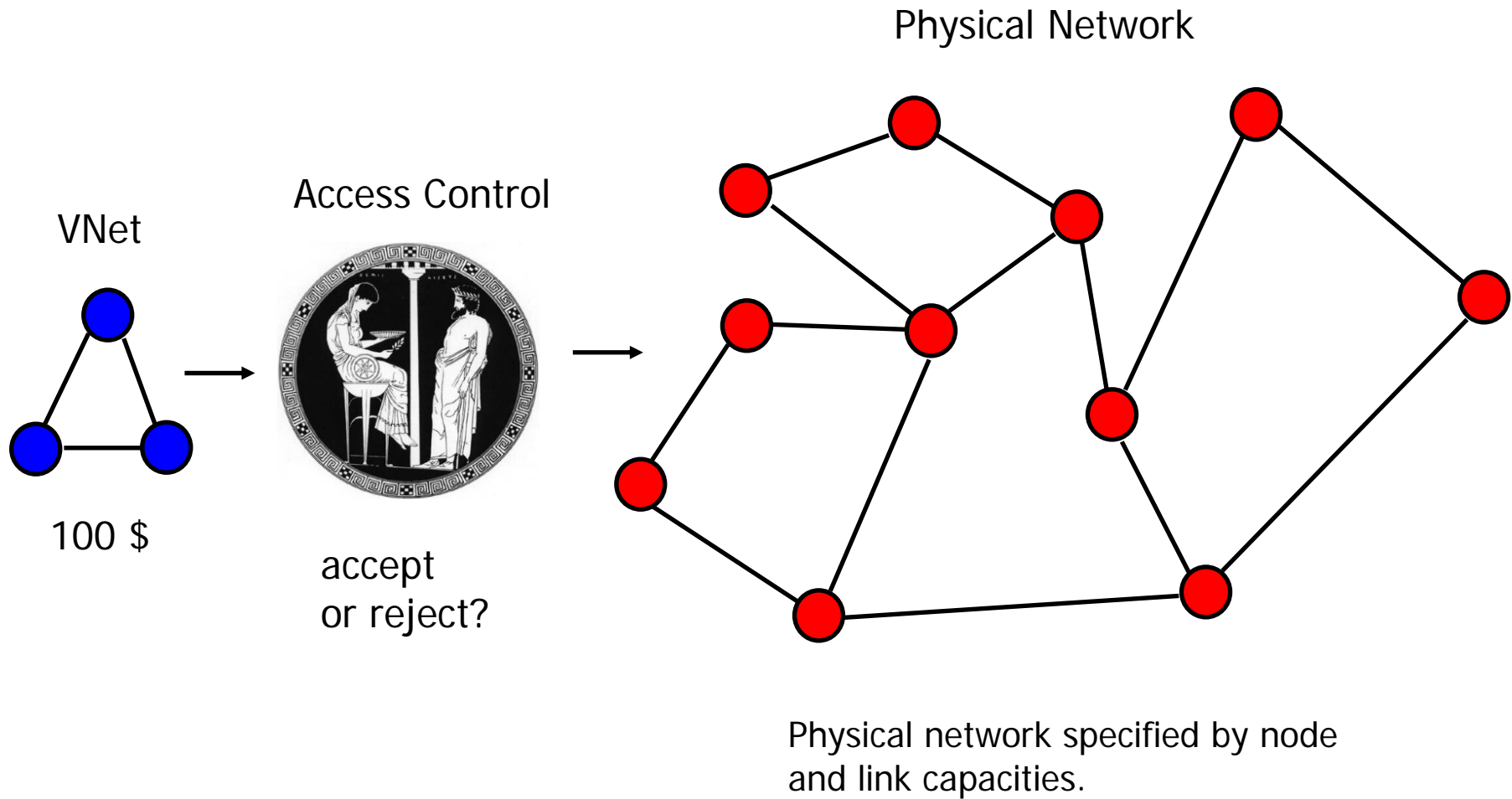


and

NTT
docomo:

