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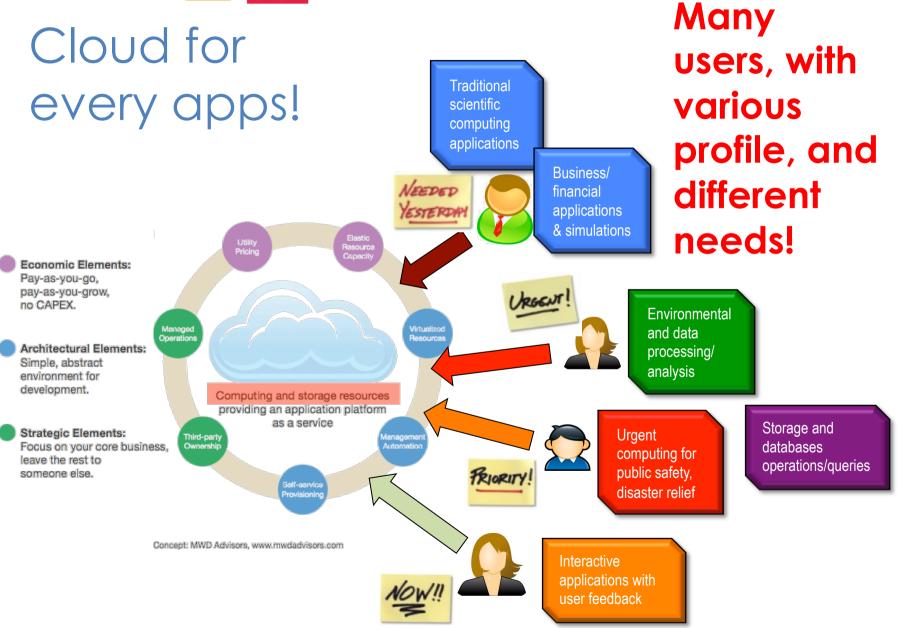
School of Computer Science & Informatics - Cardiff University, UK o.f.rana@cs.cardiff.ac.uk

Congduc Pham Laboratoire informatique-Université de Pau, France congduc.pham@univ-pau.fr Revenue creation for rate adaptive stream management in multi-tenancy environments



GECON 2013 10th International Conference on Economics of Grids, Clouds, Systems and Services September 18-20, 2013 Zaragoza, Spain





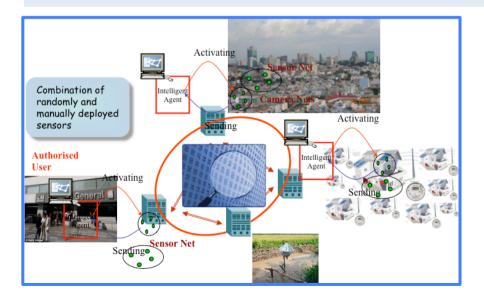






Focus on data streams management

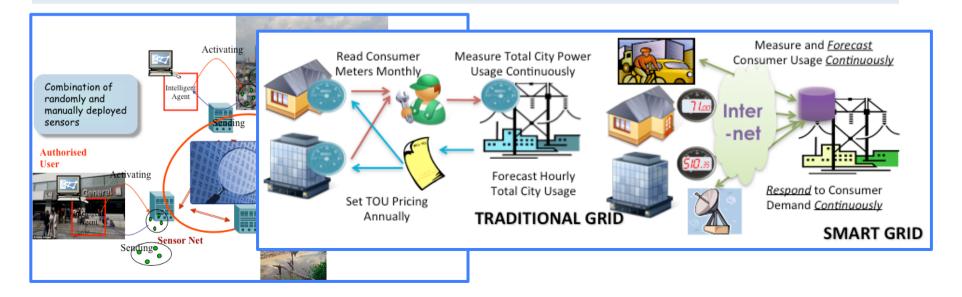
- Applications send continuous/long-lived data flows
- Data sent to a cloud **must** be processed in a **timely** manner but...
- ... data rates can be **sporadic**.





