

# Modelling and control of BigData services

Application to MapReduce performance and  
dependability

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PhD Defense



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Bogdan Robu



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# CONTEXT

## Cloud computing

- Cloud services are everywhere



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# CONTEXT

## Cloud computing

- Cloud services are everywhere



- More and more BigData cloud services



**MapReduce**



# CHALLENGES IN THE CLOUD

## Performance issues

- Recent survey from Compuware, 468 CIOs and senior IT professionals

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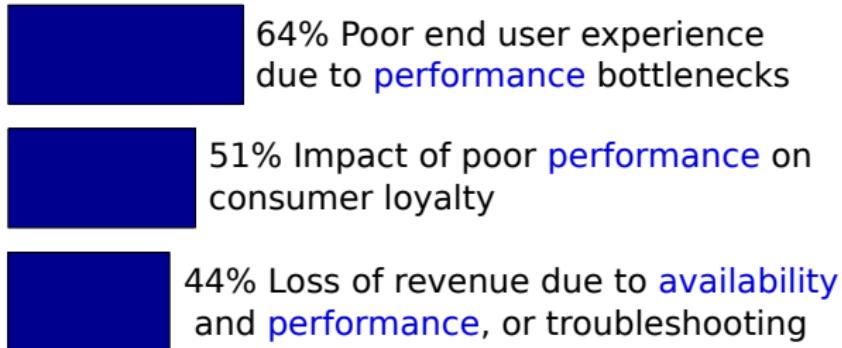
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# CHALLENGES IN THE CLOUD

## Performance issues

- Recent survey from Compuware, 468 CIOs and senior IT professionals

### Biggest concerns about managing cloud services?



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# CHALLENGES IN THE CLOUD

## Dependability issues

The screenshot shows the TechWeek Europe website. At the top, there is a navigation bar with the logo 'TechWeek europe' and a search bar. Below the navigation bar, there is a secondary menu with links: Menu, Mobility, Networks, Cloud, Security, Workspace, Projects, Events, and Tech Club. Underneath this, there is a sub-navigation bar with two tabs: 'CLOUD' and 'CLOUD MANAGEMENT'. The 'CLOUD' tab is highlighted.

## AWS Suffers Another Cloud Outage

Ben Sullivan, August 10, 2015, 3:15 pm



- September 2015, overloaded with requests

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# CHALLENGES IN THE CLOUD

## Dependability issues

The screenshot shows a news article from TechWeek Europe. The header includes the TechWeek Europe logo, a search bar, and navigation links for PERSONAL TECH, BIZ TECH, FUTURE TECH, SCIENCE, LIFE, and T-LOUNGE. The main title of the article is "Amazon Web Services Suffers Crash, Takes Down Netflix, Reddit, Tinder And Other Huge Parts Of The Internet". The article is by Remellaine Arsenio and was published on September 23, 2015, at 11:14 AM. Below the title, there are social sharing buttons for Like, Follow, Share, Tweet, Reddit, and Comments, along with a "SUBSCRIBE" button. A large image of the Amazon logo is displayed, with a caption below it stating: "The Amazon Web Services crashed Netflix, Reddit, Tinder". To the right of the image, a text block discusses the impact of the outage, mentioning that AWS powers many websites and that instead of building their own data infrastructures, marketers rely on cloud services like Amazon's.

A TAG Amazon Web Services , Tinder , Netflix , Reddit , AWS Services

## A Amazon Web Services Suffers Crash, Takes Down Netflix, Reddit, Tinder And Other Huge Parts Of The Internet

By Remellaine Arsenio, Tech Times | September 23, 11:14 AM

Like Follow Share(19) Tweet(?) Reddit 5 Comments ... SUBSCRIBE

The Amazon Web Services crashed Netflix, Reddit, Tinder

A monstrous outage from Amazon Web Services crashed down Netflix, Reddit, Tinder and other major websites, sending netizens in fury for missing movies, hook-ups and other fun online activities.

AWS powers web and mobile applications, and provides data processing and warehousing, storage and archiving to websites all over the world.

Instead of building their own data infrastructures, countless important online marketers place their trust on the cloud services offered by Amazon to lessen their long-term investments in the

- September 2015, overloaded with requests

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# CHALLENGES IN THE CLOUD

## Dependability issues

The screenshot shows a news article from CRN (Computer Reseller News) with the following details:

- Header:** TECH TIMES
- Sub-Header:** NEWS, ANALYSIS AND PERSPECTIVE FOR VARs AND TECHNOLOGY INTEGRATORS
- Article Title:** Azure Nightmare: Customer Suffers 9-Day Intermittent Outage, Gets No Help From Microsoft
- Author:** Kevin McLaughlin
- Date:** May 6, 2015, 7:20 pm EDT
- Content Summary:** A Microsoft Azure customer who experienced a nine-day service disruption last month is furious over the software giant's lack of clear and transparent communication about the issue.
- Social Sharing:** Facebook Like (75), Facebook Share (75), LinkedIn Share, Twitter Tweet (52), Google+ G+ (3), Google+ Share (3), and a Submit button.
- Related Content:**
  - SLIDE SHOWS:** 30 Notable IT Executive Moves: October 2015
  - Companies:** 5 Companies That Had A Rough Week
  - Review:** Review: GammaTech's Durabook S15AB Sports Big Display, Thin Body – And Is Built For Abuse

- September 2015, overloaded with requests

## PROBLEM STATEMENT

- Highly dynamic workloads
- Multiple criteria at the same time
  - performance, dependability, cost
- No fully automatic control solutions

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# QoS AND SLA

- Quality of Service (QoS)

- The ability to meet different requirements
  - e.g. performance, availability, cost

- Service Level Objectives(SLO)

- Specific, measurable service characteristics
  - e.g. Maximum response time 5s

- Service Level Agreement (SLA)

- Contract → formalises QoS
- Parties involved (roles), penalties
- Multiple SLOs
- e.g. Maximum response time 5s; Minimum availability 98%;

# MAPREDUCE

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- One of the most used BigData systems
  - Simple to use, scalable and fault-tolerant
  - Wide range of applications:

MapReduce is a programming model for processing large data sets.

It is designed to work across distributed systems.

It is a simple way to process large amounts of data.

It is a distributed system that can handle large amounts of data.

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# MAPREDUCE

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- Simple to use, scalable and fault-tolerant
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# MAPREDUCE

- One of the most used BigData systems
- Simple to use, scalable and fault-tolerant
- Wide range of applications:
  - log analysis, data mining, web search engines, scientific, computing, business intelligence

## MapReduce usage

Google	100.000 jobs/day
Yahoo	40.000 computers
Facebook	100 petabytes ( $10^{15}$ )
LinkedIn	$120 \cdot 10^9$ relationships/day