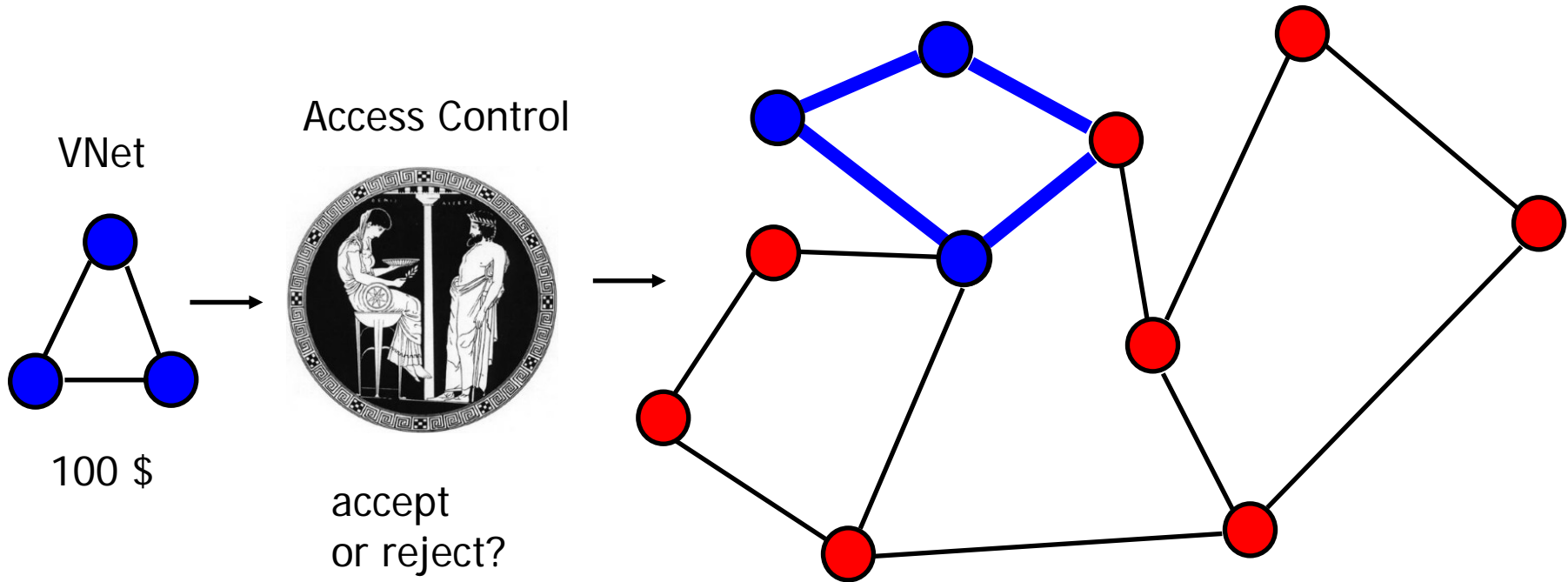
D. Jurca, A. Khan, W. Kellerer, K. Kozu and J. Widmer

This Paper: Competitive VNet Embedding



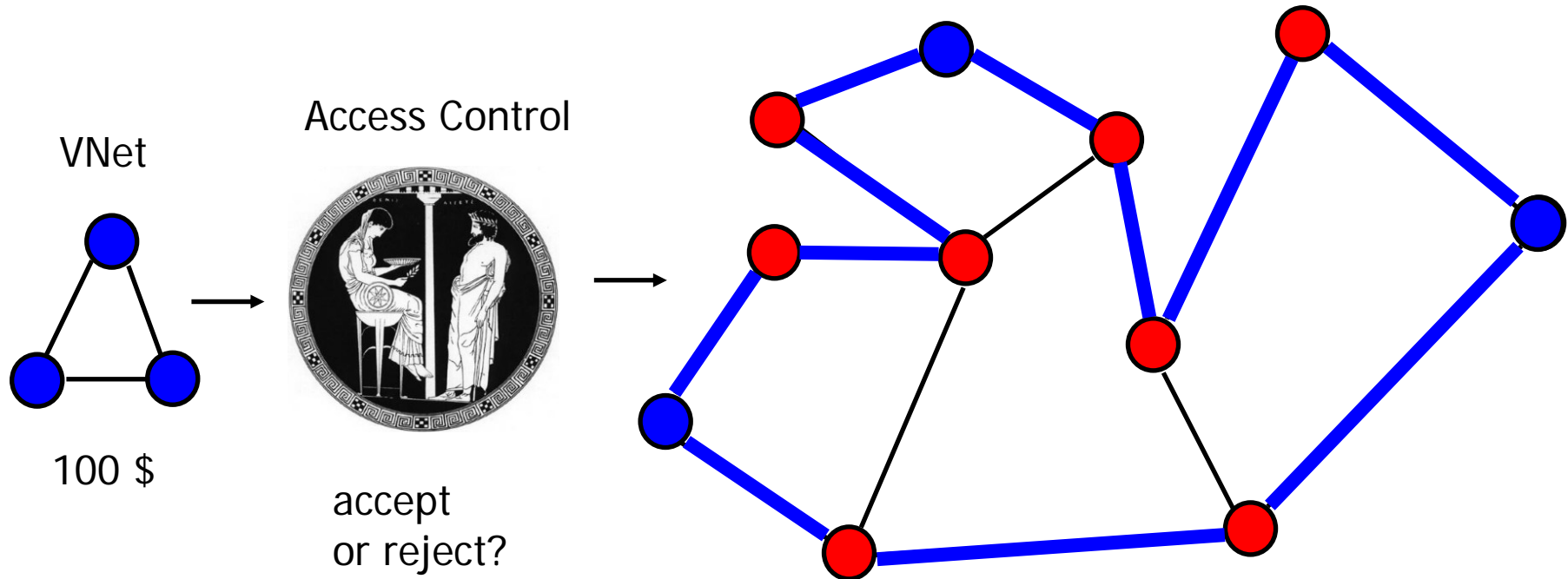
This Paper: Competitive VNet Embedding

Cheap realization => Yes!