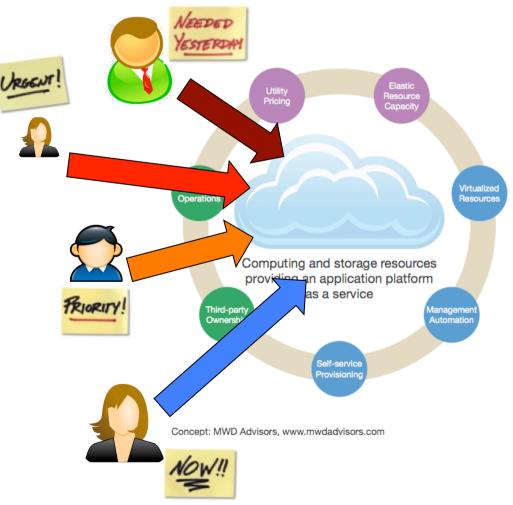
Focus on data streams management

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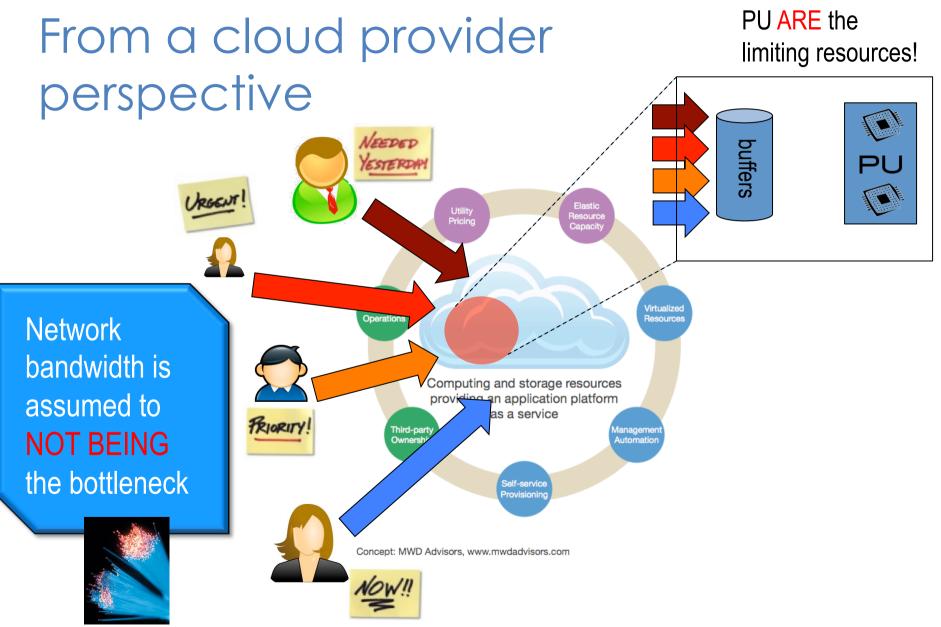




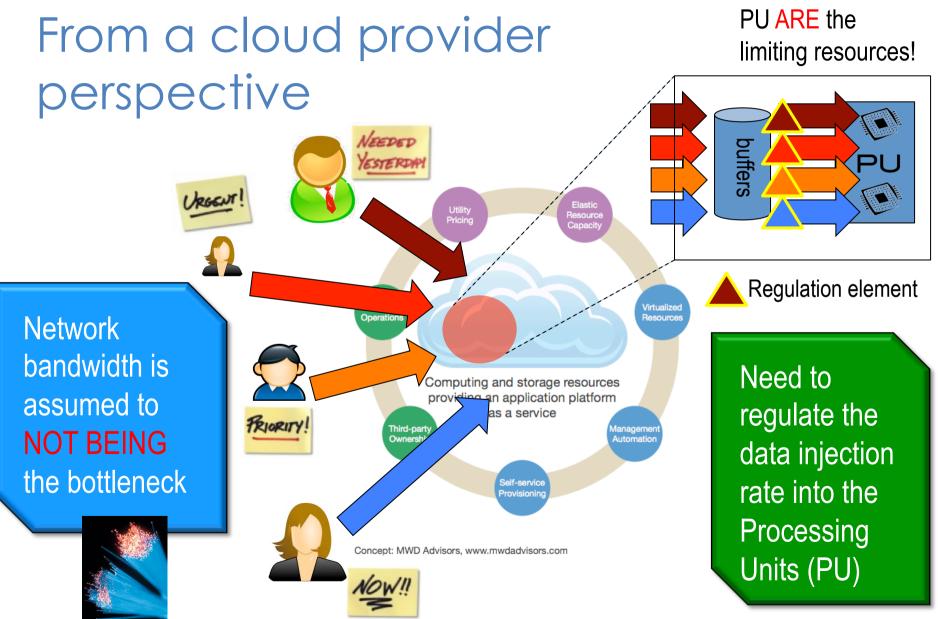
From a cloud provider perspective







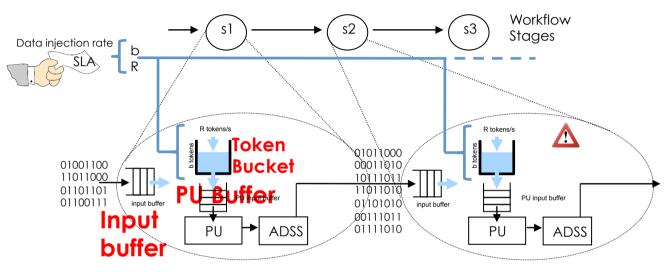






R. Tolosana, J. Banares, C. Pham, O. Rana, "Enforcing QoS in Scientific Workflow Systems Enacted Over Cloud Infrastructures", Journal of Computer and System Science (JCSS), 78(5), Elsevier.

