

# Automatic Virtual Machine Clustering based on Bhattacharyya Distance for Multi-Cloud Systems

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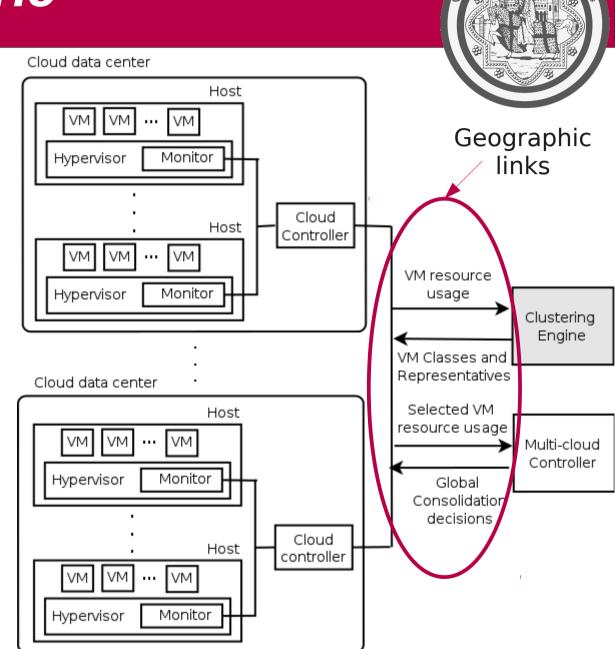
# **Cloud computing challenges**

- Large data centers (> 10<sup>5</sup> VMs)
  → huge amount of data
- Multiple data centers
  - → geographic data exchange
- → Scalability problems
- Current approach reduce amount of data in a uniform way:
  - Reduce sampling frequency
  - Reduce number of metrics considered
- → Reduced monitoring effectiveness
  - Less information available to take management decision



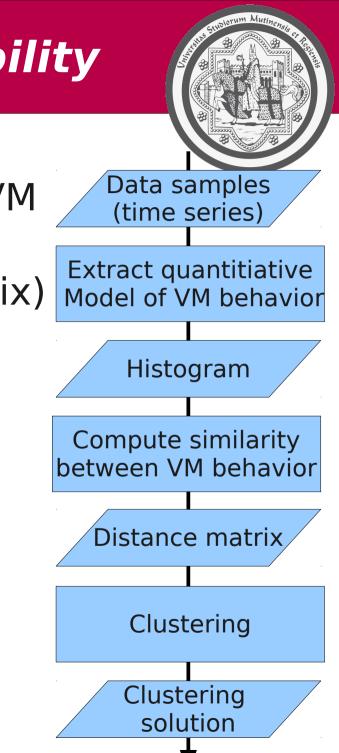
## Reference scenario

- IaaS with long term commitment
- Reactive VM relocation
  - Local scope
  - Overload mgm
- Periodic global consolidation
  - Global scope
  - Server mgm



# Impact on monitoring scalability

- Methodology:
  - Define quantitative model for VM behavior
  - Define VM similarity (dist. matrix)
  - Cluster similar VM together
- Elect a few (e.g., 3) cluster representatives
- Fine-grained monitoring of cluster representatives
- Reduced monitoring applied to other VMs
  - Reduced number of metrics
  - Lower sampling frequency



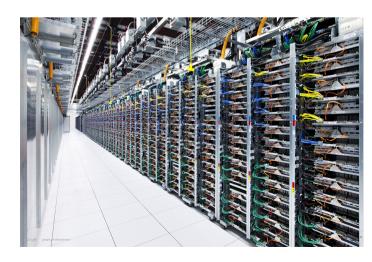
# Impact on monitoring scalability

- Case study:
  - E-health, Web-based application
  - Deployed on cloud IaaS
- Numeric example:
  - 110 VMs, K metrics, sampling frequency: 5 min.
    - $\rightarrow$  ~3.2 10<sup>4</sup> K samples/day

- 2 classes, 3 rep. per class  $\rightarrow \sim 2.1 \ 10^3$  K samples/day
- → Monitoring data reduced by 1 order of magnitude