This Paper: Competitive VNet Embedding

Expensive => No!



**It is like „online call admission“
for entire networks („telcos“)!**

Dealing with Unpredictable Demand?



How to deal with dynamic changes (e.g., mobility of users, arrival of VNets, etc.)?

Online Algorithm

Online algorithms make decisions at time t without any knowledge of inputs / requests at times $t' > t$.

Competitive Ratio

Competitive ratio r ,

$$r = \text{Cost}(\text{ALG}) / \text{cost}(\text{OPT})$$

Is the **price of not knowing the future!**

Competitive Analysis

An r -competitive online algorithm ALG gives a **worst-case performance guarantee**: the performance is at most a factor r worse than an optimal offline algorithm OPT!

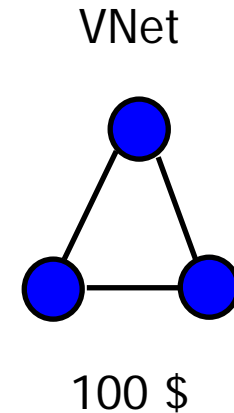
In virtual networks, many decisions need to be made online: online algorithms and network virtualization are **a perfect match!** 😊

No need for complex predictions but still good! 😊

VNet Specification and Model

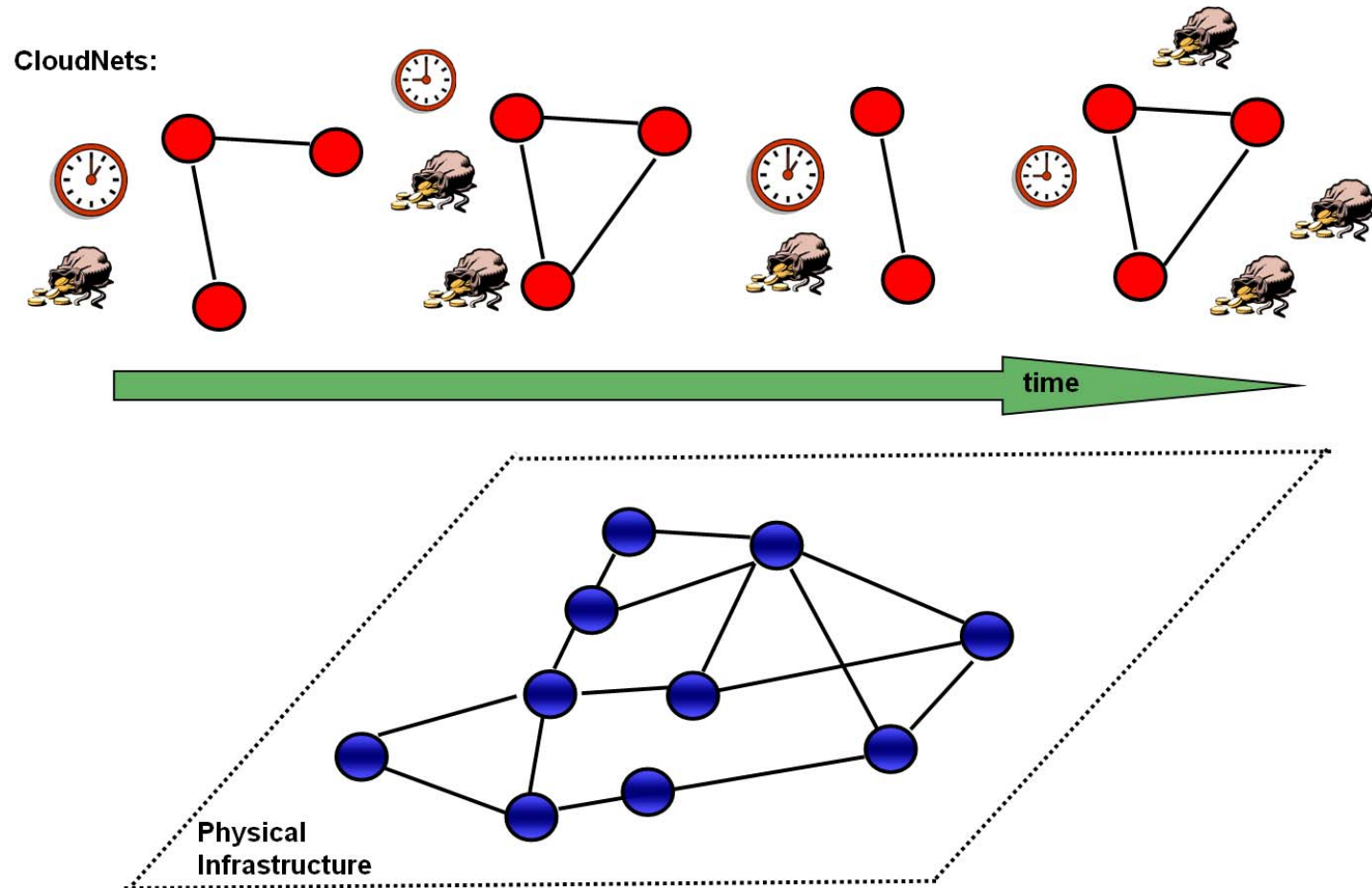
A VNet request can specify a benefit and many different QoS requirements:

- **benefit** if accepted
- a set of **terminals** to connect (or just some reqs!)
- desired bandwidth and allowed **traffic patterns**
- a **routing model**
- **duration** (from when until when?)