System Architecture



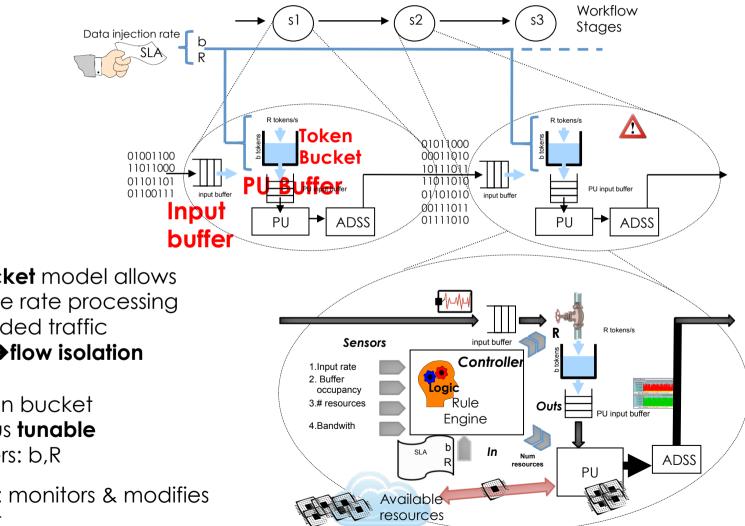
Token Bucket model allows for variable rate processing with bounded traffic envelop →flow isolation

Each token bucket provides us **tunable** parameters: b,R



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System Architecture



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Each token bucket provides us tunable parameters: b,R

Controller: monitors & modifies behaviour



Last year GECON'12

Adaptive infrastructure for shared clouds

- Multiple concurrent data streams with SLA
- Token bucket behaviour is regulated by b, R parameters, flow isolation at each node
- Rule-based SLAs can specify behaviours allowing the controller to take different actions when a threshold is reached
 - Load-shedding: drop data stored by the token bucket buffer
 - Modify the mean injection rate R

Basic revenue model

- User classes, e.g. : Gold, Silver, Bronze,...
- **Revenue:** price charge to n clients for m operations: $\sum_{i=1}^{n} \sum_{j=1}^{m} Pr(O_{ij})$
- **Cost:** for performing such operations: $c(O_{ij})$
- **Penalisation:** in case of QoS violation for client: $PSLA_{ij}$

Maximizing provider revenue

$$\sum_{i=1}^{n} \sum_{j=1}^{m} Pr(O_{ij}) - \sum_{i=1}^{n} \sum_{j=1}^{m} min(c(O_{ij}), PSLA_{ij})$$



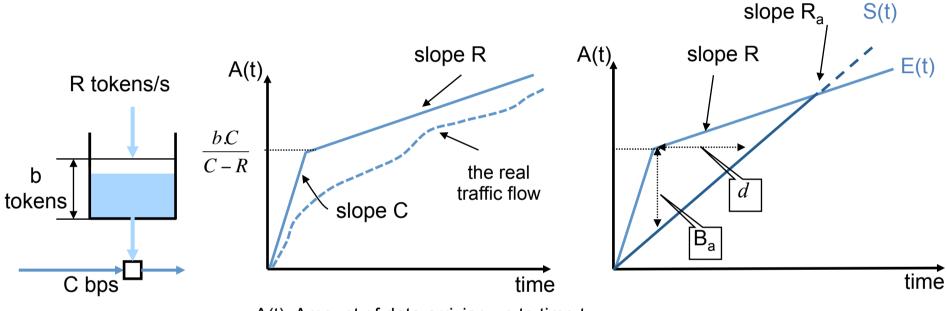
Unified resource mngt & more elaborated revenue model

- Allocation of new resources may have inadequate time scale or be very costly
- QoS requirements are often defined using the worst case scenario or statistically
- Business policies can be used to improve revenue and to provide more flexibility in SLA definition
- Cloud providers can take advantage of locally unused resources that are cheaper than allocating new resources

 redistribute unused resources
- Revenue & penalisation depends on user class, i.e. Gold users incur both high revenue and high penalty while Bronze can have low revenue and no penalty>redistribute pre-allocated ressources from less-prioritized users



Token Bucket for traffic shaping



A(t): Amount of data arriving up to time t

- Two key parameters of interest:
 - R: Also called the committed information rate (CIR), it specifies how much data can be sent or forwarded per unit time on average
 - b: it specifies for each burst how much data can be sent within a given time without creating scheduling concerns. Tokens in excess are normally dropped.