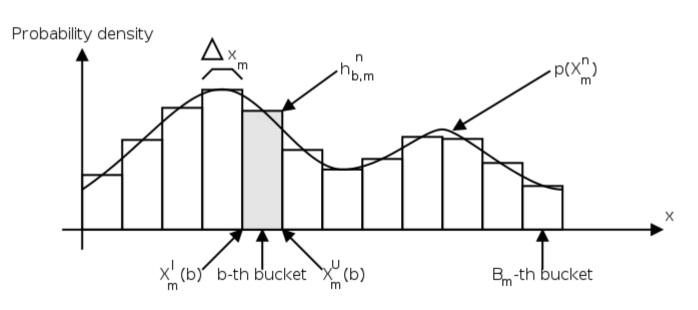


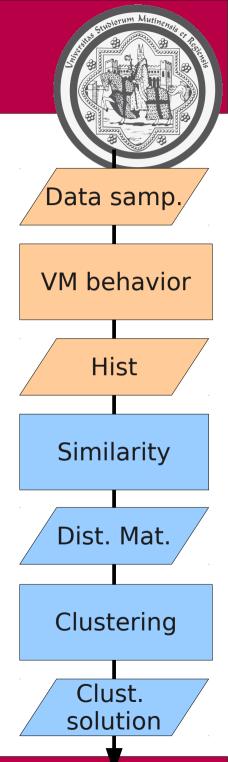




## Modeling VM behavior

- Model based on probability distribution of resource usage
  - Multiple resources considered (metrics)
- Histogram for every metric, every VM
  - Normalized histogram ( $\Sigma$ h=1)
  - B: number of buckets (critical)





#### **Defining VM similarity** Use of Bhattacharyya distance Data samp - Determine distance matrix for each couple of VMs, each metric VM behavior Euclidean combination of distance matrices Hist Sum of squares of multiple distances Similarity p(X<sup>n1</sup><sub>m</sub>) 0.5 0.5 $D_{m}(n1,n2) = -\ln(0.25 + 0.5 + 0.25)$ $D_{n}(n1,n2) = -\ln(0)$ $=-\ln(1)$ 0.25 0.25 0.25 0.25 Dist. Mat. 0 0 0 Х 0.5 0.5 p(X<sup>n2</sup>) Clustering 0.25 0.25 0.25 0.25 0 0 0 0 0 Х Clust. solution

# **Clustering algorithm**

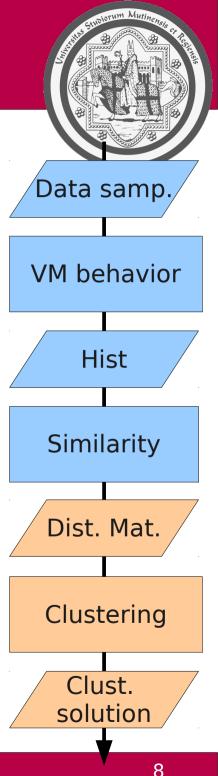
- Use of spectral clustering algorithm
  - Input: Square, symmetric distance matrix
  - Output: Cluster ID for every VM

## Additional feature:

 Number of clusters can be automatically determined through spectral gap analysis

## Open problems:

- Is it correct to consider every metric together?
- Is there a way to select the *right* metrics?



# **Choosing the right metrics**



- Multiple metrics are merged into the final distance matrix
- Not every metric provide significant information
- Proposal to identify relevant metrics
  - Consider auto-correlation: ACF decreasing rapidly  $\rightarrow$  random variations
  - Consider Coefficient of Variation: CF  $\gg$  1  $\rightarrow$  spiky and noisy behavior CF  $\ll$  1  $\rightarrow$  little information provided

### • $\rightarrow$ Merge information from metrics with

- ACF decreasing slowly
- CF ~ 1

## Case study



- IaaS cloud supporting e-health
  - Web server and DBMS
  - 110 VMs
  - 10 metrics for each VM,
  - Sampling frequency: 5 min

