

# CASE STUDY: COST CALCULATOR FOR CLOUD APPLICATIONS

Background:

Salánki, Á., Kincses, G., Gönczy, L. and Kocsis, I., 2017. Data analysis-based capacity planning of VCL clouds. *International Journal of Cloud Computing*, 6(4), pp.370-383.

# Motivation

Enterprise cloud  
Purchased CPU time Azure™



Lab

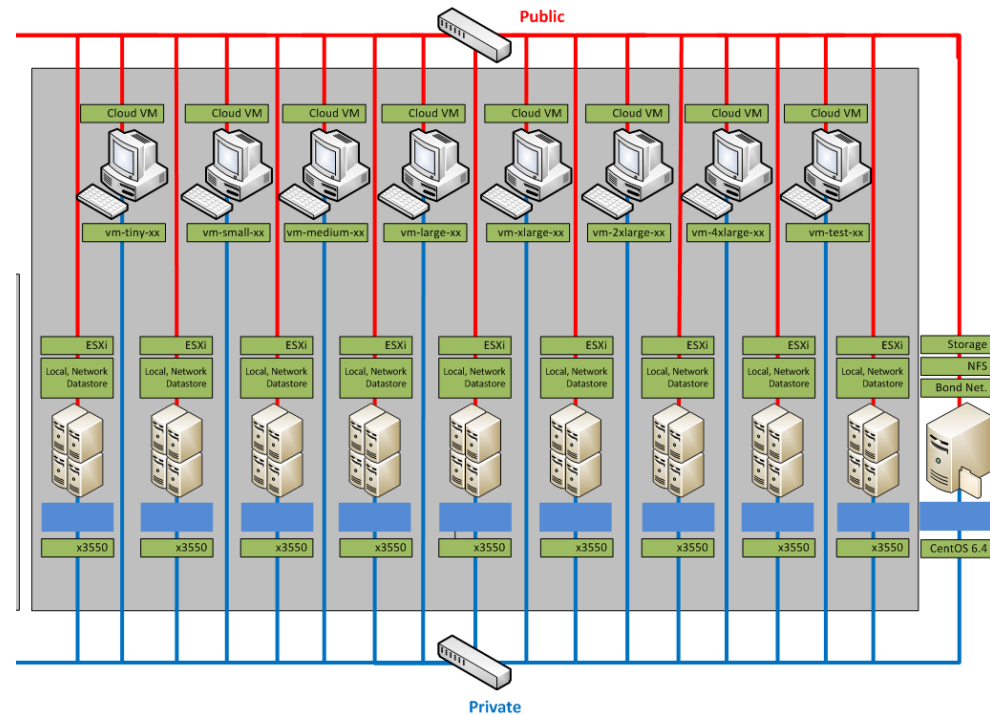


Private  
university cloud



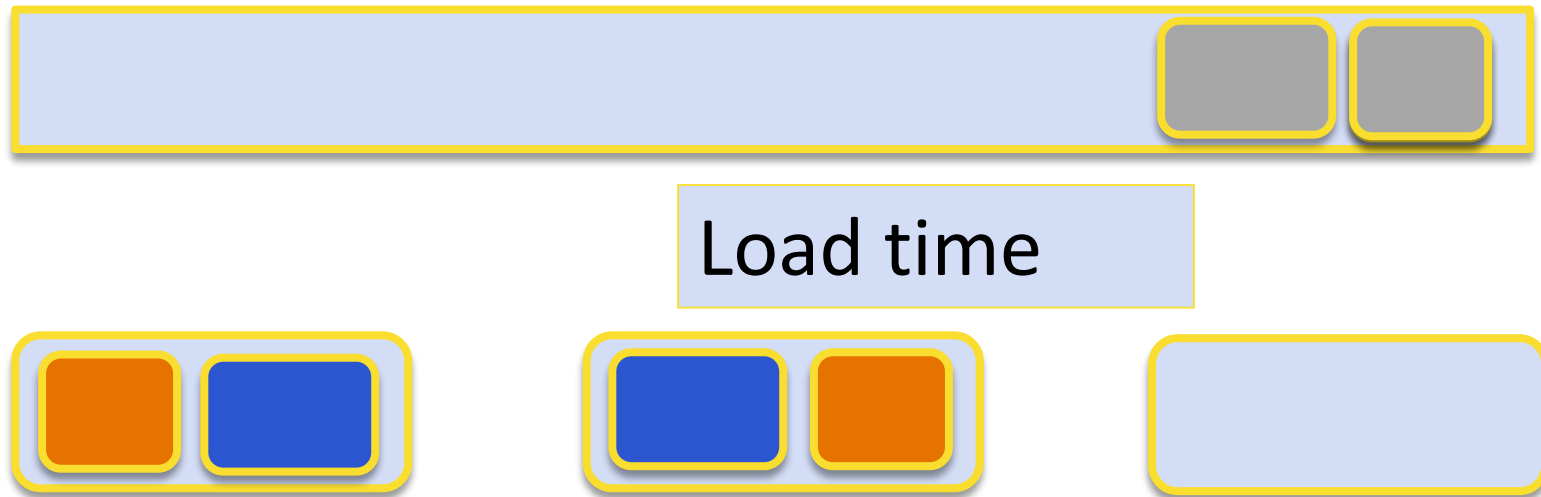
# Our VCL cloud

- Maintained by our research group
- 5 semesters
  - 2 courses/semester
- 9 hosts
- ~20 000 reservations
  - Only 22 rejected