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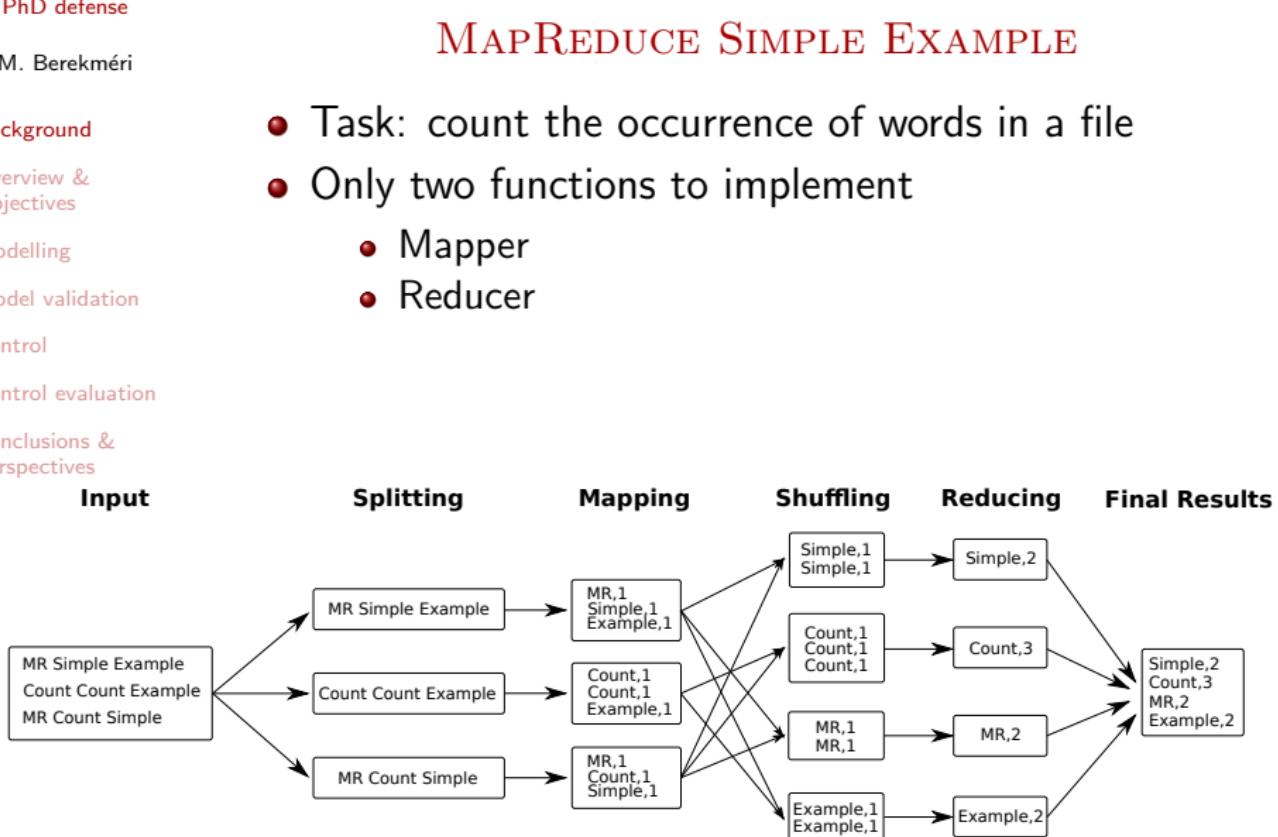
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# MAPREDUCE SIMPLE EXAMPLE

- Task: count the occurrence of words in a file
- Only two functions to implement
  - Mapper
  - Reducer



# MAPREDUCE QUALITY-OF-SERVICE

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## ● Service performance:

- response time - average time ( $y_{rt}$ ) to process a client request

$$y_{rt}[T] = \text{avg}(y_{rt_1}, y_{rt_2}, \dots, y_{rt_N})$$

## ● Service dependability:

- availability - system accessibility to users per unit of time

$$y_{av} \left[ \frac{\%}{T} \right] = \frac{N_{SuccessfulJobs}}{N_{SuccessfulJobs} + N_{RejectedJobs}} * 100$$

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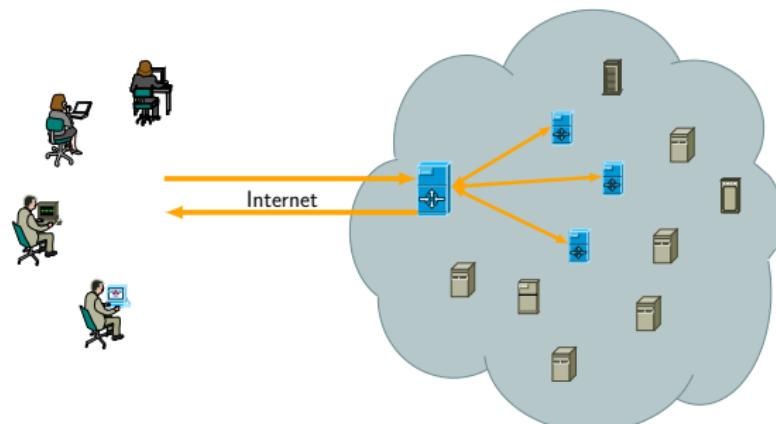
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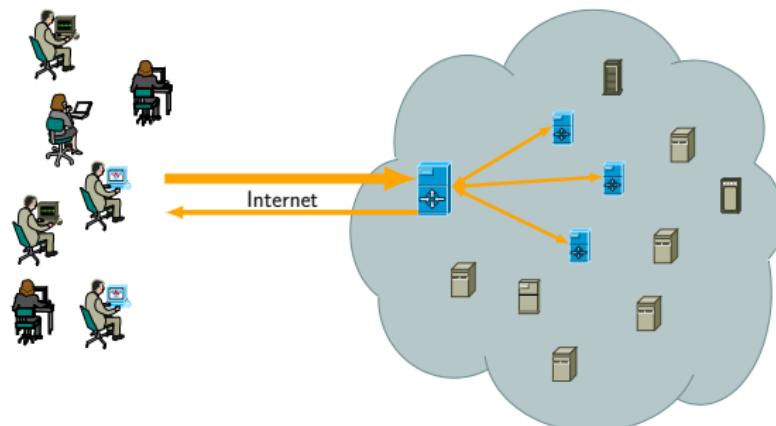
# MAPREDUCE PARAMETERS

- Many parameters (Hadoop → 170 )
  - most not configurable on-line
- Examples of on-line controllable parameters:
  - $N$  - number of processing nodes
  - $MC$ - maximum accepted client requests



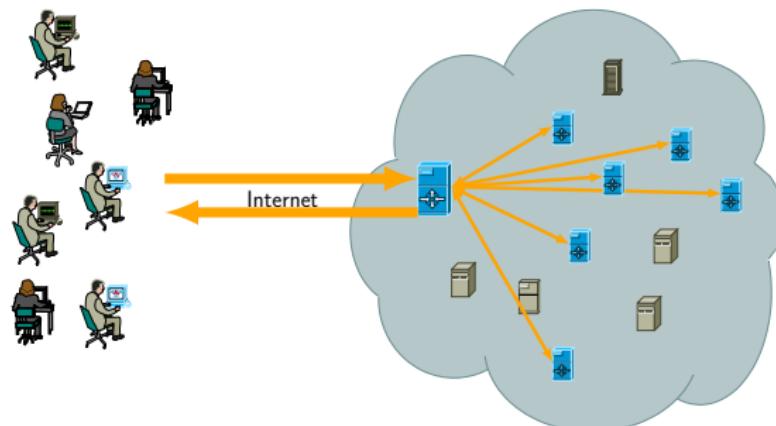
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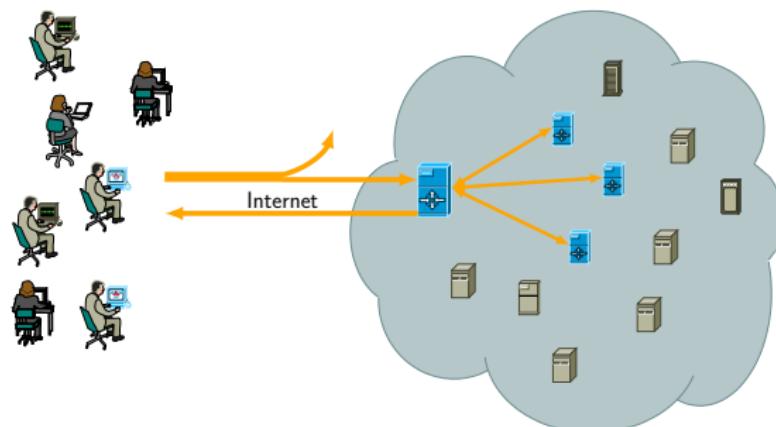
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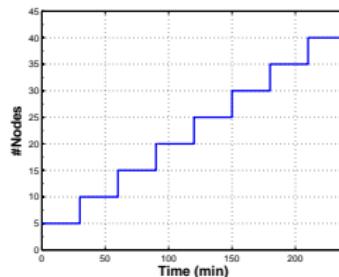
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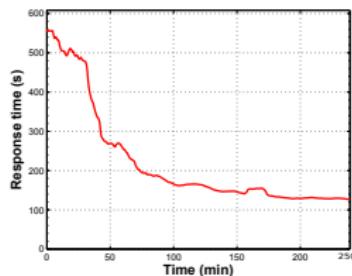