#### Goal: separate Web servers and DBMS

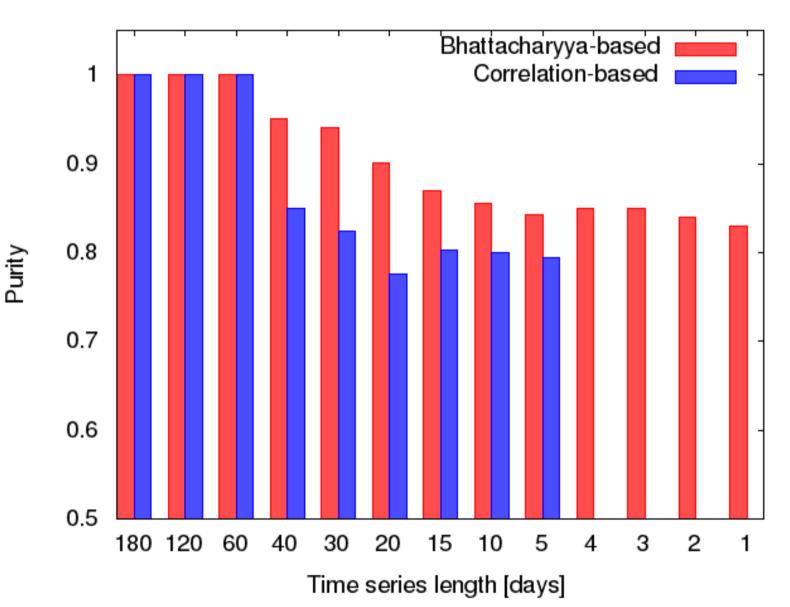
- Main metric: Purity of clustering

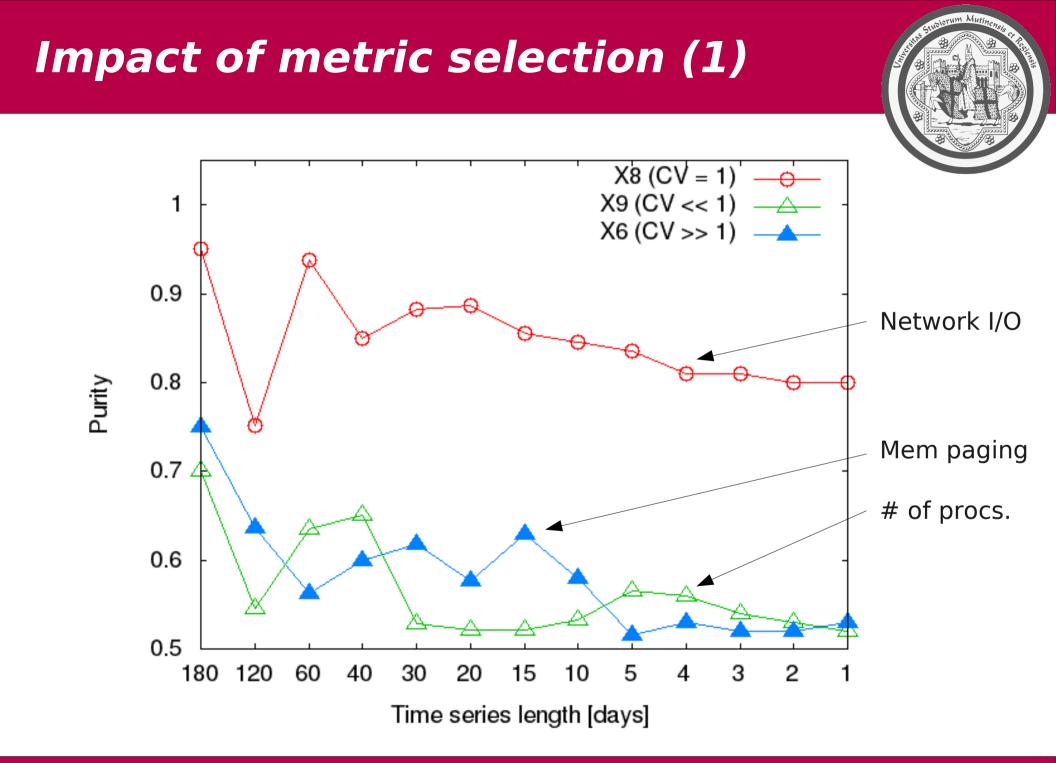
#### Three types of analyses

- Impact of time series length
- Impact of metric selection techniques
- Impact of histogram characteristics