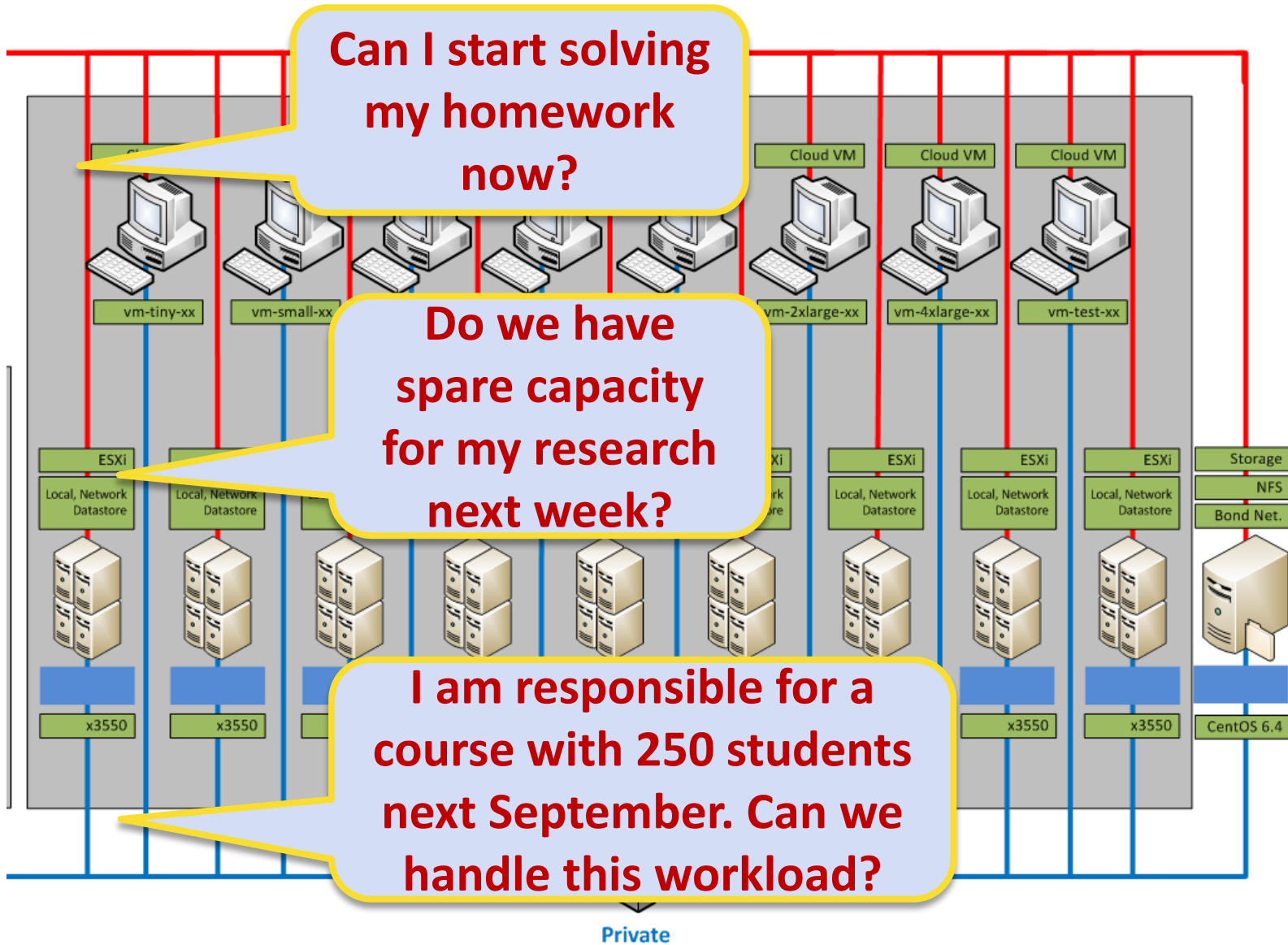


# Reservation Workflow in VCL

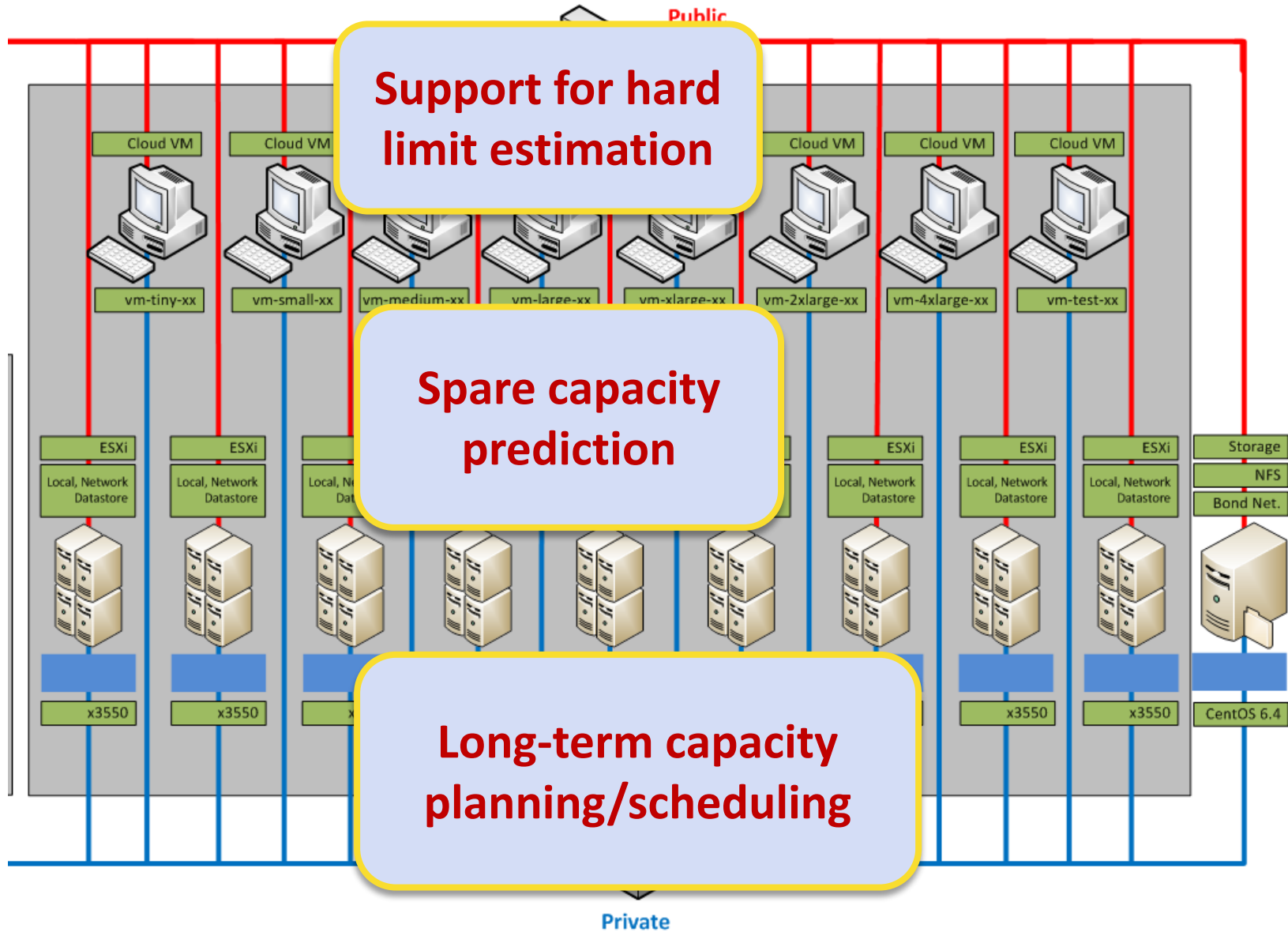
- Request
  - VM type
  - Length
  - Immediately or later
- Hard reservation limit



# Capacity Planning



# Capacity Planning



# The Available Dataset



deadlines, #students



reservation type, time to load, etc.



VM1

VM2

Host2

VM1

VM2

Host1

cpu usage, memory usage, etc.

cpu usage, memory usage, etc.



# Data Analysis Steps



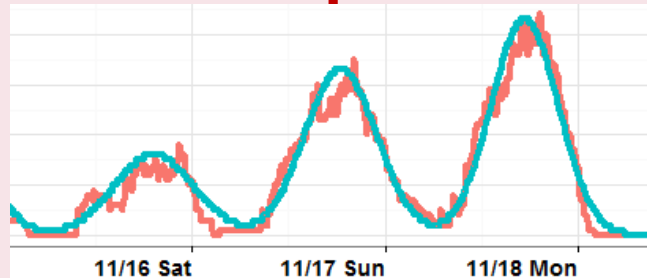
Host2



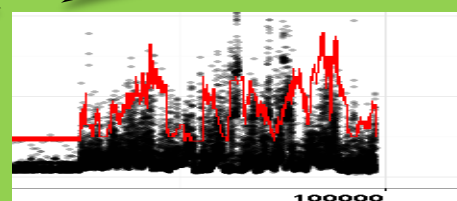
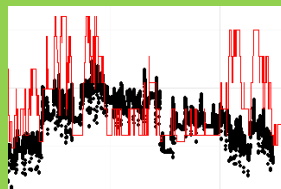
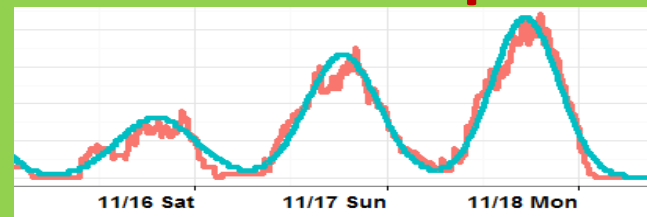
Host1