# MOTIVATING EXAMPLE

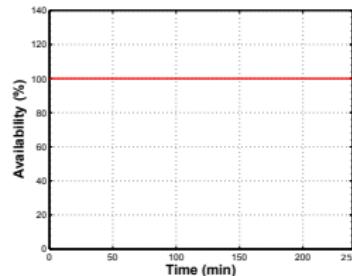
## Impact of nodes variation on performance and dependability



Nodes



Response time

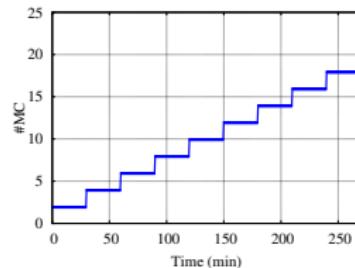


Availability

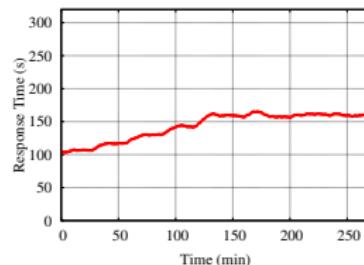
#MC=10, #Clients=10

# MOTIVATING EXAMPLE

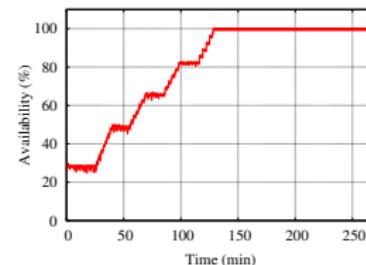
## Impact of max clients variation on performance and dependability



MaxClients



Response time



Availability

#Clients=10, #Nodes=20

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# CONTROL THEORETICAL APPROACH

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Develop and apply control strategies for MapReduce

- But why control theory?

- Unified mathematical framework
- Automated tools for modelling and control synthesis
- Proven algorithms

- Challenges:

- No physics behind algorithms, applications
- Difficult to use classical techniques

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## RELATED WORK

### Improving MapReduce

- Bottleneck management [Sharma, CLOUD'12]
- Data placement [ADAPT, ICDCS'12]
- Fault recovery [Ruiz, ICDE'11]
- Cost-based optimization [Herodotou, VLDB'11]

- ✗ No performance or dependability guarantees
- ✗ Heuristic, best-effort

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## RELATED WORK

- Modelling MapReduce performance

- Fine grained, analytical models
  - [Herodotos, VLDB'11]
  - [Vianna, IJPP'13]
- Course grained, regression models
  - [Verma, Middleware'11]
  - [Xu, IP&DPS'12]

 Single-output models

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## RELATED WORK

- Controlling MapReduce performance
  - [SteamEngine, HiPC'11]
  - [ARIA, ICAC'11]
  - [Jockey, ACM'12]

✗ Don't ensure multiple objectives

## RELATED WORK

### Open issues

- Multi-input, Multi-output (MIMO) models
- Controllers capable of ensuring multiple SLOs at the same time
- Handling the trade-off between contradictory SLOs

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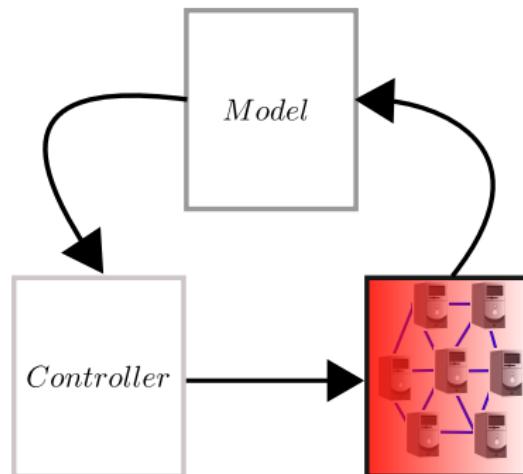
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# OBJECTIVES

- Experimental MapReduce environment
  - MIMO MapReduce model
  - Controllers → ensure a multi-objective SLA
    - Performance and dependability SLOs, meanwhile minimising cost
    - Faced with dynamic client variations



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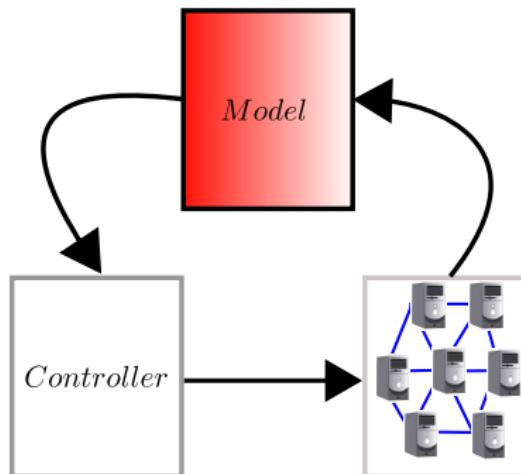
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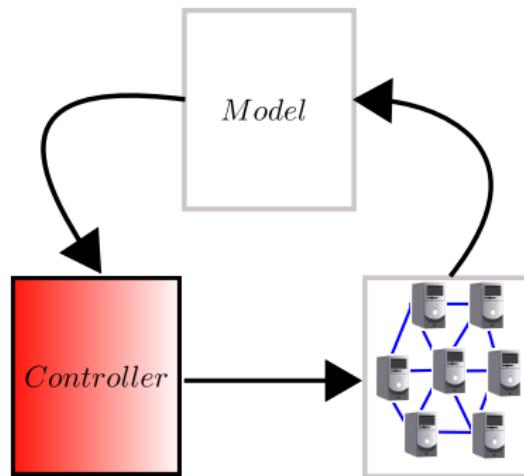
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# OBJECTIVES