If VNets with these specifications arrive over time,
which ones to accept online?!

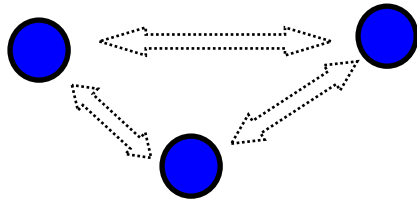
VNet Specification and Model



Supported VNet Traffic Patterns

Customer Pipe

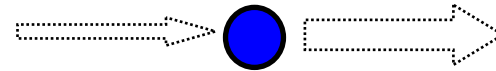
Every pair (u,v) of nodes requires a certain bandwidth.



Detailed constraints, only this **traffic matrix** needs to be fulfilled!

Hose Model

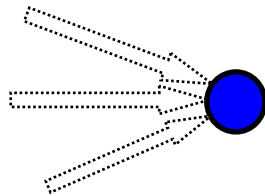
Each node v has **max ingress** and **max egress bandwidth**: each traffic matrix fulfilling them must be served.



More flexible, must support many traffic matrices!

Aggregate Ingress Model

Sum of ingress bandwidths must be at most a parameter I .

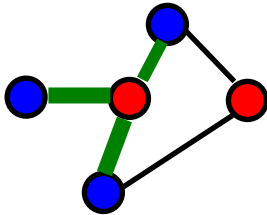


Simple and flexible! Good for **multicasts** etc.: no overhead, duplicate packets for output links, not input links already!

Supported VNet Routing Models

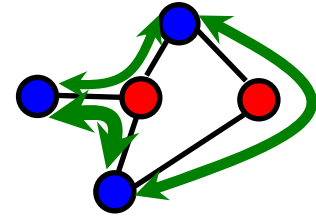
Tree

VNet is embedded as **Steiner tree**:



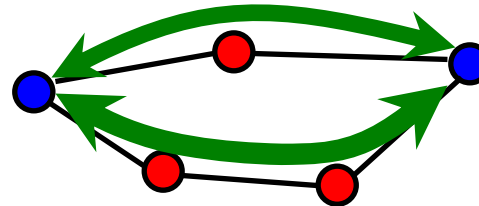
Single Path

Each pair of nodes communicates along a single path.



Multi Path

A **linear combination** specifies split of traffic between two nodes.

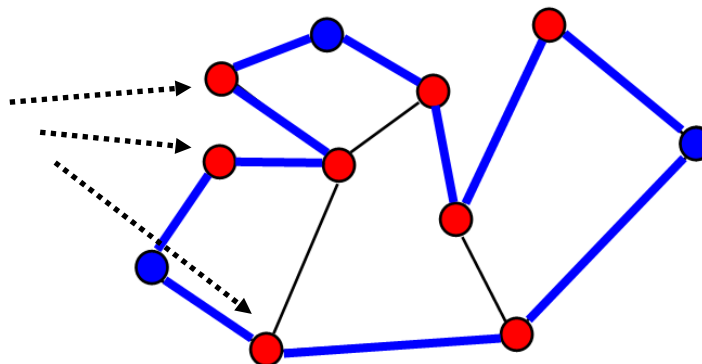


VNets arriving over time can request different models!

Theorem

The presented online algorithm GIPO is **log-competitive** in the amount of resources in the physical network!
If capacities can be exceeded by a log factor, it is even **constant competitive**.

Also works for **router loads**
(determined by **packet rate**)!



Analysis Overview

Algorithm design and analysis follows **online primal-dual** approach recently invented by **Buchbinder&Naor!**

(Application to general VNet embeddings, traffic&routing models, router loads, duration, approx oracles, ...)

1. Formulate **dynamic** primal and dual LP

$\begin{aligned} \min Z_j^T \cdot \mathbf{1} + X^T \cdot C \quad s.t. \\ Z_j^T \cdot D_j + X^T \cdot A_j \geq B_j^T \\ X, Z_j \geq \mathbf{0} \end{aligned}$ <p>(I)</p>	$\begin{aligned} \max B_j^T \cdot Y_j \quad s.t. \\ A_j \cdot Y_j \leq C \\ D_j \cdot Y_j \leq \mathbf{1} \\ Y_j \geq \mathbf{0} \end{aligned}$ <p>(II)</p>
---	---

Fig. 1: (I) The primal covering LP. (II) The dual packing LP.