## Workload prediction

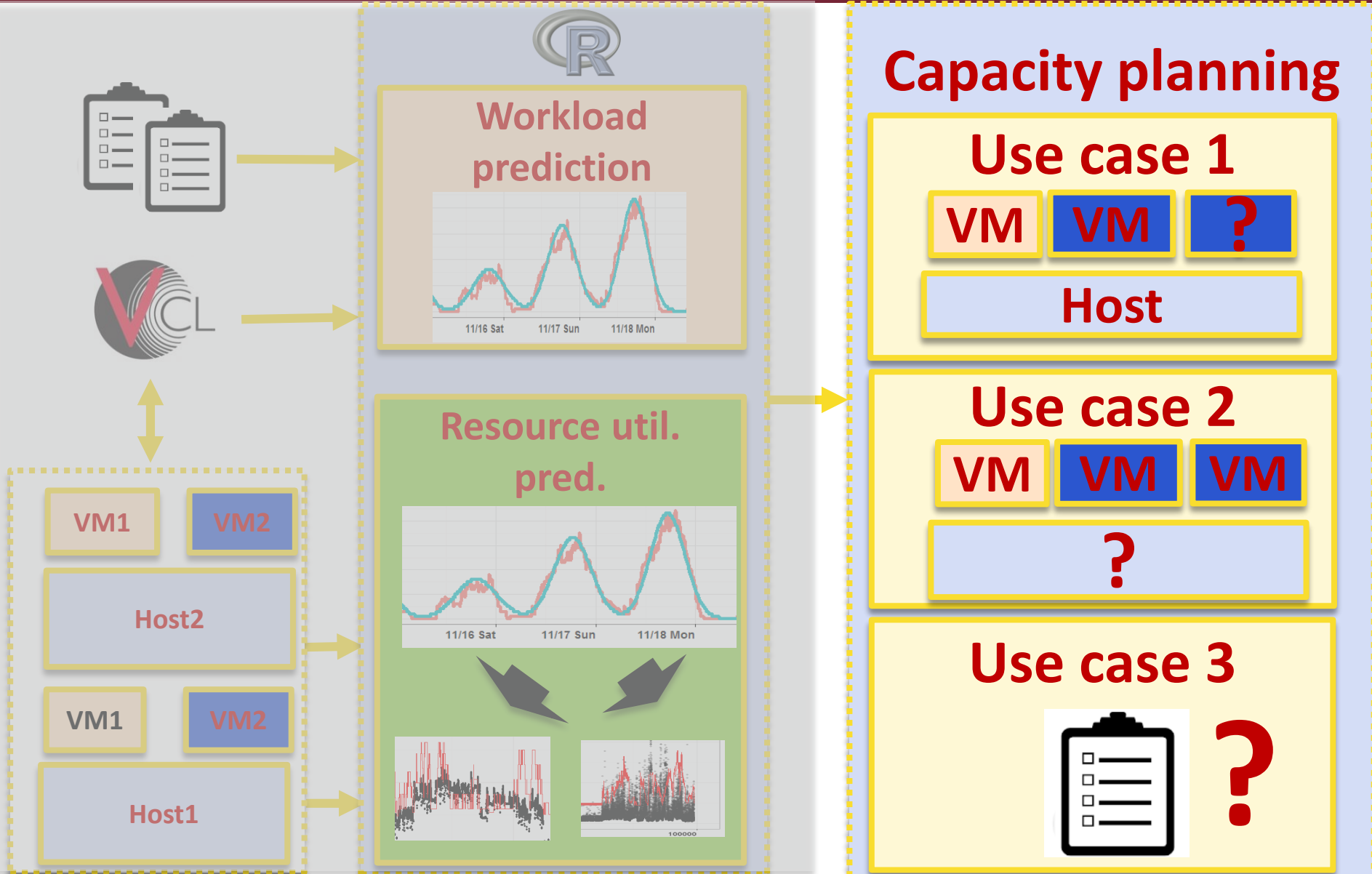


## Resource util. pred.



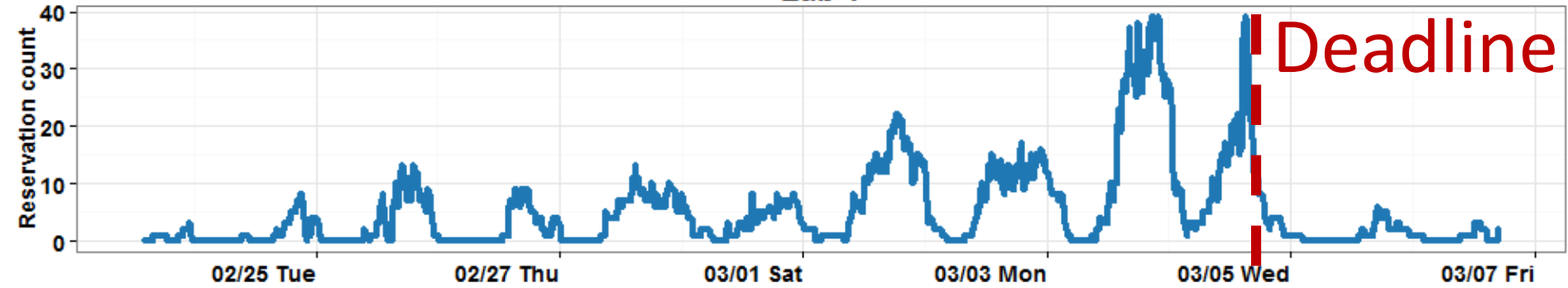


# Workflow

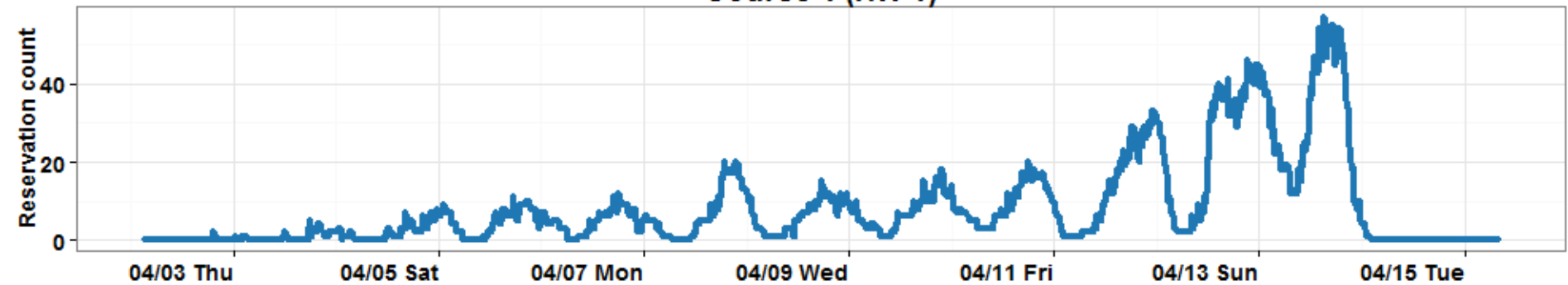


# Workload prediction

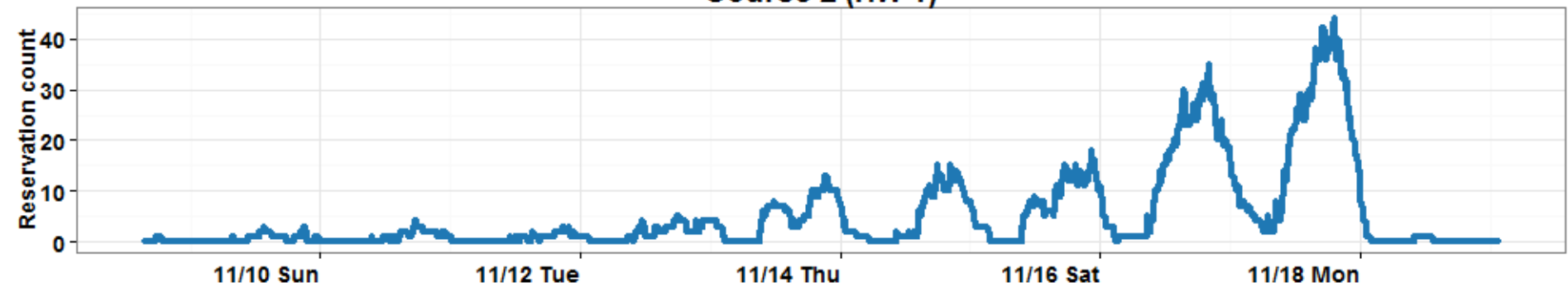
Lab 1



Course 1 (HW 1)

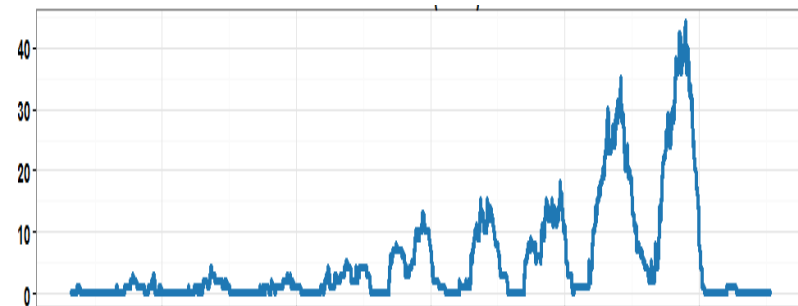
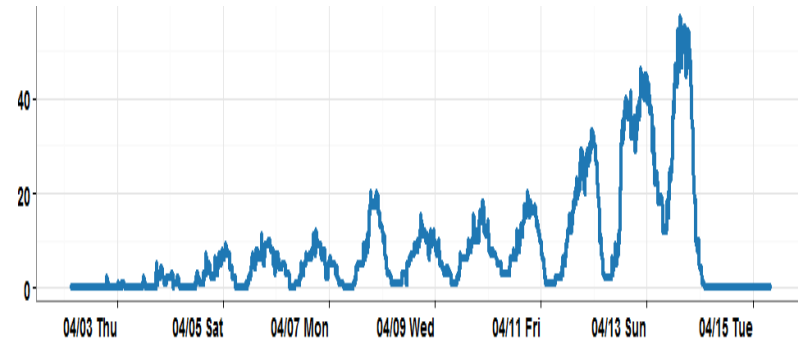
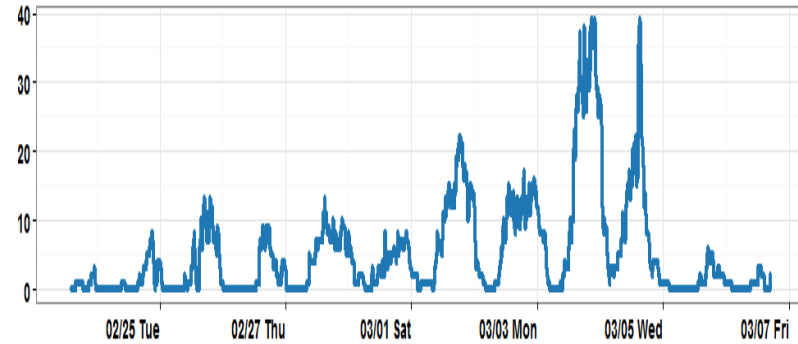