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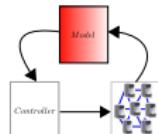
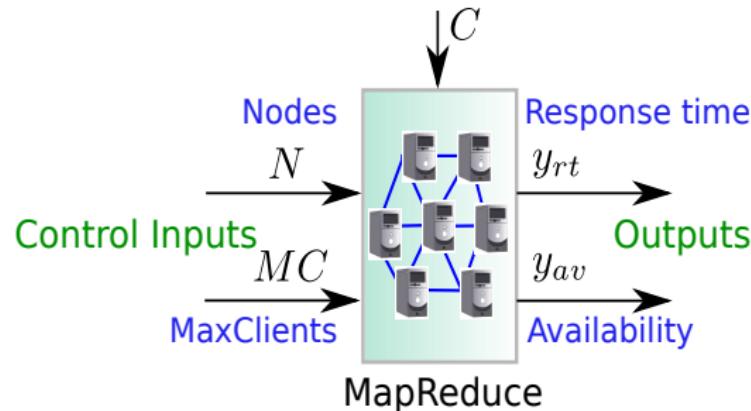
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# MODEL STRUCTURE

**Exogeneous Inputs****Clients**

- $N$  number of nodes in the cluster
- $MC$  maximum number of concurrent clients
- $C$  number of concurrent clients

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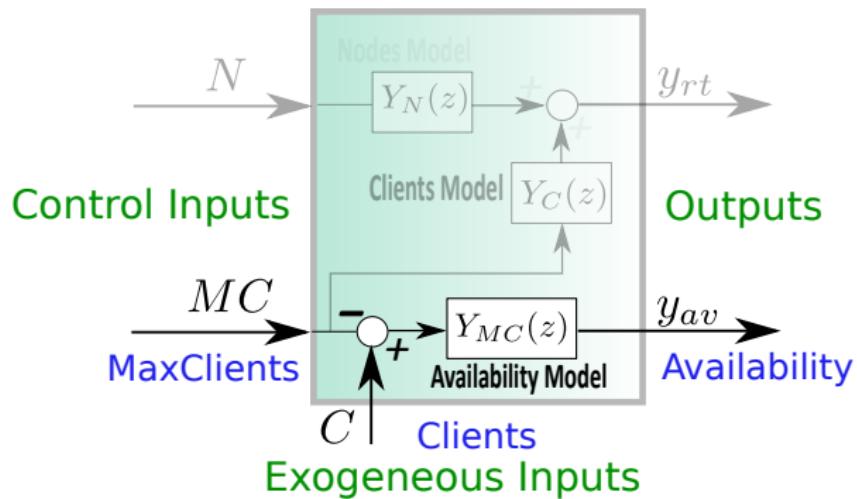
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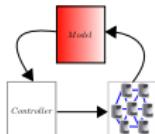
# MAPREDUCE DEPENDABILITY MODEL



- Mathematical relation between number of clients, max clients and availability

$$y_{av} = Y_{MC}(z) \cdot (C - MC)$$

where:  $MC \leq C$



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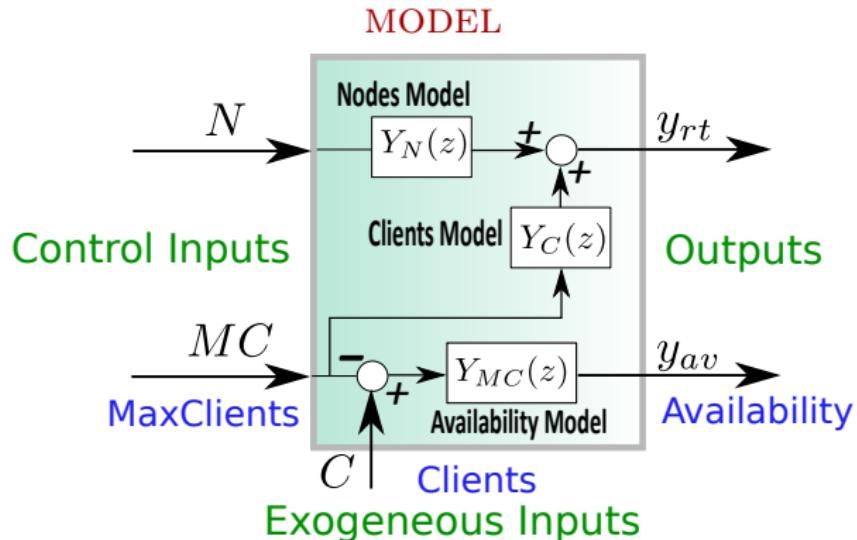
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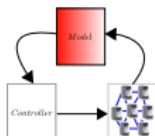
# MAPREDUCE PERFORMANCE AND DEPENDABILITY



$$y_{rt} = Y_C(z) \cdot MC + Y_N(z) \cdot N$$

$$y_{av} = Y_{MC}(z) \cdot (C - MC)$$

where:  $MC \leq C$



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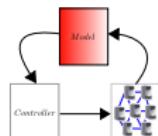
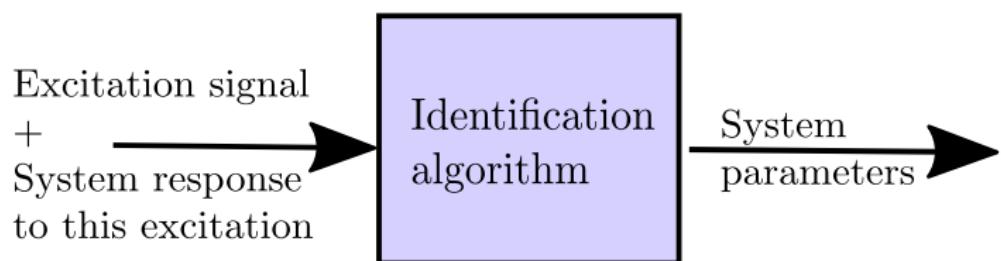
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# HOW TO DETERMINE THE MODEL STRUCTURE AND PARAMETERS?

- Automated off-line identification and on-line adaptation tools



- ① Prediction error method → off-line identification
- ② Recursive least square → on-line adaptation

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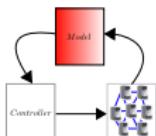
# TOWARDS IDENTIFYING MODEL PARAMETERS

## ● Transfer function formulation

$$y_{rt} = Y_C(z) \cdot MC + Y_N(z) \cdot N$$

$$y_{av} = Y_{MC}(z) \cdot (C - MC)$$

## ● State space formulation - matricial, compact



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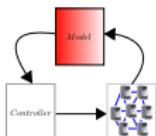
# TOWARDS IDENTIFYING MODEL PARAMETERS

## ● Transfer function formulation

$$y_{rt} = z^{-\tau_{rtc}} \frac{b_c}{(z + a_c)} \cdot MC + Y_N(z) \cdot N$$

$$y_{av} = Y_{MC}(z) \cdot (C - MC)$$

## ● State space formulation - matricial, compact



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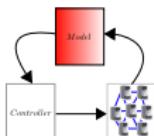
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$$y_{rt} = z^{-\tau_{rtc}} \frac{b_c}{(z + a_c)} \cdot MC + z^{-\tau_{rtn}} \frac{b_n}{(z + a_n)} \cdot N$$

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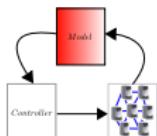
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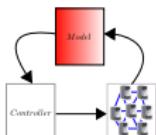
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- State space formulation - matricial, compact

$$x_{k+1} = A_d \cdot x_k + B_d \cdot u_k$$

$$y_k = C_d \cdot x_k$$



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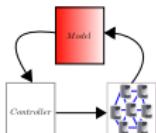
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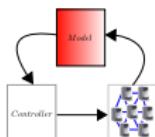
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- Transfer function formulation

$$y_{rt} = z^{-\tau_{rtc}} \frac{b_c}{(z + a_c)} \cdot MC + z^{-\tau_{rtn}} \frac{b_n}{(z + a_n)} \cdot N$$

$$y_{av} = \frac{b_{mc}}{(z + a_{mc})} \cdot (C - MC)$$

- State space formulation - matricial, compact



$$\begin{aligned}x_{k+1} &= A_d \cdot x_k + B_d \cdot u_k \\y_k &= C_d \cdot x_k\end{aligned}$$