2. Derive GIPO algorithm which always produces **feasible primal** solutions and where **Primal $\geq 2 \cdot$ Dual**

Algorithm 1 The General Integral (all-or-nothing) Packing Online Algorithm (GIPO).

Upon the j th round:

1. $f_{j,\ell} \leftarrow \operatorname{argmin}\{\gamma(j, \ell) : f_{j,\ell} \in \Delta_j\}$ (oracle procedure)
2. If $\gamma(j, \ell) < b_j$ then, (accept)
 - (a) $y_{j,\ell} \leftarrow 1$.
 - (b) For each row e : If $A_{e,(j,\ell)} \neq 0$ do

$$x_e \leftarrow x_e \cdot 2^{A_{e,(j,\ell)}/c_e} + \frac{1}{w(j, \ell)} \cdot (2^{A_{e,(j,\ell)}/c_e} - 1).$$

- (c) $z_j \leftarrow b_j - \gamma(j, \ell)$.
 3. Else, (reject)
 - (a) $z_j \leftarrow 0$.
-

Ideas of GIPO

GIPO invokes an **oracle procedure** to determine cost of VNet embedding!

Algorithm 1 The General Integral (all-or-nothing) Packing Online Algorithm (GIPO).

Upon the j th round:

1. $f_{j,\ell} \leftarrow \operatorname{argmin}\{\gamma(j, \ell) : f_{j,\ell} \in \Delta_j\}$ (oracle procedure)
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- (c) $z_j \leftarrow b_j - \gamma(j, \ell)$.
3. Else, (reject)
 - (a) $z_j \leftarrow 0$.

Algorithm efficient... except for oracle (static, optimal embedding)!
What if we only use a **suboptimal embedding** here?!

Effect of Approximate Oracles

Problem: computation of optimal embeddings **NP-hard!**
Thus: use approximate embeddings! (E.g., Steiner tree)

GIPO:

Algorithm 1 The General Integral (all-or-nothing) Packing Online Algorithm (GIPO).

Upon the j th round:

1. $f_{j,\ell} \leftarrow \arg\min\{\gamma(j, \ell) : f_{j,\ell} \in \Delta_j\}$ (oracle procedure)

2. If $\gamma(j, \ell) < b_j$ then, (accept)

(a) $y_{j,\ell} \leftarrow 1$.

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(c) $z_j \leftarrow b_j - \gamma(j, \ell)$.

3. Else, (reject)

(a) $z_j \leftarrow 0$.

Embedding Approx.:

**<insert your favorite
approx algo>**

Approx ratio r

Competitive ratio ρ

Lemma

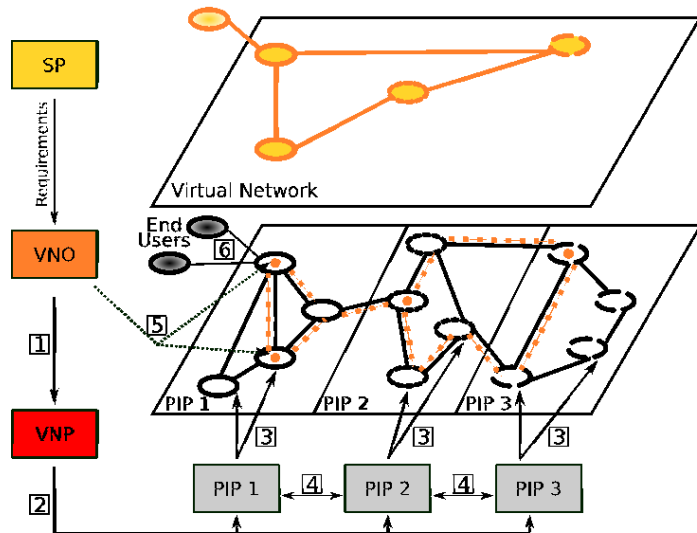
The approximation does not reduce the overall competitive ratio by much: we get **$\rho * r$ ratio!**

Conclusion

VNet admission control problem = „online call admission for telcos“

Summary:

Very general online VNet embedding algorithm.



Future work:

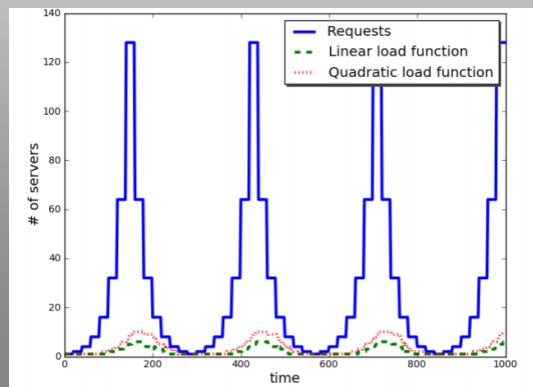
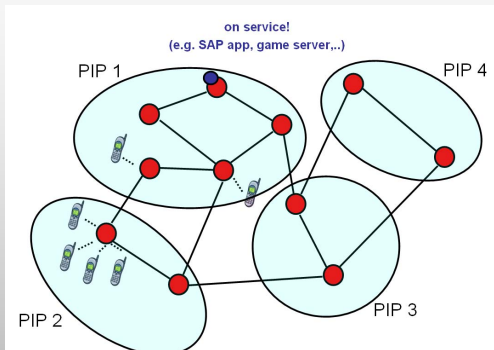
1. There is a **lower bound** of $\log(n \cdot T)$ from online circuit switching
2. More complex embedding constraints, full-duplex links mapped on asymmetric, half-duplex network? Or architecture **compatibility**?
3. With **preemption** better competitive ratio possible?
4. Non-linear objective functions? Maybe also with Buchbinder&Naor framework, using **semi-definite** programming!
5. Embedding support in **prototype** only offline so far!