Course 2 (HW 1)



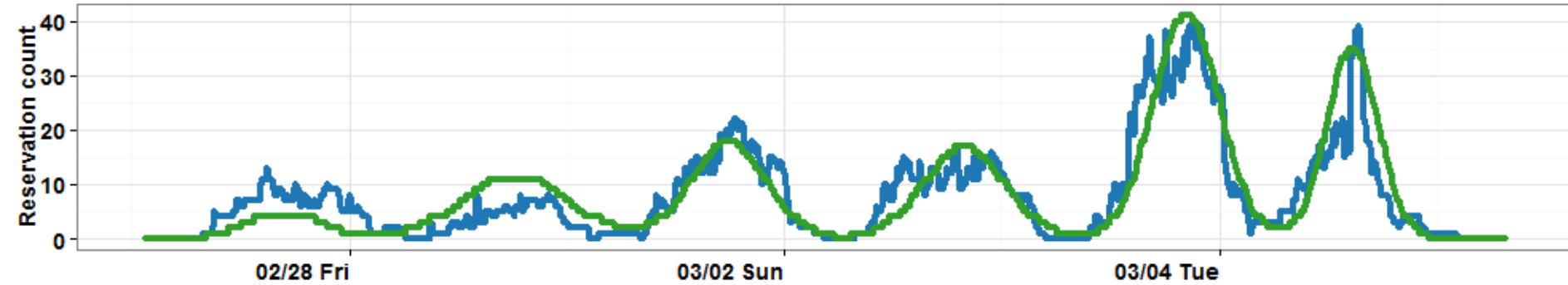
# Workload prediction

- Daily workload follows a Gaussian-like distribution

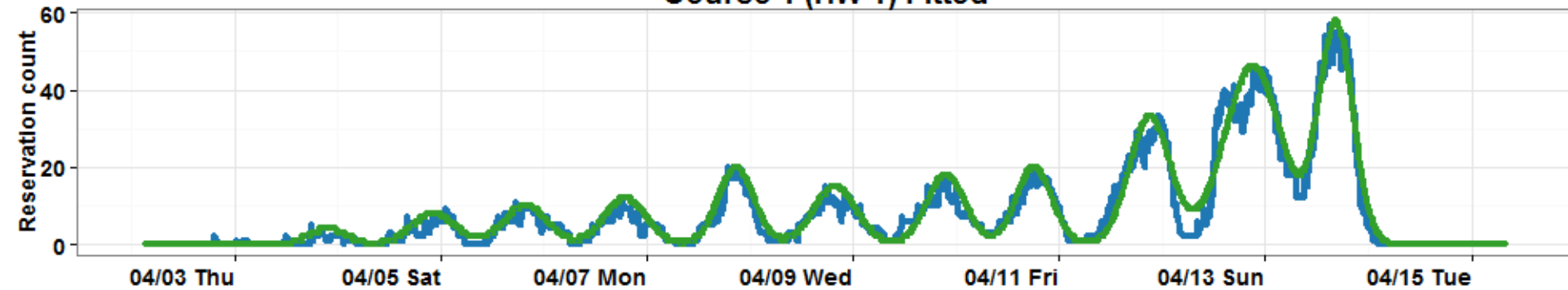


# Model fitting

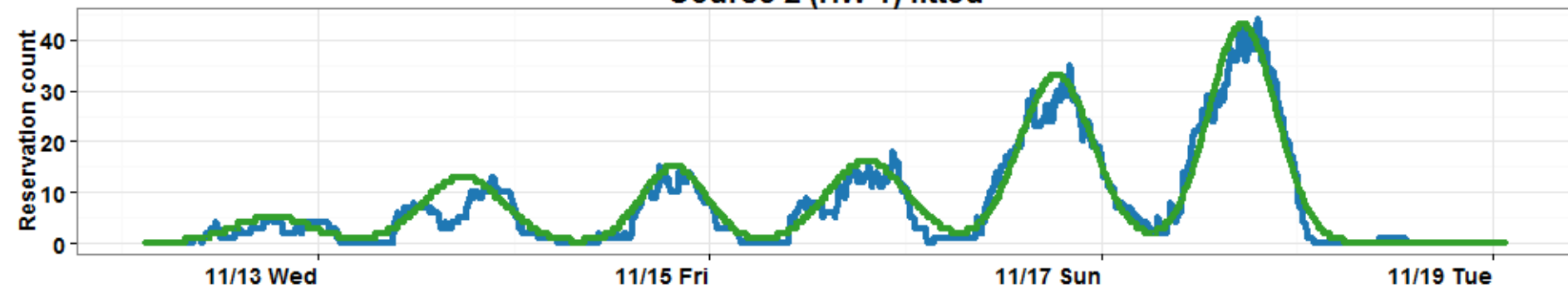
Lab 1 fitted



Course 1 (HW 1) Fitted

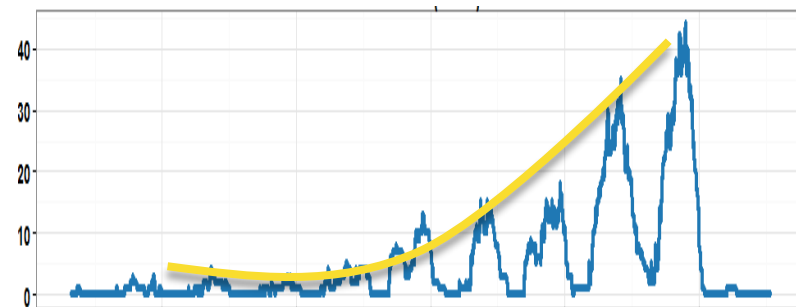
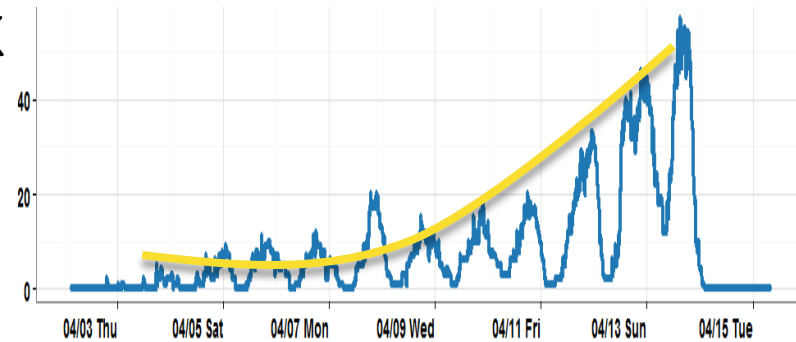
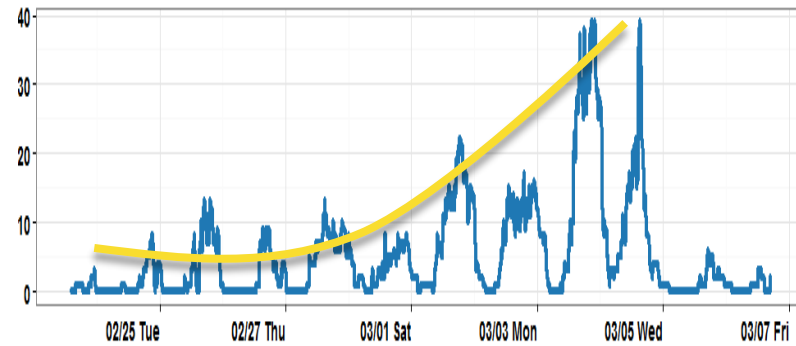