Shaping in image

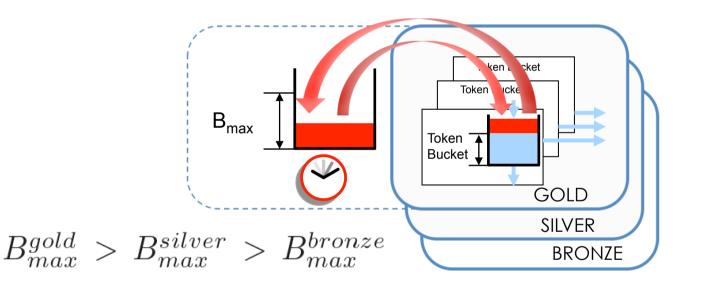
Constant

Dynamic			

Dynamic shaping allows for variable data injection rate (variable sensing rate for instance). Need however to bound the amount of provided service per time period T.



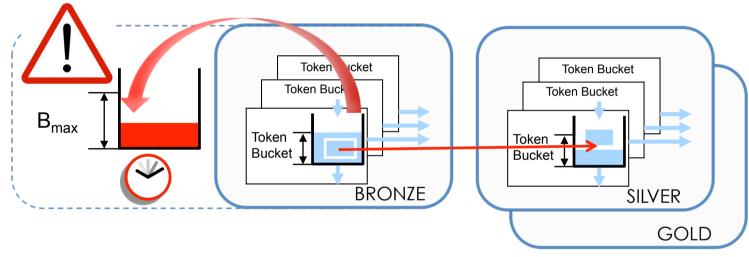
Unified resource mngt with TB → Redistribute unused resources



- Under-utilization of resource in a flow will produce tokens in excess
- Within a service class,
 - Tokens in excess of all flows are collected and stored up to B_{max} tokens
 - Token's lifetime is limited to a few control intervals to limit inconsistency



Unified resource mngt with TB →Take resources from lower classes



- Silver class has higher revenue and higher penalty than Bronze class for example: shortage of resource in Silver class is more costly
- Taking resources from Bronze to Silver is more revenue-efficient
 - Tokens are taken directly from a Bronze flow's token bucket
 - Can put a limit to the number of tokens the system can take
- Safer than taking tokens from the Bronze unused token bucket



More elaborated revenue model



Buy remote resources (from other Cloud provider)

Allocate new local resources (launch new VMs)

Redistribute pre-allocated ressources from less-prioritized users

Redistribute unused resources



More elaborated revenue model



Buy remote resources (from other Cloud provider)

Allocate new local resources (launch new VMs)

Redistribute pre-allocated ressources from less-prioritized users

Redistribute unused resources



Instant revenue & global revenue

 $Instant \ Revenue = \sum_{i=1}^{n} (CostPU_{client} - CostPU_{provider}) * \#PU \\ -\sum_{i=1}^{n} \#penalties_i * CostPenalties_i \\ -\Delta \#PU * CostPU_{provider}$

- **Global revenue** is the accumulated InstantRevenue over time.
- With the unified resource management proposition, we can
 - Reduce the number of penalties by using unused resources
 - **Reduce** Δ **#PU** by taking resources from less priority flows
- Using a single performance criterion, i.e. global revenue, simplifies the optimization problem