Other algorithmic VNet problems studied so far...

Online Migration and Allocation

Goal:

Online VNet migration in multi-provider environments and online server allocation



VISA 2010, Hot-ICE 2011, IPTCOMM 2011

Even More General Embeddings

Goal:

E.g., different link types

Nodes:

$$\begin{aligned} \text{map_node: } & \sum_{v \in NE_S} \text{new}(u, v) = 1 & \forall u \in NE_{VN} \\ \text{set_new: } & \text{alloc}_{r_S}(u, v, r_V) \leq \text{cap}_{r_S}(v) \text{new}(u, v) & \forall u \in NE_{VN}, v \in NE_S, r_V \in R_V, r_S \in R_S \\ \text{req_min: } & \text{alloc}_{r_V}(u, v) \geq \text{new}(u, v) \text{req}(u, r_V, s) & \forall u \in NE_{VN}, r_V \in R_V, r_S \in R_S, s = \text{minimum} \\ \text{req_max: } & \text{alloc}_{r_V}(u, v) \leq \text{new}(u, v) \text{req}(u, r_V, s) & \forall u \in NE_{VN}, r_V \in R_V, r_S \in R_S, s = \text{maximum} \\ \text{req_con: } & \text{alloc}_{r_V}(u, v) = \text{new}(u, v) \text{req}(u, r_V, s) & \forall u \in NE_{VN}, r_V \in R_V, r_S \in R_S, s = \text{constant} \end{aligned}$$

Mapping:

$$\begin{aligned} \text{relate_V: } & \text{alloc}_{r_V}(u, v) \geq \min_{r_S} \text{alloc}_{r_S} \cdot \text{new}(u, v) & \forall u \in NE_V, v \in NE_S, r_V \in R_V \\ \text{allowed: } & \text{suit}(u, v) \geq \text{new}(u, v) & \forall u \in NE_V, v \in NE_S \\ \text{no_capacity: } & \sum_{u \in NE_V} \sum_{r_V \in R_V} \text{alloc}_{r_S}(u, v, r_V) \leq \text{cap}_{r_S}(v) & \forall v \in NE_S, r_S \in R_S \\ \text{capacity: } & \sum_{v \in NE_S} \sum_{u \in NE_V} \sum_{r_V \in R_V} \text{alloc}_{r_S}(u, v, r_V) \leq \text{cap}(r_S) & \forall r_S \in R_S \\ \text{load: } & \text{weight}_{r_S} / \text{cap}(r_S) \cdot \sum_{v \in NE_S} \sum_{u \in NE_V} \sum_{r_V \in R_V} \text{alloc}_{r_S}(u, v, r_V) \leq \text{load}(r_S) & \forall r_S \in R_S \\ \text{max_load: } & \text{load}(r_S) \leq \text{max_load} & \forall r_S \in R_S \end{aligned}$$

Resource-Variable Relation:

$$\begin{aligned} \text{resource: } & \sum_{r_S \in R_S} \text{prop}(r_V, r_S) \text{alloc}_{r_S}(u, v, r_V) = \text{alloc}_{r_V}(u, v) & \forall u \in NE_V, v \in NE_S, r_V \in R_V \\ \text{flow_res: } & \sum_{r_S \in R_S} \text{prop}(r_V, r_S) \text{flow}_{r_S}(f, v, w, r_V) = \text{flow}_{r_V}(f, v, w) & \forall f \in Fl(u), (v, w) \in NE_S^2, r_V \in R_V, \forall u \in NE_{VL} \end{aligned}$$