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# TOWARDS IDENTIFYING MODEL PARAMETERS

- Transfer function formulation

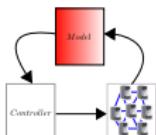
$$y_{rt} = z^{-\tau_{rtc}} \frac{b_c}{(z + a_c)} \cdot MC + z^{-\tau_{rtn}} \frac{b_n}{(z + a_n)} \cdot N$$

$$y_{av} = \frac{b_{mc}}{(z + a_{mc})} \cdot (C - MC)$$

- State space formulation - matricial, compact

$$x_{k+1} = A_d \cdot x_k + B_d \cdot u_k$$

$$y_k = C_d \cdot x_k$$



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# EXPERIMENTAL SETUP

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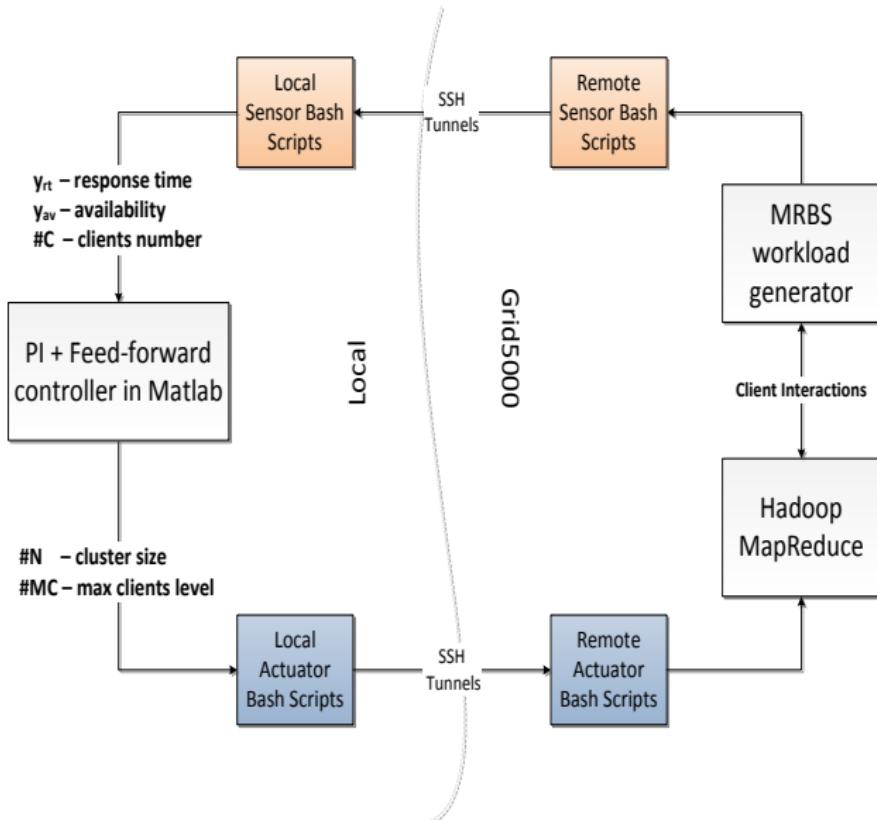
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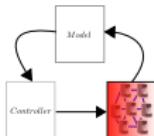
## EXPERIMENTAL SETUP

### • MRBS

- Performance and dependability benchmark suite
- Realistic multi-user workloads
- Data intensive business intelligence benchmark
- Clients request: one or more jobs( $\sim 10\text{GB}$  )
- French nationwide research cluster: Grid5000

### Node configuration

Quad-core Intel 2.53GHz CPU	15GB RAM
Infiniband 20G network	298GB disk



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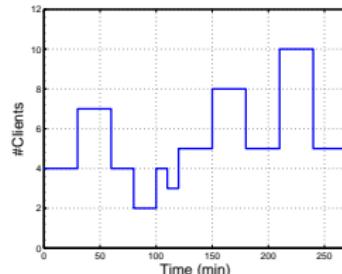
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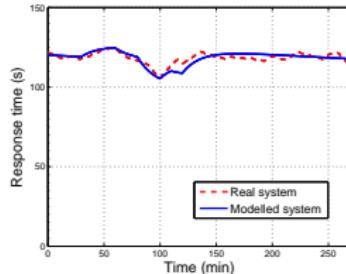
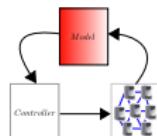
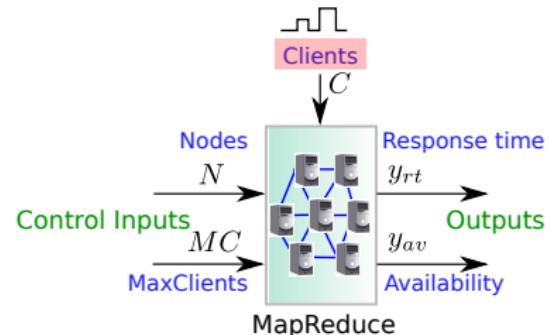
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# MODEL VALIDATION

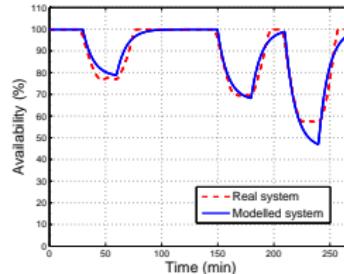
## Accuracy in capturing clients variation



Clients



Response time



Availability

#MC=5, #Nodes=20

Background

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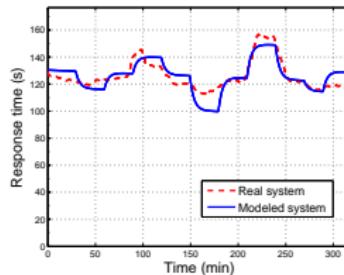
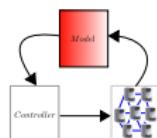
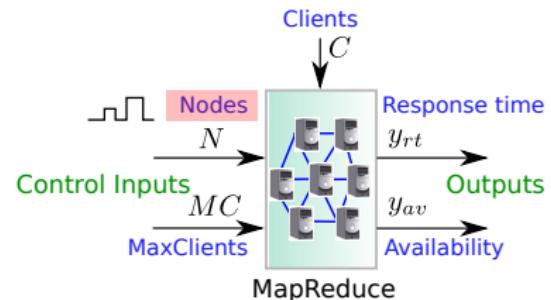
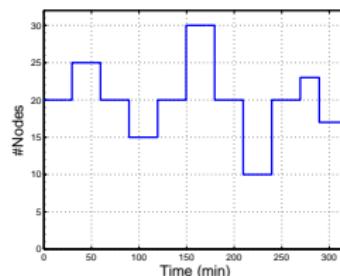
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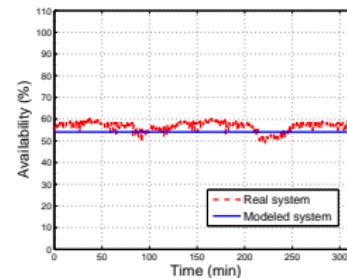
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# MODEL VALIDATION