Course 2 (HW 1) fitted



# Workload prediction

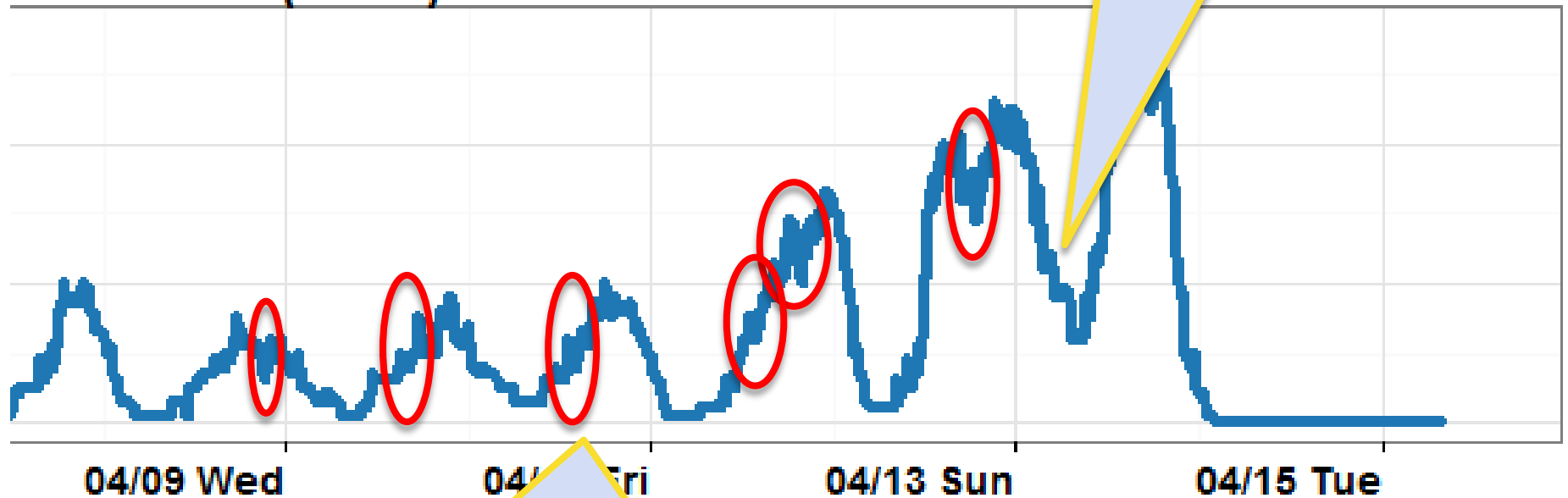
- Daily workload follows a Gaussian-like distribution
- Exponential increase in peak numbers
- maximum location between 7 PM and 11 PM
- ~4 hours as standard deviation



# Workload prediction

## Course 1 (HW 1)

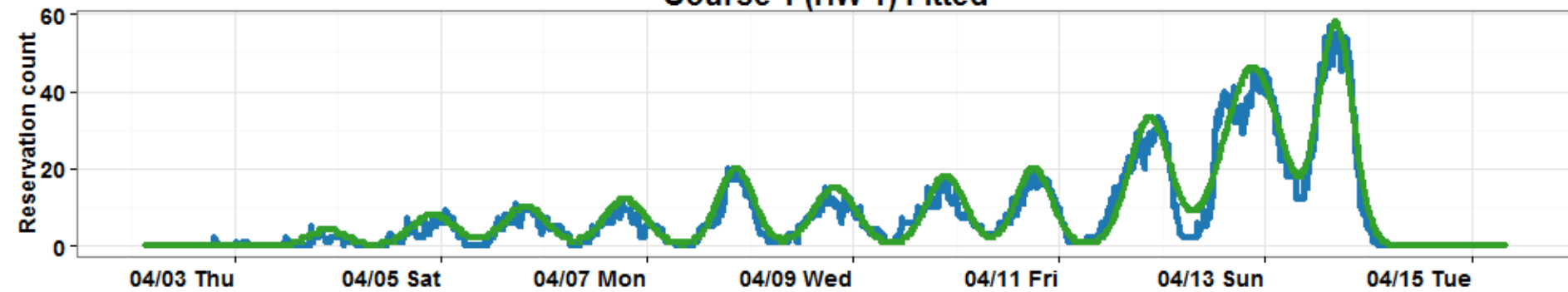
Students work even in the night



They have lunch and dinner

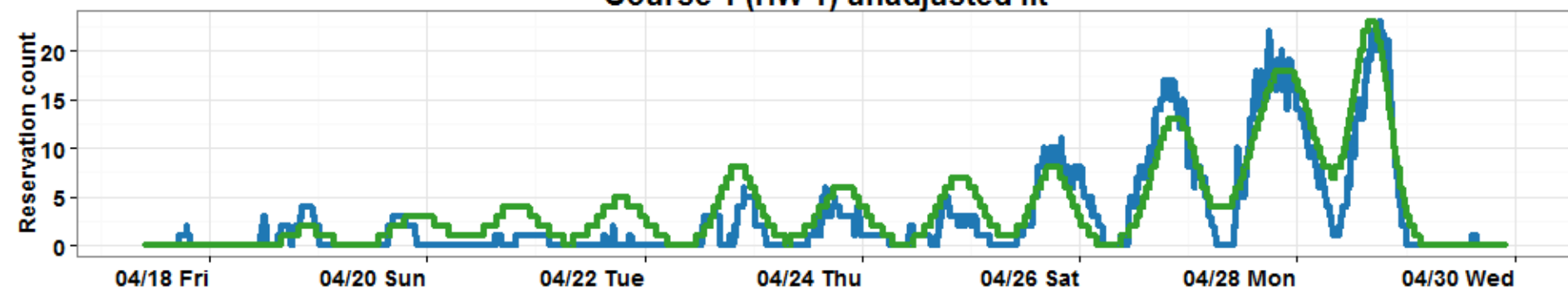
# Workload prediction

Course 1 (HW 1) Fitted



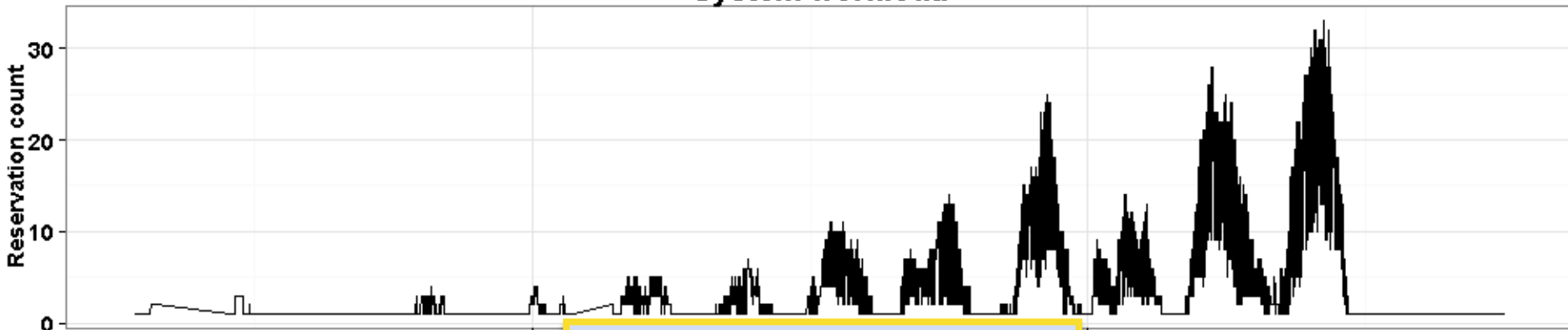
Changes in students' behavior?