Links:

$$\begin{aligned} \text{map_link: } & \sum_{v \in NE_S} \text{new}(u, v) \geq 1 & \forall u \in NE_{VL} \\ \text{map_flow: } & \text{new}(f, v) \leq \text{new}(u, v) & \forall f \in Fl(u), v \in NE_S, \forall u \in NE_{VL} \\ \text{map_src: } & \text{new}(f, v) \geq \text{new}(q_f, v) & \forall f \in Fl(u), v \in NE_S, q_f \text{ source of } f; \forall u \in NE_{VL} \\ \text{map_sink: } & \text{new}(f, v) \geq \text{new}(d_f, v) & \forall f \in Fl(u), v \in NE_S, d_f \text{ sink of } f; \forall u \in NE_{VL} \\ \text{req_min: } & \sum_{u \in NE_S} (\text{flow}_{r_V}(f, v, w) - \text{flow}_{r_V}(f, v, u)) \geq \text{new}(q_f, v) \text{req}(u, r_V, s) - \text{new}(d_f, v) & \forall f \in Fl(u), v \in NE_S, r_V \in R_V, \forall u \in NE_{VL}, s = \text{minimum} \\ \text{req_max: } & \sum_{u \in NE_S} (\text{flow}_{r_V}(f, v, w) - \text{flow}_{r_V}(f, v, u)) \leq \text{new}(q_f, v) \text{req}(u, r_V, s) + \text{new}(d_f, v) & \forall f \in Fl(u), v \in NE_S, r_V \in R_V, \forall u \in NE_{VL}, s = \text{maximum} \\ \text{req_const: } & \sum_{u \in NE_S} (\text{flow}_{r_V}(f, v, w) - \text{flow}_{r_V}(f, v, u)) = \text{new}(q_f, v) \text{req}(u, r_V, s) - \text{new}(d_f, v) \text{req}(u, r_V, s) & \forall f \in Fl(u), v \in NE_S, r_V \in R_V, \forall u \in NE_{VL}, s = \text{constant} \end{aligned}$$

Link Allocation:

$$\begin{aligned} \text{exp_out: } & \sum_{u \in NE_S} \text{flow}_{r_S}(f, v, w, r_V) \leq \text{alloc}_{r_S}(u, v, r_V) & \forall f \in Fl(u), v \in NE_S, r_V \in R_V, r_S \in R_S, \forall u \in NE_{VL} \\ \text{exp_in: } & \sum_{u \in NE_S} \text{flow}_{r_S}(f, v, w, r_V) \leq \text{alloc}_{r_S}(u, v, r_V) & \forall f \in Fl(u), v \in NE_S, r_V \in R_V, r_S \in R_S, \forall u \in NE_{VL} \\ \text{direction: } & \text{flow}_{r_S}(f, v, w, r_V) \leq \text{new}(u, v) \text{cap}_{r_S}(u, w) & \forall f \in Fl(u), (u, w) \in NE_S^2, r_V \in R_V, r_S \in R_S, \forall u \in NE_{VL} \\ \text{relate_f: } & \sum_{u \in NE_S} \text{flow}_{r_S}(f, v, w, r_V) + \text{flow}_{r_S}(f, v, r_V) \geq \text{new}(f, v) & \forall f \in Fl(u), \forall u \in NE_{VL}, v \in NE_S, r_V \in R_V, r_S \in R_S \end{aligned}$$

Migration:

$$\begin{aligned} \text{new: } & \sum_{v \in NE_S} \text{old}(u, v) \geq \text{mig}(u) & \forall u \in NE_V \\ \text{migrated: } & \text{old}(u, v) - \text{new}(u, v) \leq \text{mig}(u) & \forall u \in NE_V, v \in NE_S \end{aligned}$$

ArXiv 2011

Thank you!

Further reading: project website!

<http://www.net.t-labs.tu-berlin.de/~stefan/virtu.shtml>

Simplified LP

$$\min \sum_{e \in E} x_e \cdot c(e) + \sum_{v \in V} x_v \cdot c(v) + \sum_i z_i \cdot d_i \quad s.t.$$

$$\text{(Covering Const.) } \forall i \forall \Delta \in \Delta_i z_i + \alpha(i, \Delta) \geq b_i$$

$$\forall i \forall \Delta \in \Delta_i x_e, x_v, z_i \geq 0$$

(I)

maximize
benefit!

$$\max \sum_i b_i \cdot \sum_{\Delta_{ij} \in \Delta_i} f_{ij} \quad s.t.$$

(Vertex Capacity Const.)

$$\forall v \in V \quad flow(v) \leq c(v)$$

(Edge Capacity Const.)

$$\forall e \in E \quad flow(e) \leq c(e)$$

(Demand Const.)

$$\forall i \quad \sum_{\Delta_{ij} \in \Delta_i} f_{ij} \leq d_i$$

$$f \geq 0$$

(II)

realization of i-th
request (will be integer,
accept fully or not at all)

... while ensuring
capacity and
no more than demand!

Fig. 1: (I) The Primal linear embedding program. (II) The Dual linear embedding program.