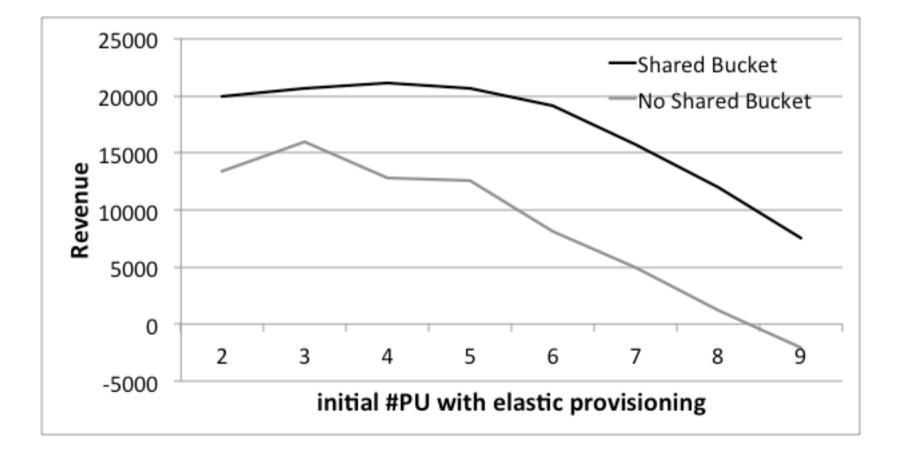


Simulation settings

1 data chunk = 1 tokenPoisson Data injection rate b=10 tokens R=20 tokens/s OFF ON 4 U[2s,5s] 20u/s 0 data chunk/s N 15u/s 600u/s On average, 4 PU cket ken L Token icke are required to B_{max}=80 process the 4 Gold Token Bucket flows