Course 1 (HW 1) unadjusted fit



# Resource Utilization Prediction

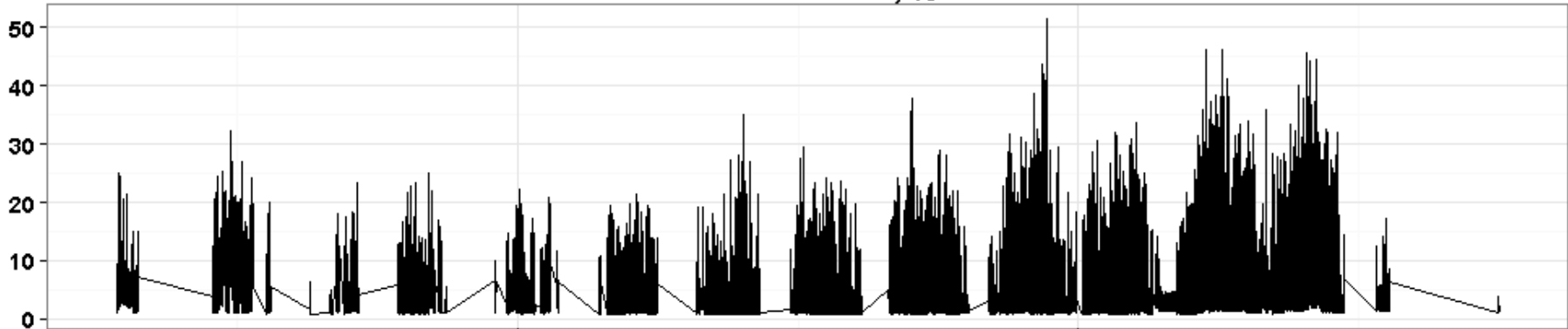
System workload



1384000000

$f(\text{workload})??$

CPU utilization, %



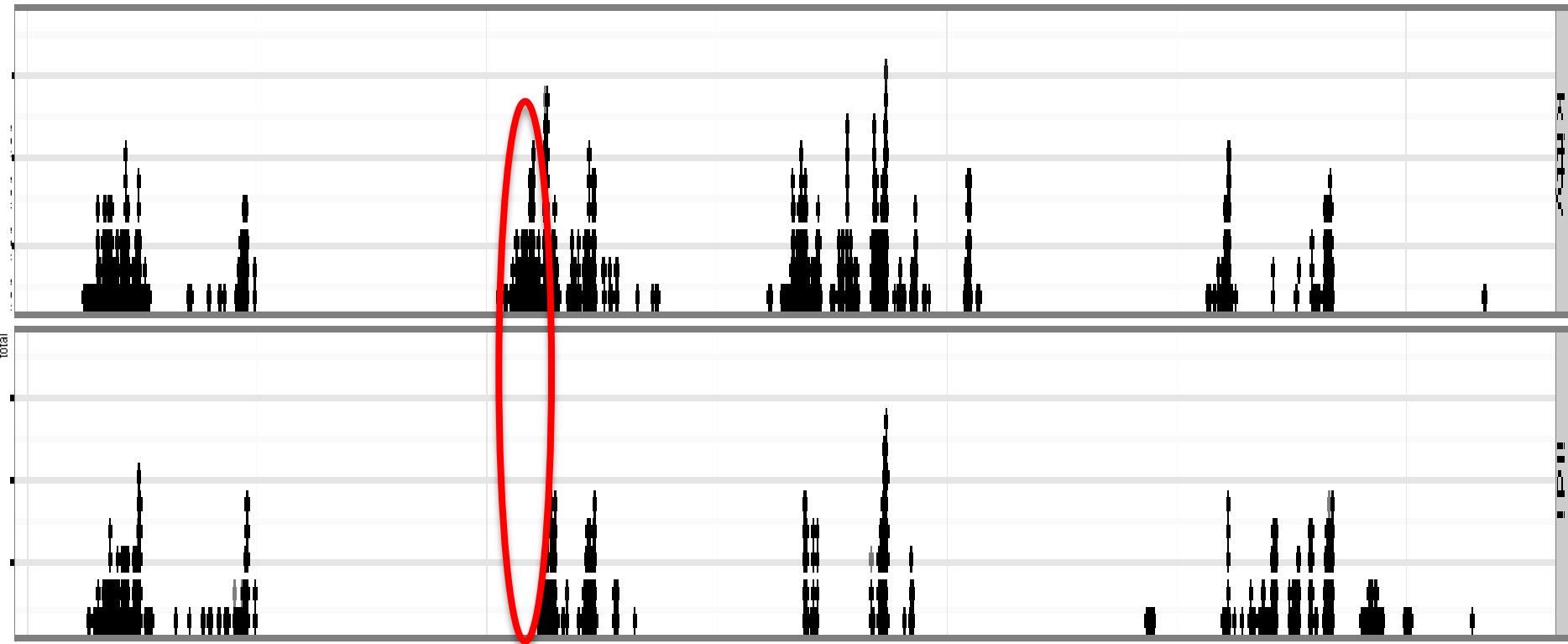
1384000000

1384500000



# Challenges

- It is a cloud
  - Statistical multiplexing 😊



2012/2013/2

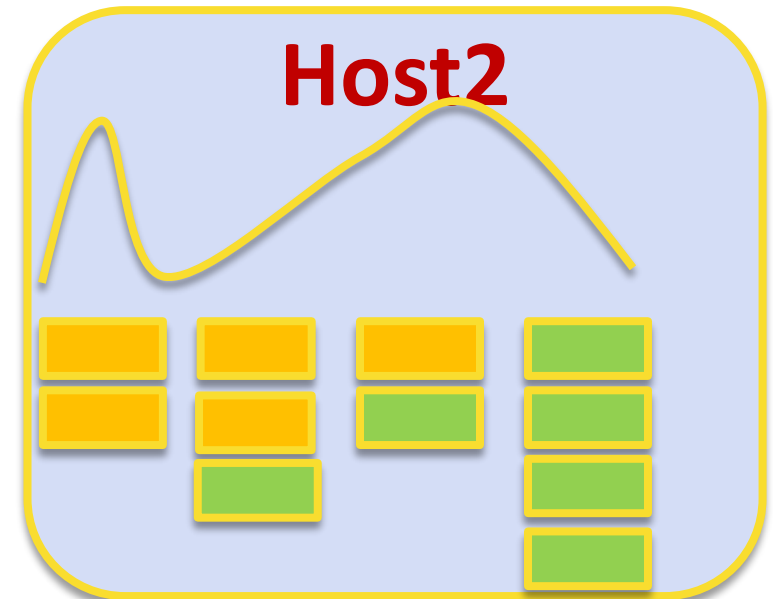
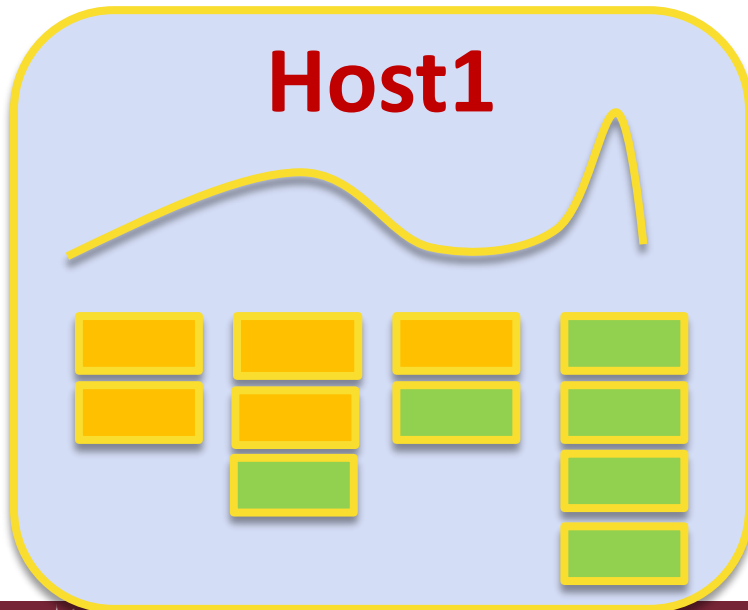
2013/2014/1