Simplified LP

essentially, exponential load...

$$\min \sum_{e \in E} x_e \cdot c(e) + \sum_{v \in V} x_v \cdot c(v) + \sum_i z_i \cdot d_i \quad s.t.$$

$$\text{(Covering Const.) } \forall i \forall \Delta \in \Delta_i z_i + \alpha(i, \Delta) \geq b_i$$

$$\forall i \forall \Delta \in \Delta_i x_e, x_v, z_i \geq 0$$

(I)

$$\max \sum_i b_i \cdot \sum_{\Delta_{ij} \in \Delta_i} f_{ij} \quad s.t.$$

$$\text{(Vertex Capacity Const.)} \quad \forall v \in V \quad flow(v) \leq c(v)$$

$$\text{(Edge Capacity Const.)} \quad \forall e \in E \quad flow(e) \leq c(e)$$

$$\text{(Demand Const.)} \quad \forall i \quad \sum_{\Delta_{ij} \in \Delta_i} f_{ij} \leq d_i$$

$$f \geq 0$$

(II)

Fig. 1: (I) The Primal linear embedding program. (II) The Dual linear embedding program.

Simplified Algo

Algorithm 1 The ISTP Algorithm.

Input: $G = (V, E)$ (possibly infinite), sequence of requests $\{r_i\}_{i=1}^{\infty}$ where $r_i \triangleq (U_i, c_i, d_i, b_i)$.

Upon arrival of request r_i :

1) $j \leftarrow \operatorname{argmin}\{\alpha(i, j) : \Delta_{ij} \in \Delta_i\}$ (find a lightest realization over the terminal set U_i using an oracle).

2) If $\alpha(i, j) < b_i$ then, (accept r_i)

a) $f_{ij} \leftarrow d_i$.

b) For each $e \in E(\Delta_{ij})$ do

$$x_e \leftarrow x_e \cdot 2^{d_i/c(e)} + \frac{1}{|V(\Delta_{ij})|} \cdot (2^{d_i/c(e)} - 1).$$

c) For each $v \in V(\Delta_{ij})$ do


$$x_v \leftarrow x_v \cdot 2^{c_i/c(v)} + \frac{d_i/c_i}{|V(\Delta_{ij})|} \cdot (2^{c_i/c(v)} - 1).$$

d) $z_i \leftarrow b_i - \alpha(i, j)$.


3) Else, (reject r_i)

a) $z_i \leftarrow 0$.

oracle
(triangle only)



update primal
variables if accepted



Simplified Analysis

Step (2b) increases the cost $\sum_e x_e \cdot c(e)$ as follows
(change $\Delta(x_e) = \sum_e (x_e^t - x_e^{t-1}) \cdot c(e)$):

$$\begin{aligned}
 \Delta(x_e) &\leq \sum_{e \in \Delta} \left[x_e \cdot (2^{d_i/c(e)} - 1) + \frac{1}{|V(\Delta_{ij})|} \cdot (2^{d_i/c(e)} - 1) \right] \cdot c(e) \\
 &= \sum_{e \in \Delta} \left(x_e + \frac{1}{|V(\Delta_{ij})|} \right) \cdot (2^{d_i/c(e)} - 1) \cdot c(e) \\
 &\leq c_{\min}(e) \cdot (2^{d_i/c_{\min}(e)} - 1) \sum_{e \in \Delta} \left(x_e + \frac{1}{|V(\Delta_{ij})|} \right) \\
 &\leq d_i \cdot (2^1 - 1) \sum_{e \in \Delta} \left(x_e + \frac{1}{|V(\Delta_{ij})|} \right) \\
 &\leq d_i \cdot \sum_{e \in \Delta} x_e + d_i \cdot \sum_{e \in \Delta} \frac{1}{|V(\Delta_{ij})|} \\
 &\leq d_i \cdot \sum_{e \in \Delta} x_e + d_i \cdot \quad (1)
 \end{aligned}$$

Step (2c) increases the cost $\sum_v x_v \cdot c(v)$ as follows
(change $\Delta(x_v) = \sum_v (x_v^t - x_v^{t-1}) \cdot c(v)$):

$$\begin{aligned}
 \delta(x_v) &\leq \sum_{v \in \Delta} \left[x_v \cdot (2^{c_i/c(v)} - 1) + \frac{d_i/c_i}{|V(\Delta_{ij})|} \cdot (2^{c_i/c(v)} - 1) \right] \cdot c(v) \\
 &= \sum_{v \in \Delta} \left(x_v + \frac{d_i/c_i}{|V(\Delta_{ij})|} \right) \cdot (2^{c_i/c(v)} - 1) \cdot c(v) \\
 &\leq c_{\min}(v) \cdot (2^{c_i/c_{\min}(v)} - 1) \sum_{v \in \Delta} \left(x_v + \frac{d_i/c_i}{|V(\Delta_{ij})|} \right) \\
 &\leq c_i \cdot (2^1 - 1) \sum_{v \in \Delta} \left(x_v + \frac{d_i/c_i}{|V(\Delta_{ij})|} \right) \\
 &\leq c_i \cdot \sum_{v \in \Delta} x_v + c_i \cdot \sum_{v \in \Delta} \frac{d_i/c_i}{|V(\Delta_{ij})|} \\
 &\leq c_i \cdot \sum_{v \in \Delta} x_v + d_i \cdot \quad (2)
 \end{aligned}$$

after each request,
primal variables
constitute feasible
solutions...