

Neutrality in Future Public Clouds: Implications and Challenges

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 - ▶ subsidiaries/affiliates of the few public cloud providers compete with their own tenants, e.g., both Amazon Prime and Netflix use Amazon EC2.
- ▶ We expect antitrust concerns to lead to neutrality regulations in the public-cloud marketplace as they did in the commercialized, public Internet marketplace.
 - ▶ The FCC deemed the Internet a utility in 2015 in a move to shore-up neutrality rules.
- ▶ There is relatively scant existing consideration of neutrality issues re. virtualized services in the public cloud, beyond how *network* neutrality relates to the cloud.

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- ▶ So a more challenging auditing/enforcement setting.

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 - ▶ allocates any discretionary resources,
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 - ▶ deals with congestion.

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 - ▶ deals with congestion.
- ▶ Without a single (or equivalent lumped) resource type, it may be challenging to identify groups of similar SLAs in order to define fair treatment.

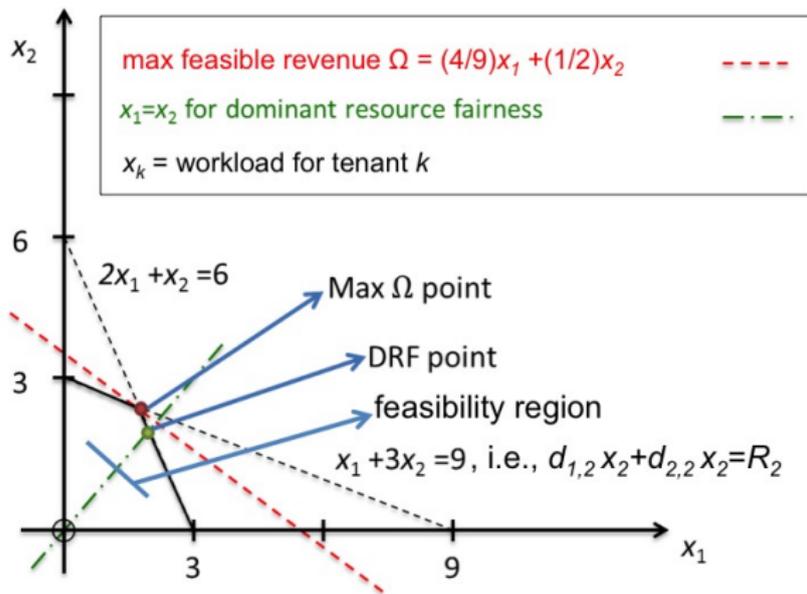
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- ▶ A neutral provider's behavior must not be based on “inside information” or preferences, *e.g.*,
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 - ▶ [Mogul et al. '15] consider a network bandwidth allocation problem wherein a cloud provider reckons tenant sensitivity to network bandwidth underprovisioning and uses this to allocate bandwidth differentially to improve its profits.
- ▶ Given resource-oriented SLAs (not performance oriented), if cloud reckons “sensitivity” based on:
 - ▶ precise knowledge of tenant's performance & requirements, then not neutral
 - ▶ measured resource usage, then may be neutral
 - ▶ Amazon “burstable” VM instances use token-bucket regulation to govern access to network IO and CPU, and so may be more aggressively consolidated without workload inferences (neutrally, see technical report)

Resource congestion, choice of boundary Nash equilibria



x = incident tenant workload, R = resource (CPU, memory) capacity,
 d = tenant resource demand per unit workload

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- ▶ The tenants have the same SLAs corresponding to one guaranteed core, and the cloud has a discretionary third core that it can share between the two tenants.
- ▶ Empirically, we found that a single core can achieve throughput of 75k ops/s with satisfactory mean response times of $400\mu\text{s}$.

Resv vs CFS based sharing of discretionary core - set-up

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- ▶ The premise here is that with greater demand variability, tenant 2 will out-compete tenant 1 for the discretionary CPU core under CFS;
- ▶ but this may not be entirely the case when the tenant demands are positively correlated, *i.e.*, how workload is consolidated by the cloud will impact such “iso-neutral” decision-making.

Resv vs CFS based sharing of discretionary core - results

Table: Latency of tenants, uncorrelated demand scenario:

	CFS3:7		CFS1:1		Resv1:1	
	95-th	avg	95-th	avg	95-th	avg
Tenant 1	484	318	449	382	451	365
Tenant 2	418	254	435	284	444	299

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- ▶ See paper for an example involving memory allocation.

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 - ▶ select resource granularities
 - ▶ simplify by use of a lumped resource, *e.g.*, energy
 - ▶ use of hardware verification technology
 - ▶ role of third-party verifiers
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- ▶ General scalability and accuracy issues