## Accuracy in capturing nodes variation



Response time



Availability

#MC=5, #Clients=10

Background

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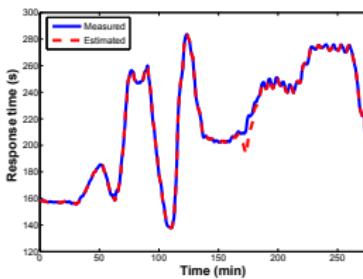
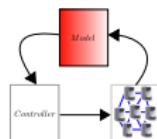
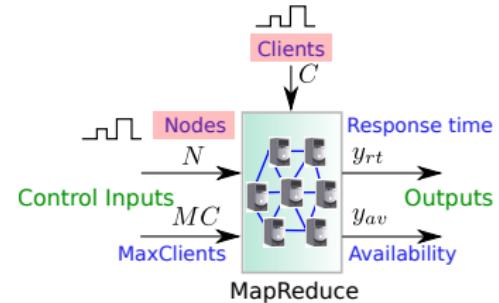
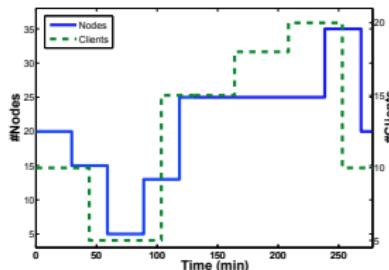
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# ON-LINE PARAMETER ADAPTATION VALIDATION

## Accuracy in capturing nodes and clients variation



Response time

#MC=30

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# CONTROLLERS

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## ● 3 control types

### ● *MR-Perf*

- Time-based feedback-feedforward controller
- Ensures performance

### ● *MR-Perf-Cost*

- Event-based feedback-feedforward controller
- Ensures performance, minimal costs

### ● *MR-Ctrl*

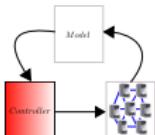
- Constrained optimal control
- Ensures performance and dependability, while minimising costs

# MR-Perf-Cost

## EVENT-BASED CONTROL

- SLA:

- Keep response time under a given threshold
- Minimal costs (minimal node count and variation)

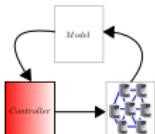


# MR-Perf-Cost

## EVENT-BASED CONTROL

### SLA:

- Keep response time under a given threshold
- Minimal costs (minimal node count and variation)

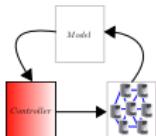
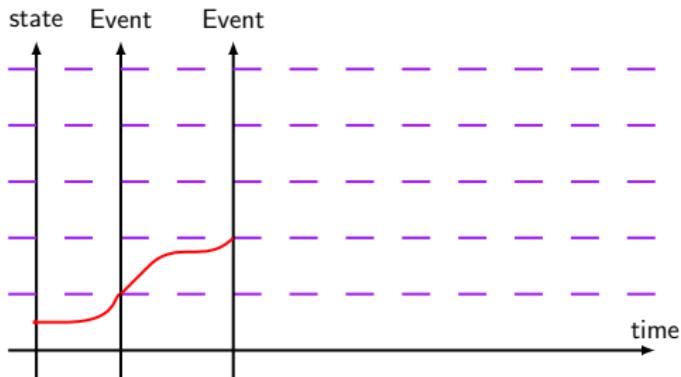


# MR-Perf-Cost

## EVENT-BASED CONTROL

- SLA:

- Keep response time under a given threshold
- Minimal costs (minimal node count and variation)

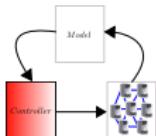
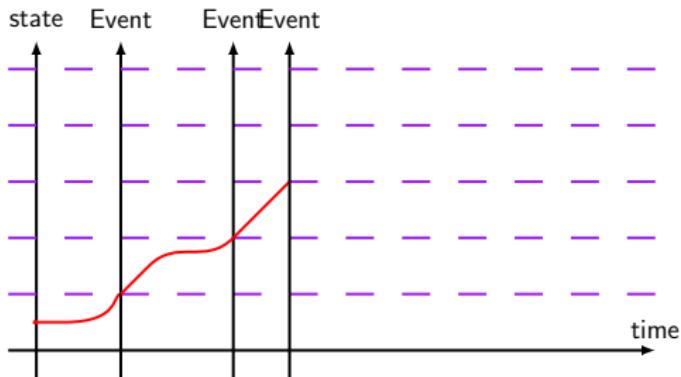


# MR-Perf-Cost

## EVENT-BASED CONTROL

### SLA:

- Keep response time under a given threshold
- Minimal costs (minimal node count and variation)

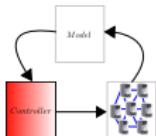
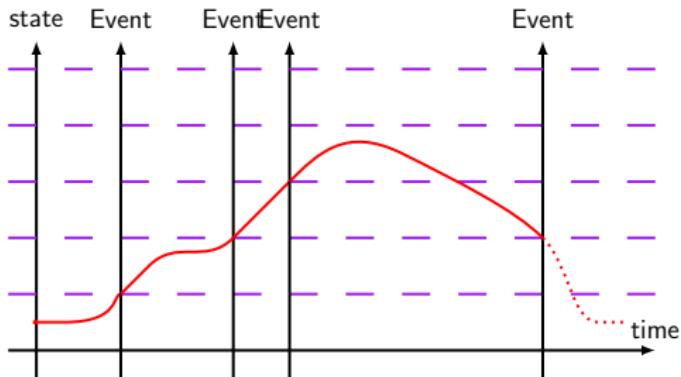


# MR-Perf-Cost

## EVENT-BASED CONTROL

### SLA:

- Keep response time under a given threshold
- Minimal costs (minimal node count and variation)



*MR-Perf-Cost*

Background

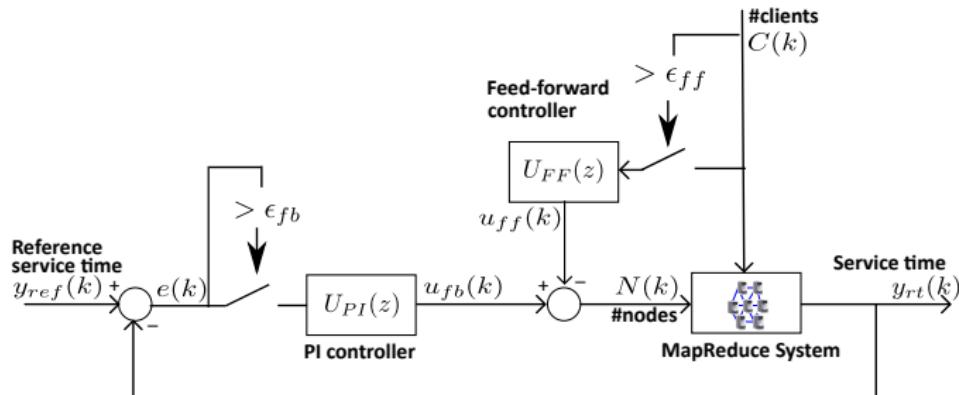
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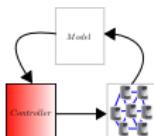
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- Event-based PI control