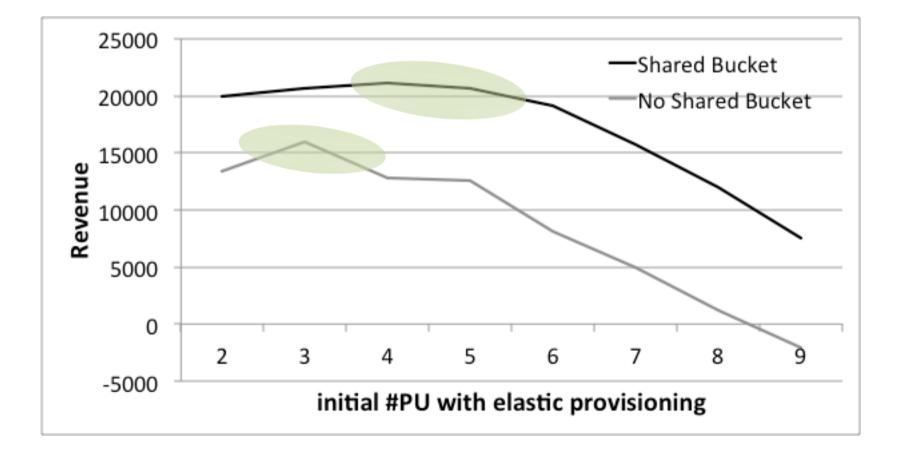


Simulation results - 300s with elastic provisioning



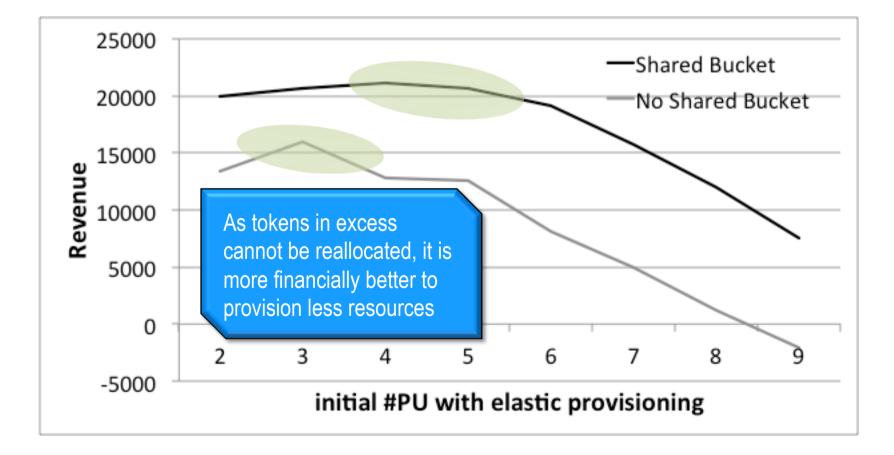


Simulation results - 300s with elastic provisioning



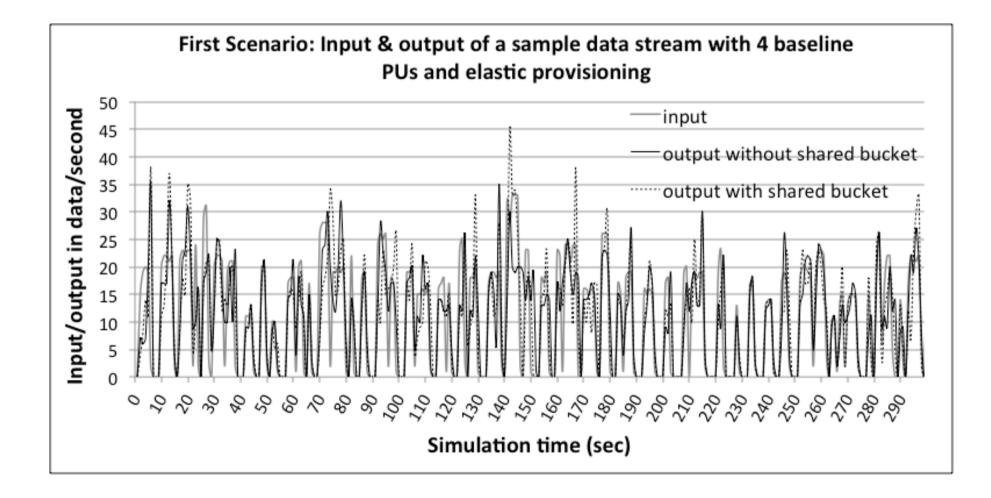


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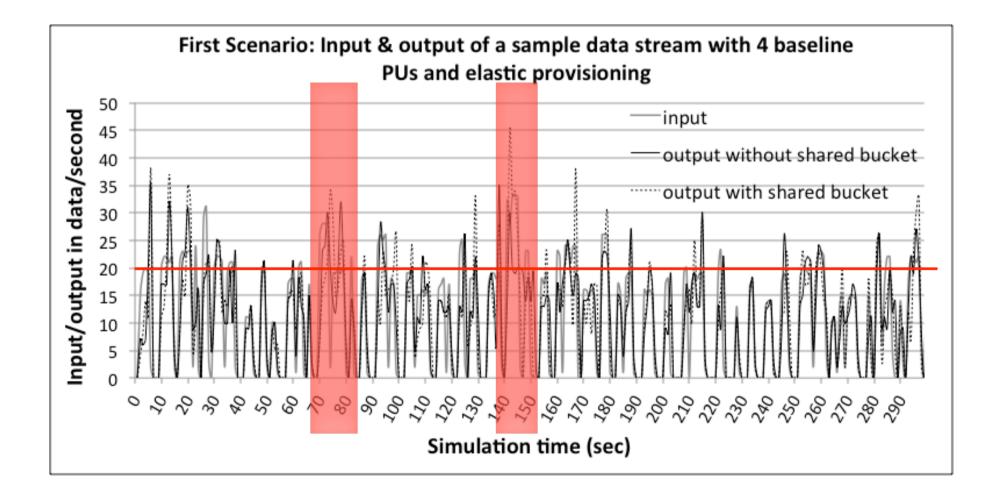