2013/2014/2

2014/2015/1

# Challenges

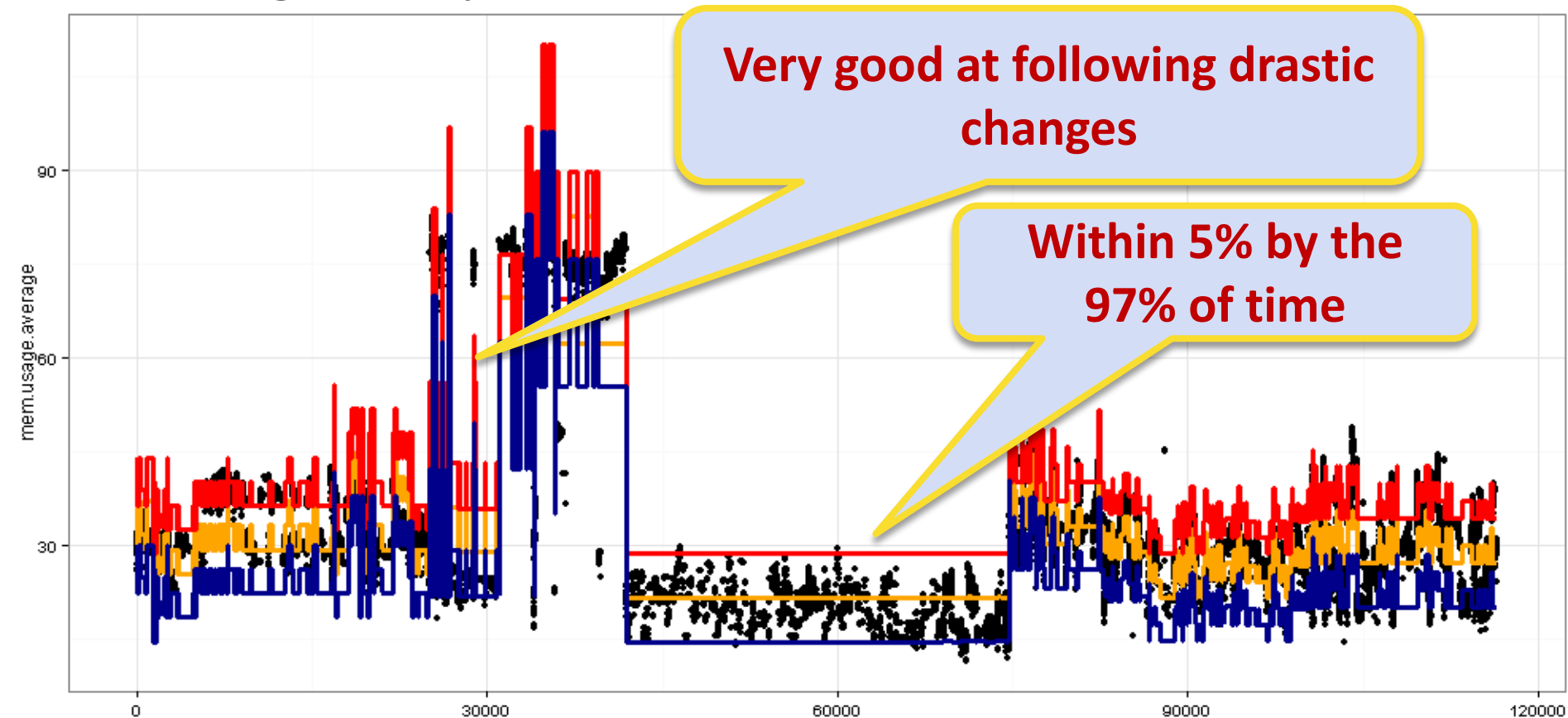
- It is a cloud
- Hosts show different behavior
  - Warm spare
  - Different user behavior
  - ???



# Resource utilization analysis: memory

## ■ Linear model

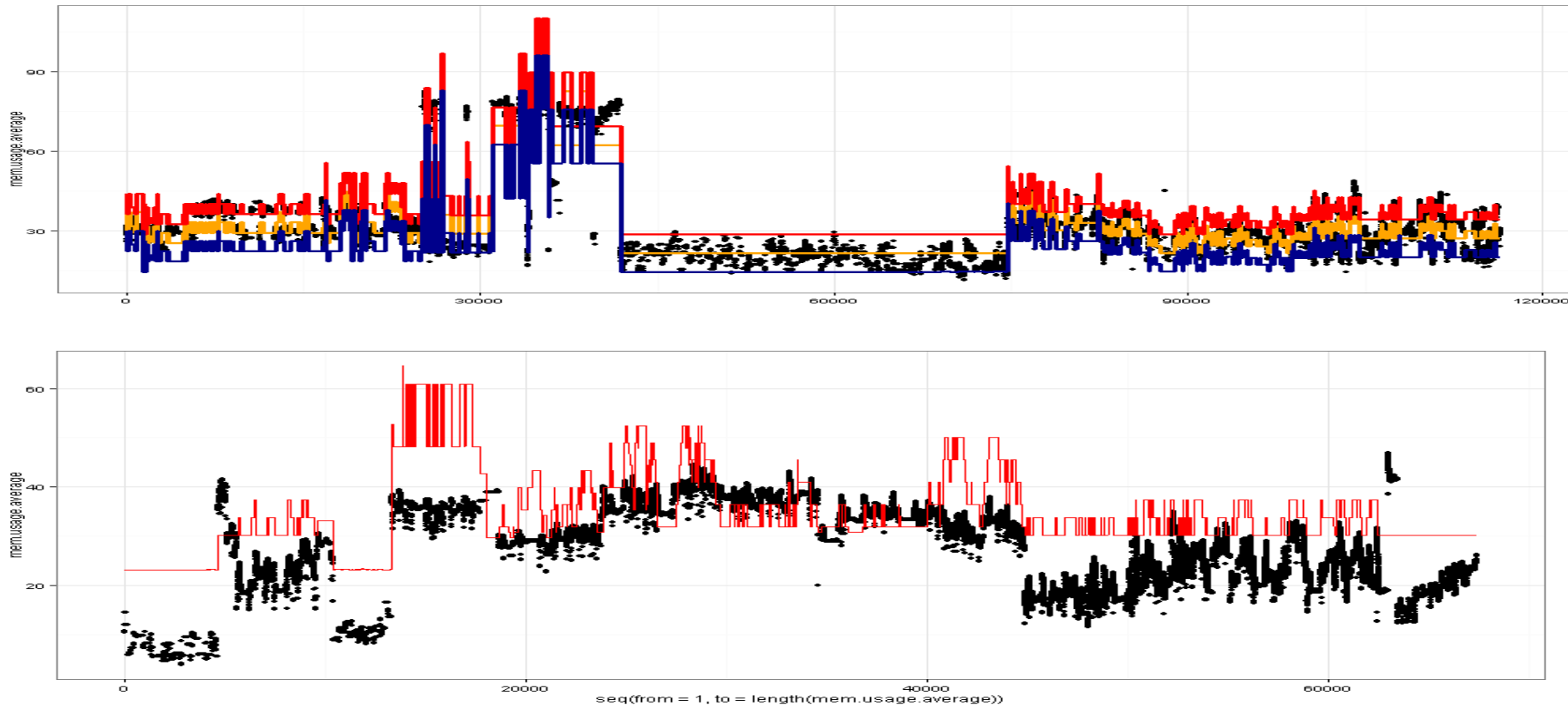
- $Mem(VM_1) + Mem(VM_2) + \dots + Mem(mgmt)$
- Weighted by the workload