$$u_{fb}(k) = u_{fb}(k-1) + K_p \cdot (e(k) - e(k-1)) + K_i \cdot h \cdot e(k)$$

- Event-based feedforward control

$$u_{ff}(k) = -\frac{b_C}{b_N} \cdot C$$



*MR-Perf-Cost*

Background

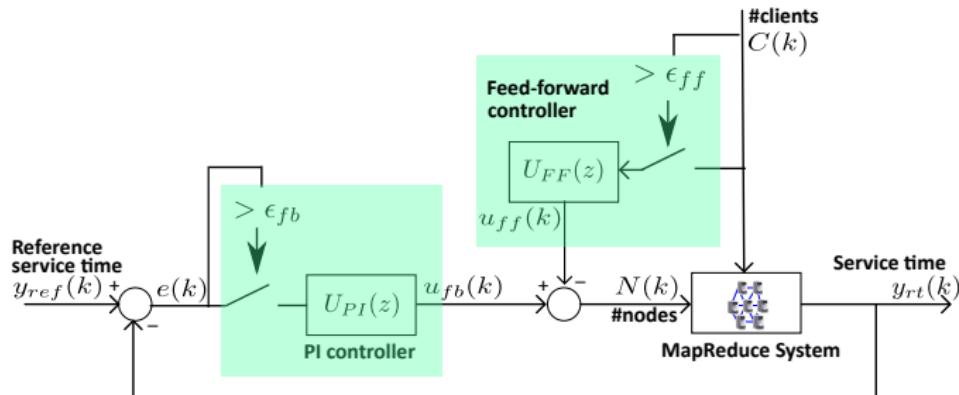
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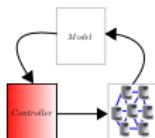
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$$u_{fb}(k) = u_{fb}(k-1) + K_p \cdot (e(k) - e(k-1)) + K_i \cdot h \cdot e(k)$$

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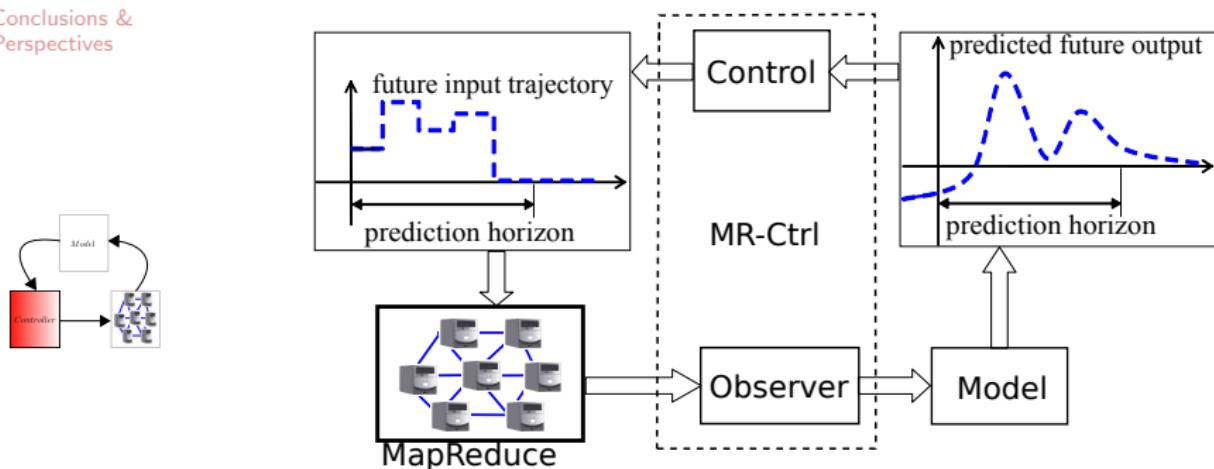
Control evaluation

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## Constrained optimal control

## ● SLA:

- Keep response time under a given threshold
- Keep availability above a given threshold
- Minimal costs (node count and variation)



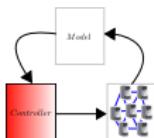
*MR-Ctrl*

## Optimisation strategy

- Objective:
  - Optimal control trajectory to ensure SLA
- Optimisation based on objective function:

$$J = \min_{U_k} \{ (Y_k - Y_{ref})^T \cdot Q \cdot (Y_k - Y_{ref}) + U_k^T \cdot R \cdot U_k \}$$

- Subject to constraints:  $MC \leq C$  and  $N \leq N_{max}$
- Quadratic programming → Optimal Control Synthesis



$$J = \min_u \{ u^T \cdot H \cdot u + 2 \cdot f^T \cdot u \}$$

$$H = \Delta^T \cdot Q \cdot \Delta$$

$$f^T = 2 \cdot (X_k^T \cdot \Gamma^T \cdot Q \cdot \Delta - Y_{ref}^T \cdot Q \cdot \Delta)$$

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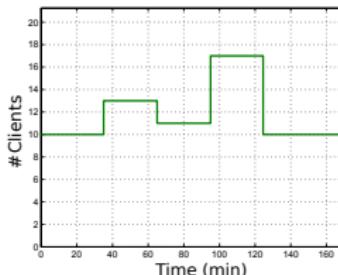
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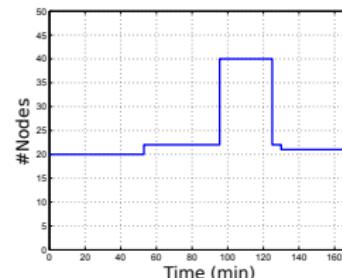
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# MR-Perf-Cost

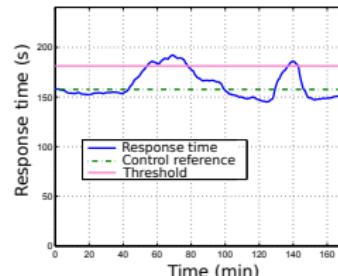
## Experimental validation



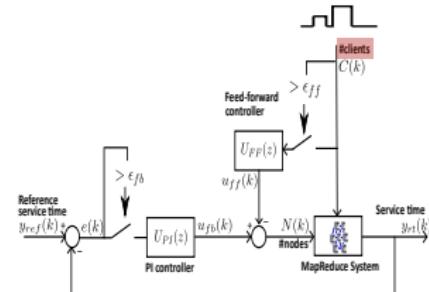
Clients



Nodes



Response time



## *MR-Perf-Cost*

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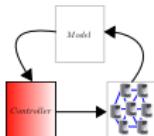
✓ Can handle highly dynamic workloads

✓ Automatic control solution

✓ Minimal costs

✓ Easy to implement

✗ Single output objectives only



# *MR-Perf-Cost*

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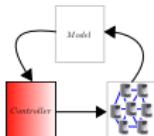
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## *MR-Perf-Cost*

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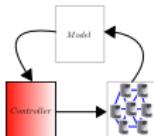
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# *MR-Perf-Cost*

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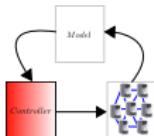
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## *MR-Perf-Cost*

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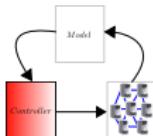
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- ✓ Can handle highly dynamic workloads
- ✓ Automatic control solution
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- ✗ Single output objectives only