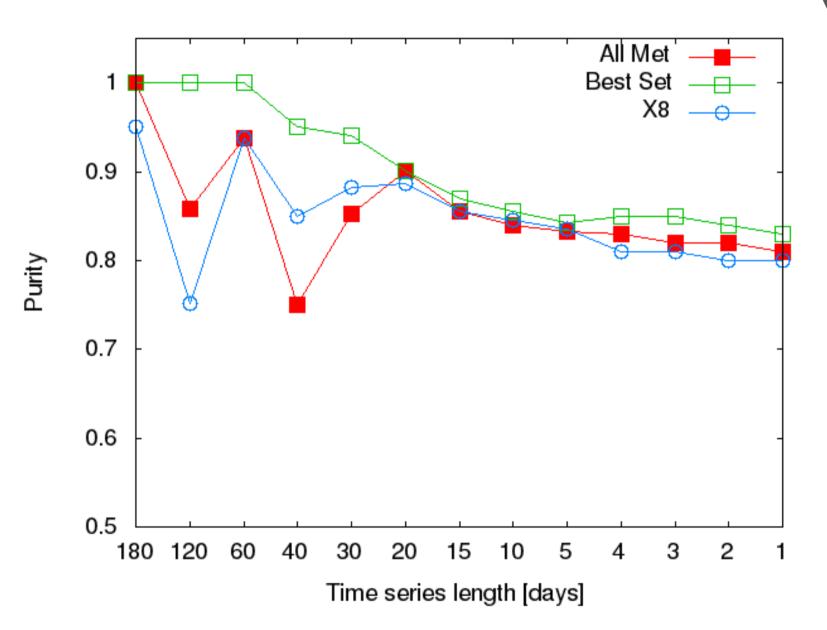
## Impact of time series length





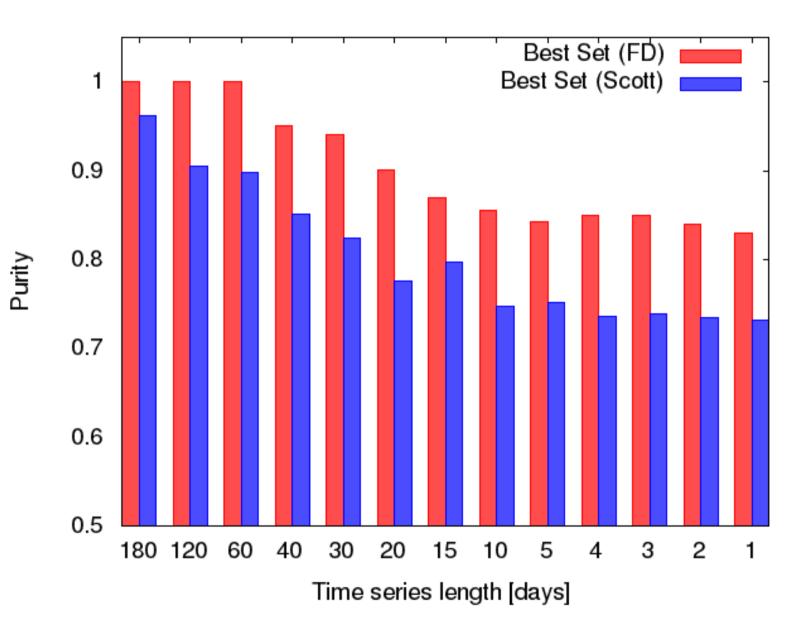


## Impact of metric selection (2)



MultiCloud - 22 april 2013 - Prague

### Impact of histogram characteristics



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# **Conclusion and future work**



- Scalability in (multi)cloud systems
  → open issue
- Proposal of novel methodology to improve scalability through clustering of similar VMs
- Experimental results are encouraging
  Purity >0.83 even for very short time series
- Future research directions:
  - Validation with more data set (Help!)
  - Improving stability of the results w.r.t histogram parameters
  - Evaluate different models for VM behavior
  - Application of clustering to improve scalability of VM management



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