# Resource utilization analysis: memory

## ■ Linear model

- $Mem(VM_1) + Mem(VM_2) + \dots + Mem(mgmt)$
- Weighted by the workload



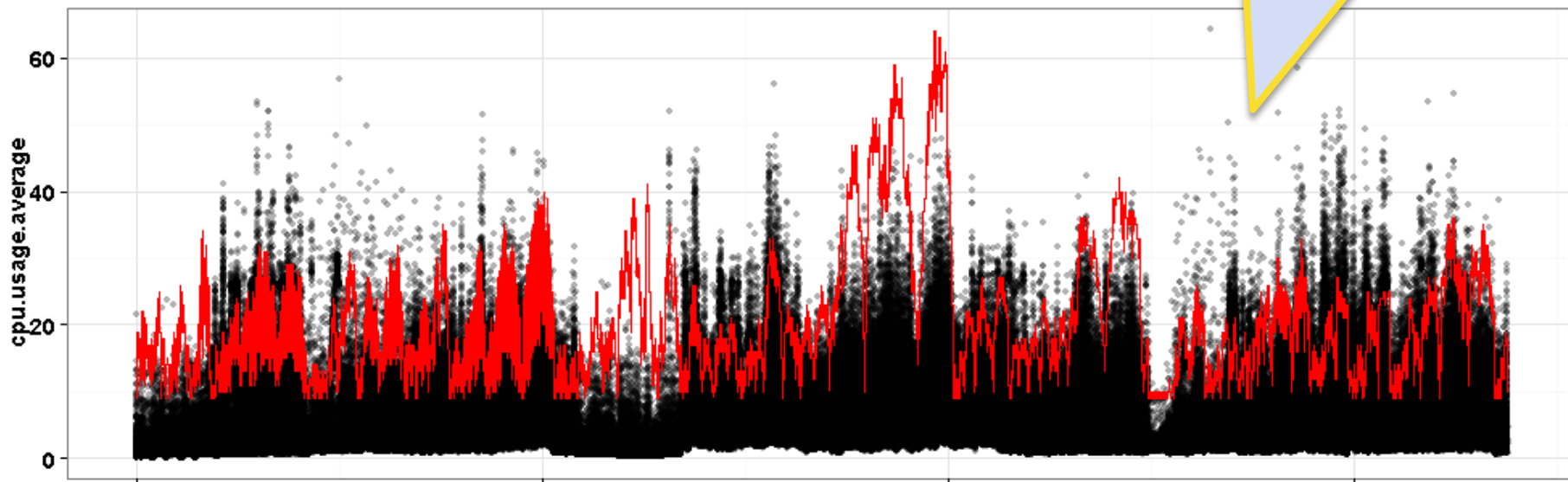
# Resource utilization analysis: CPU

- Linear model

- $CPU(VM_1) + CPU(VM_2) + \dots + CPU(mgmt)$

- Weighted by the workload

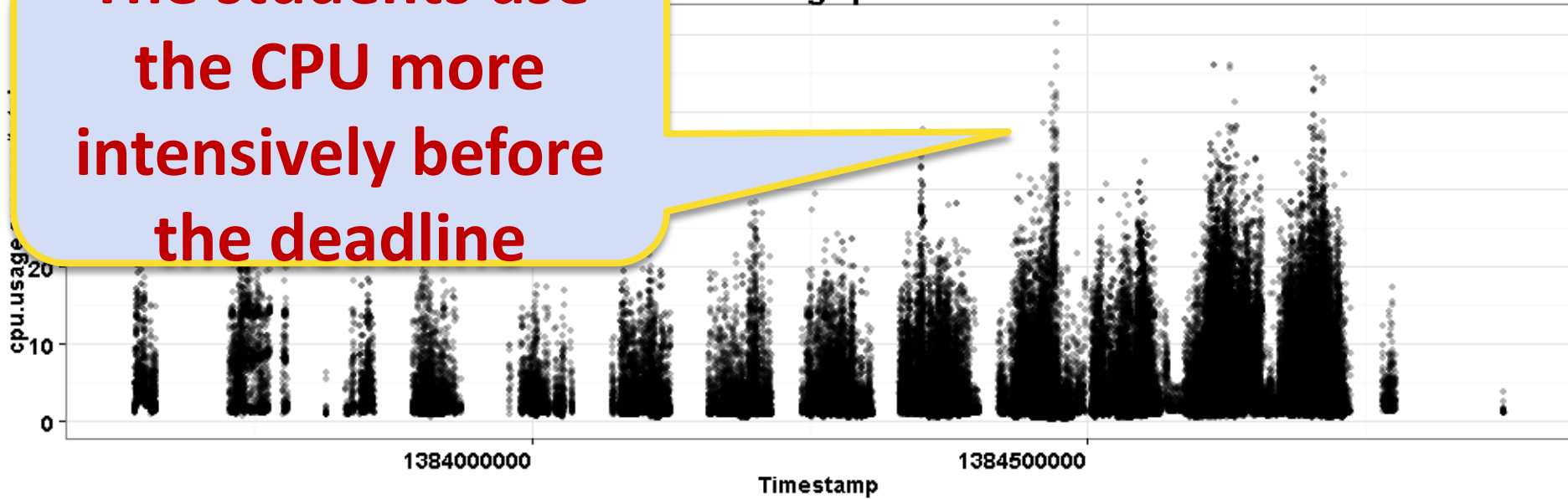
CPU is much more sensitive than memory



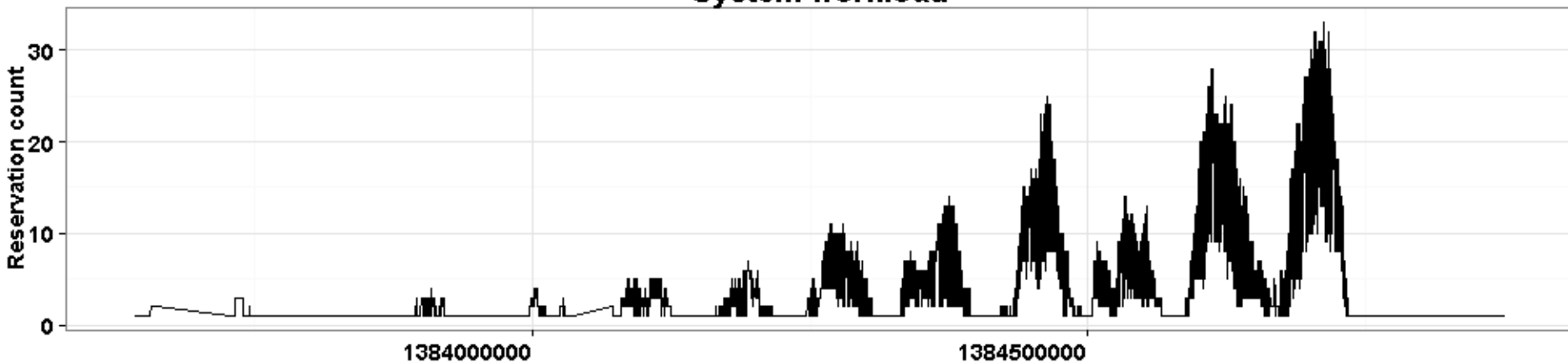
# Resource utilization analysis: CPU

**The students use the CPU more intensively before the deadline**

CPU usage per VM



System workload

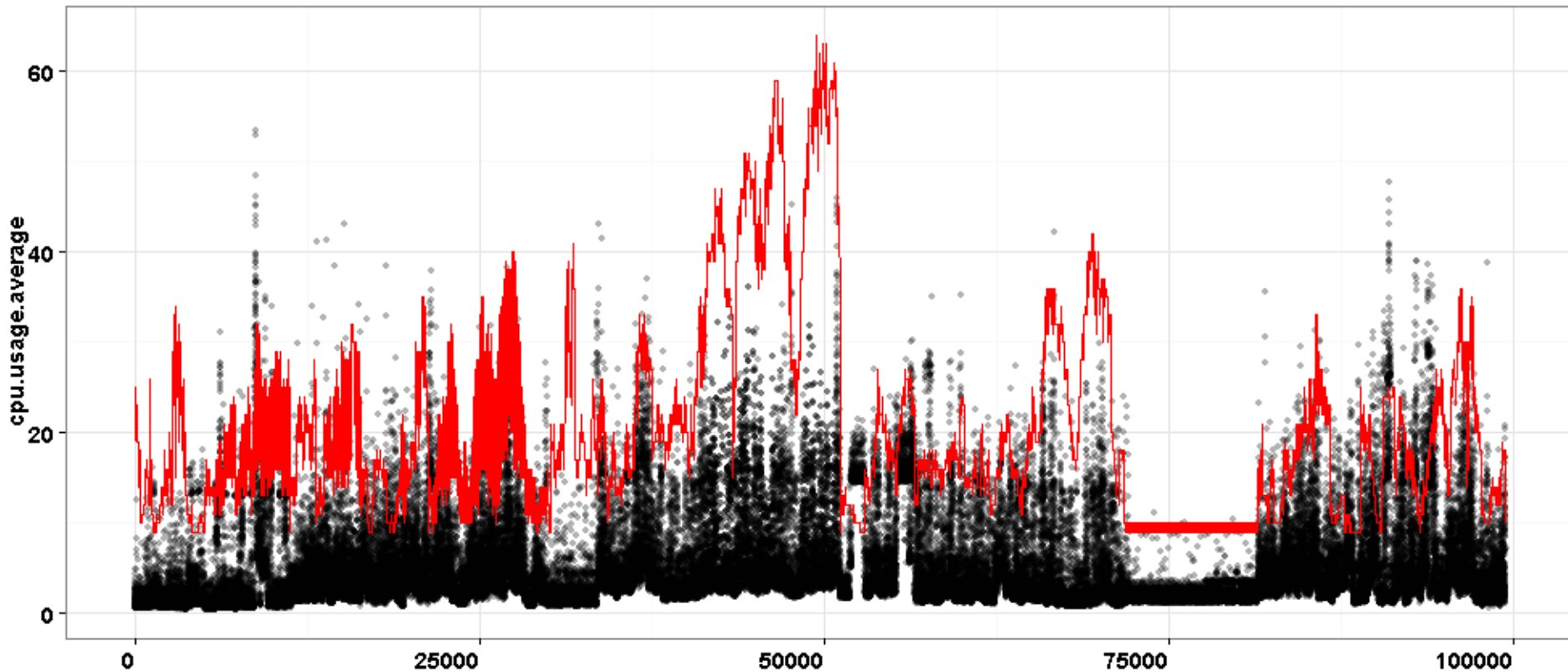


# Resource utilization analysis: CPU

- Linear model

- $CPU(VM_1, wl) + CPU(VM_2, wl) + \dots$

- Weighted by the workload



# Summary

- Data-driven static capacity planning
  - „user behavior” analysis
  - resource fingerprint estimation
- Conclusions:
  - student behavior can be modelled
  - resource allocation were sometimes (too) strict
- Dynamic capacity planning?
  - Long loading time → failed reservations soon
  - When to burst out to a public cloud?