Simulation results - throughput 4PU

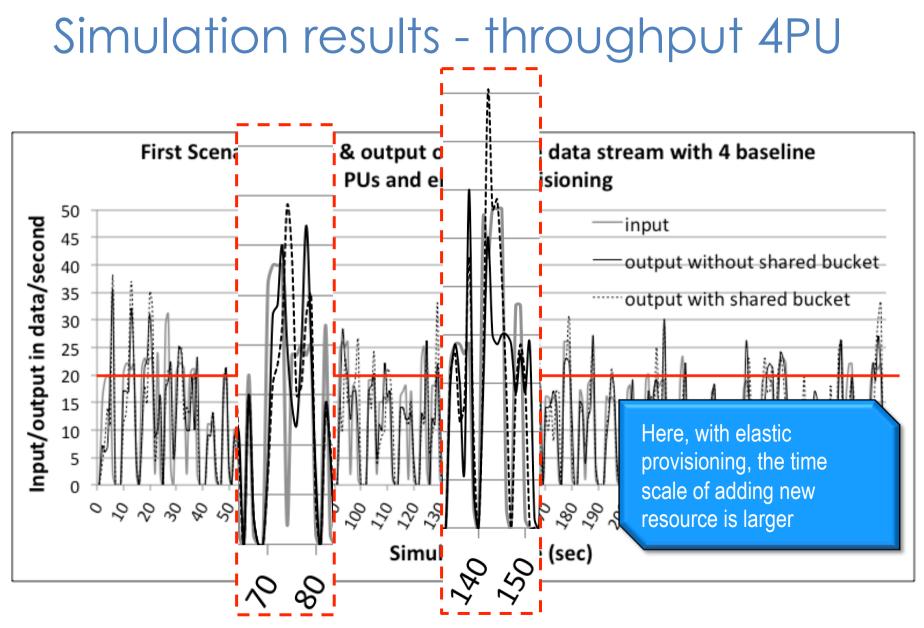




Simulation results - throughput 4PU

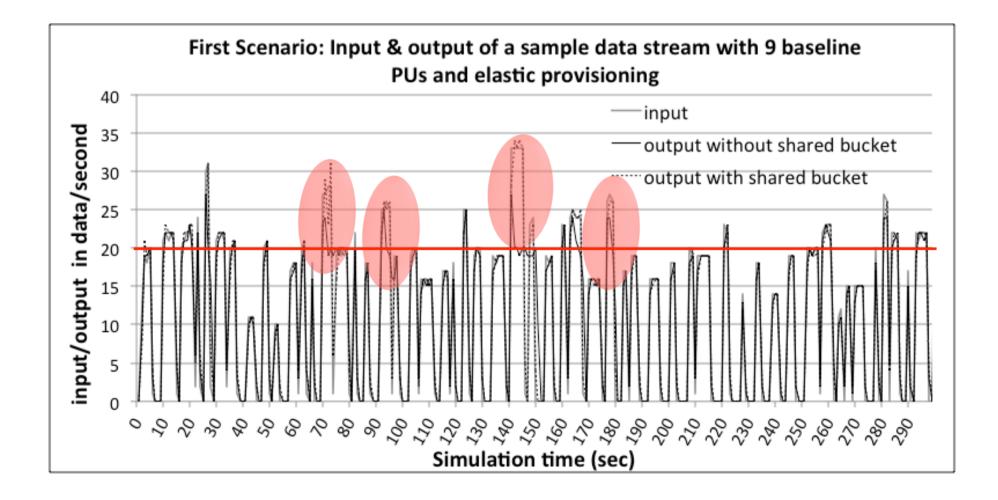






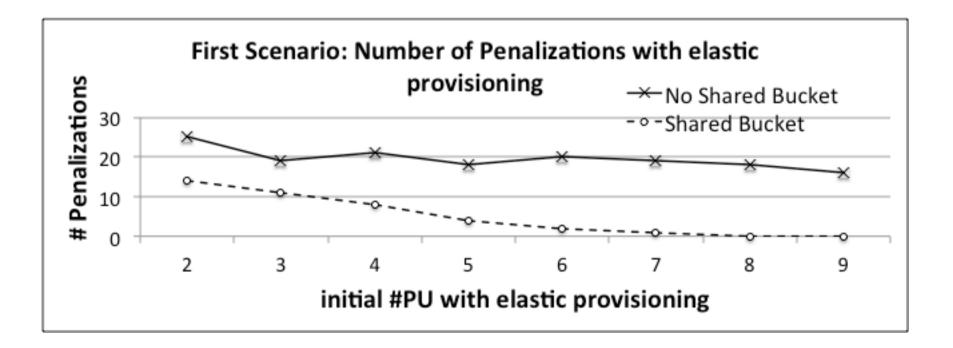


Simulation results - throughput 9PU



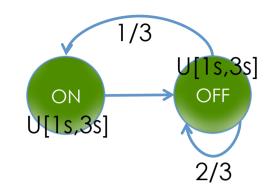


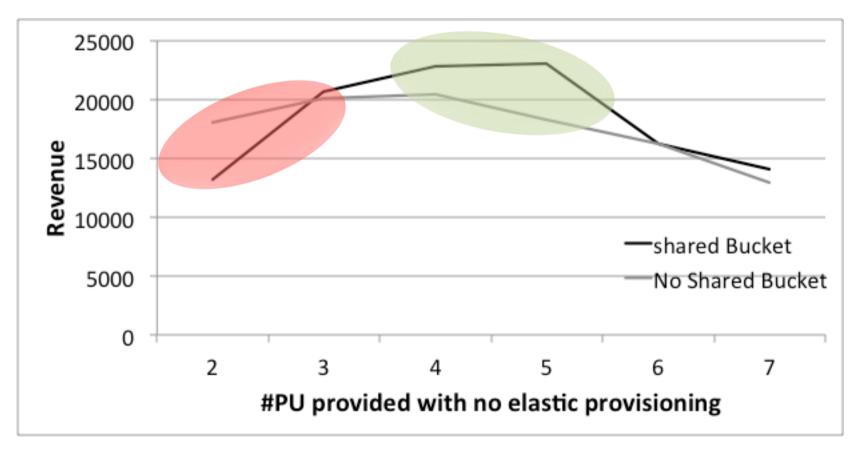
Simulation results number of penalisations