*MR-Ctrl*

## Validation scenario

- Force trade-off between two objectives → #N=40

Background

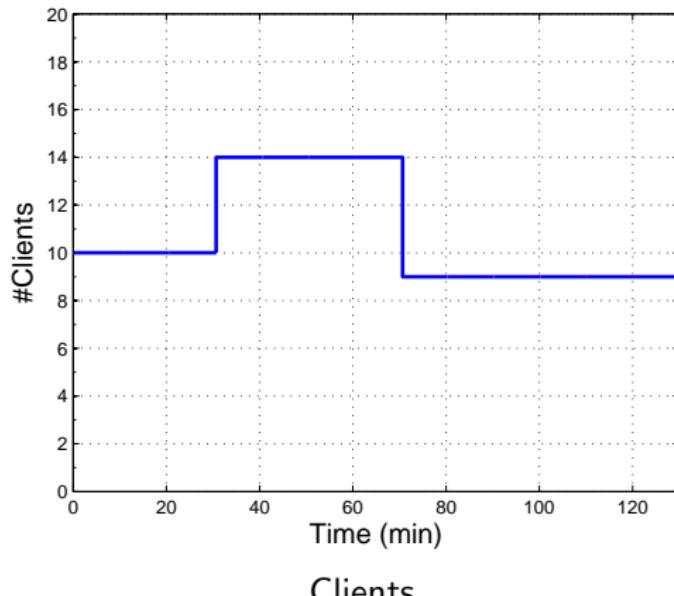
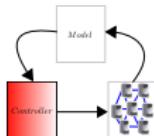
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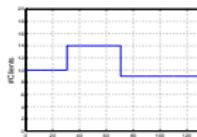
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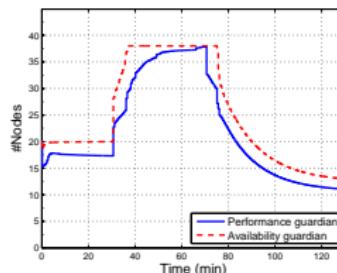
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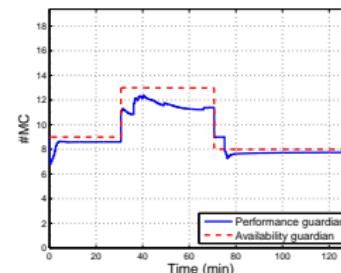
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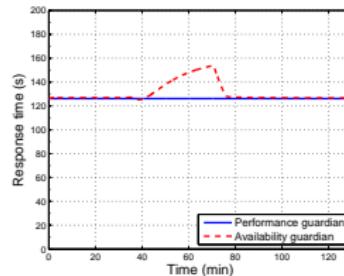
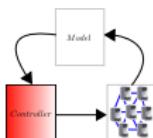
## Validation in simulation



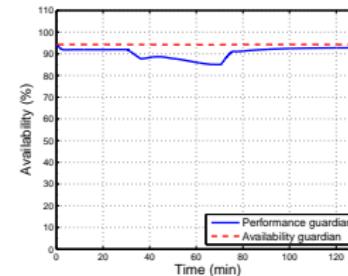
Nodes



MaxClients



Response time



Availability

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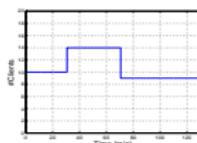
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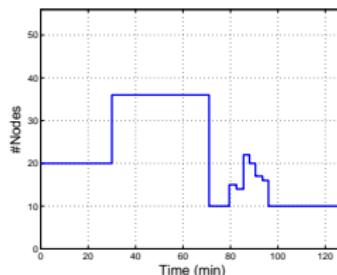
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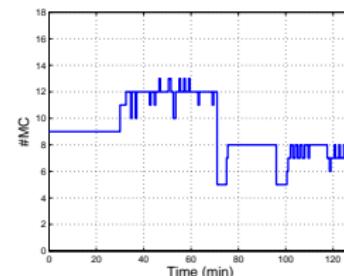
Conclusions &amp; Perspectives

*MR-Ctrl*

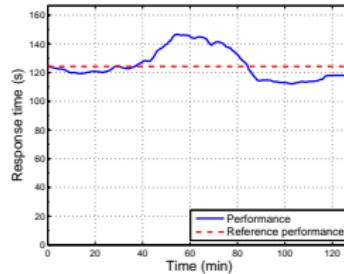
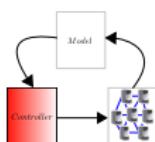
## Experimental validation



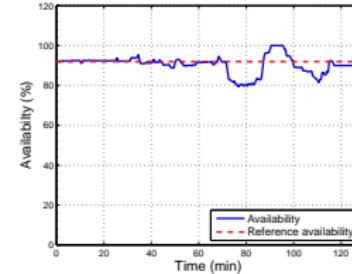
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## ✓ Handle highly dynamic workloads

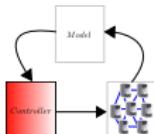
✓ Ensure multiple objectives at the same time

✓ Trade-off between contradictory objectives easily quantified

✓ Automatic control solution

✓ Explicit cost minimisation criteria

✓ Handle varying control or output constraints



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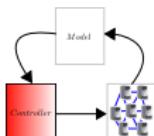
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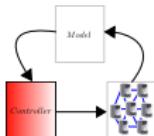
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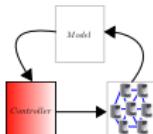
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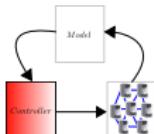
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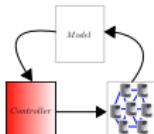
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- ✓ Handle highly dynamic workloads
- ✓ Ensure multiple objectives at the same time
- ✓ Trade-off between contradictory objectives easily quantified
- ✓ Automatic control solution
- ✓ Explicit cost minimisation criteria
- ✓ Handle varying control or output constraints



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# CONCLUSIONS

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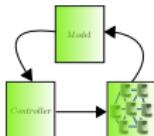
- Our model:

- Captures the system dynamics with high accuracy
- On-line adaptation of the parameters is recommended

- Our control:

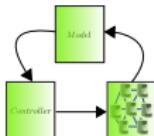
- Handles dynamic workloads
- Successfully ensures a multi-objective SLA

- General approach



# MAIN CONTRIBUTIONS

- ① An autonomous on-line control framework, that allows to measure and control the performance and availability of a MapReduce cluster
- ② Novel approach for the multi-input multi-output modelling of BigData services
- ③ Automatic control algorithms → ensures both performance and dependability objectives, while minimising costs



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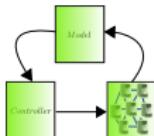
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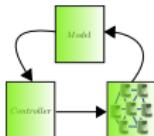
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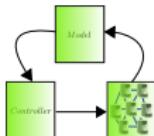
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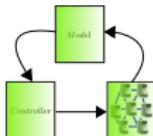
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- Other QoS metrics: e.g. throughput, reliability
  - Other service aspects: e.g. security
  - Other BigData services: e.g. YARN, SPARK
  - Other infrastructures: e.g. systems of systems (EU Project AMADEOS)
  - Package of off the shelf control solutions



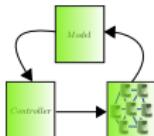
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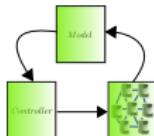
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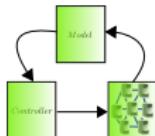
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# MAIN RESULTS

- International conference papers:

- “Application du contrôle pour garantir la performance des systèmes Big Data”, ComPAS 2014, Neuchâtel, Switzerland, April 22-25, 2014
- “A Control Approach for Performance of Big Data Systems”, IFAC 2014, Cape-Town, South Africa, August 24-29, 2014

- Journal papers:

- “Feedback Autonomic Provisioning for Guaranteeing Performance in MapReduce Systems” IEEE Transactions on Cloud Computing (accepted)
- “Event-Based Control in BigData Cloud Systems: Application to MapReduce” IEEE Transactions on Industrial Informatics (submitted)