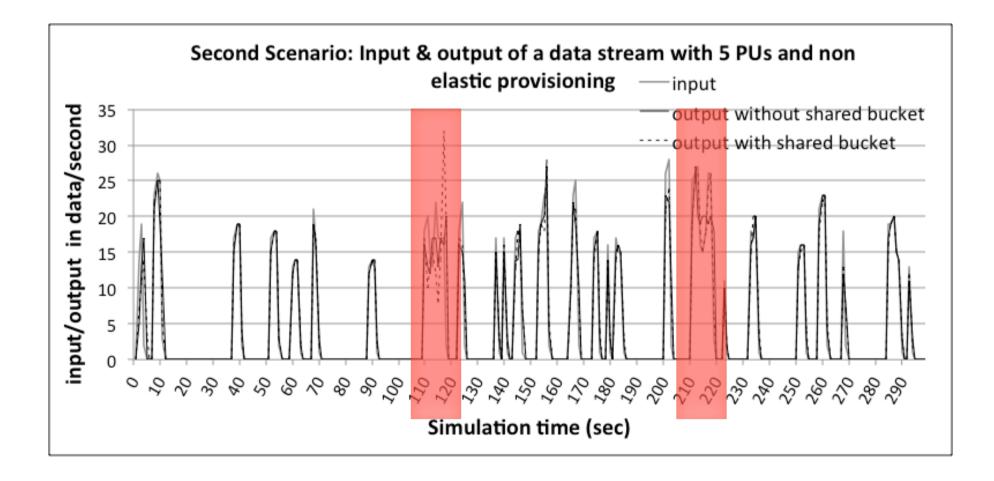
Simulation results - 300s without elastic provisioning



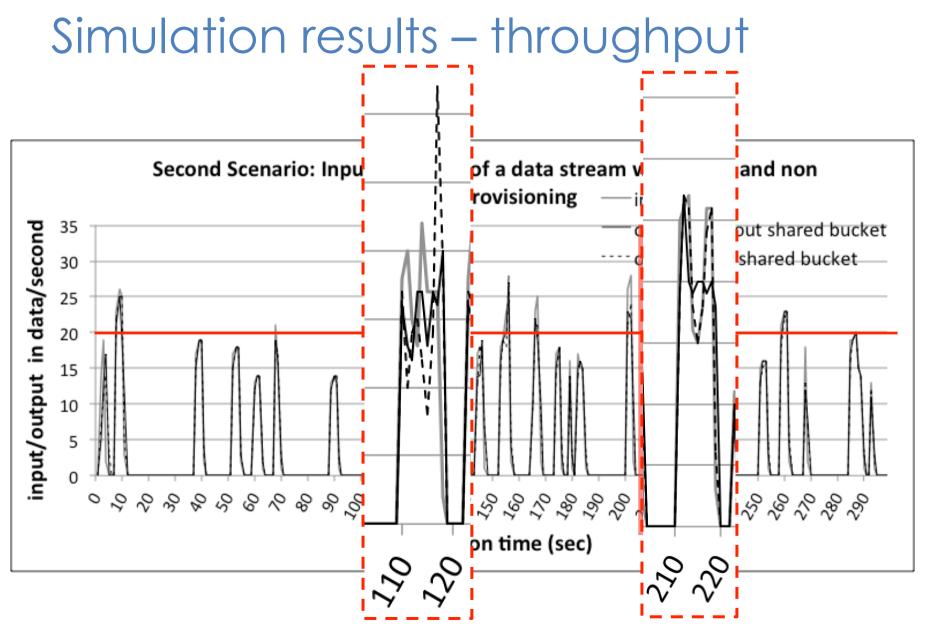




Simulation results – throughput









On-going works: implementing the architecture

- Models are implemented and simulated with Renew
- Renew can be used as the control logic and to execute commands for various purposes:
 - Execute any UNIX commands
 - Launch any script/executable (Python, Java, C,...)
- OpenNebula is used to deploy VMs
- Renew can call OpenNebula command to launch initial VMs, launch additional VMs on the same cloud (local resources), launch VMs on a remote cloud (remote resources), ...
- Renew can execute Java program for sending, receiving, processing data chunks according to the control logic



Conclusions

Unified resource management with Token Bucket

- We extend our architecture with a unified and more flexible approach based to tokens management
- Redistribution of unused resources and redistribution of pre-allocated resources from low priority classes are **handled consistently** with the Token Bucket main regulation & isolation mechanism
- Business rules can complement optimization process to improve revenue

Revenue-centric model

- Adding new resources (local or remote) as well as optimizing preallocated resources can be classified according to their cost
- Global revenue taking into account all possible resource-related actions can be maximized when using the proposed unified resource magement approach



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Revenue creation for rate adaptive stream management in multitenancy environments



GECON 2013