



Prediction-based Optimization in Mobile Networks

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[Developing the
Science of Networks]

Outline

- **Introduction**
- Prediction model
- Optimization framework
- Single user optimization with uncertainties
- Multi-user optimization
- Conclusions



Multimedia in mobile networks

- Main source of mobile traffic [1]
- Expected to continue its growth (>70%) [1]
- Predictable requests and multimedia bitrate [2]
- Adaptable video quality [3]
- Buffering



- [1] "Cisco Visual Networking Index: Global Mobile Data Traffic Forecast Update," 2014.
[2] M. Ahmed, S. Spagna, F. Huici, and S. Niccolini, "A peek into the future: predicting the evolution of popularity in user generated content," in ACM WSDM 2013.
[3] R. Pantos and W. May, "HTTP live streaming," IETF Draft, June, 2010.

Objectives

- Optimizing resource allocation for multimedia streaming will improve both network efficiency and users quality of experience (less streaming interruptions, higher average bitrate)
- Develop a general prediction model [1] accounting for different time scales and reliability
- Single user optimization [2] dealing with uncertainties
- Multi user optimization [3] accounting for variable quality

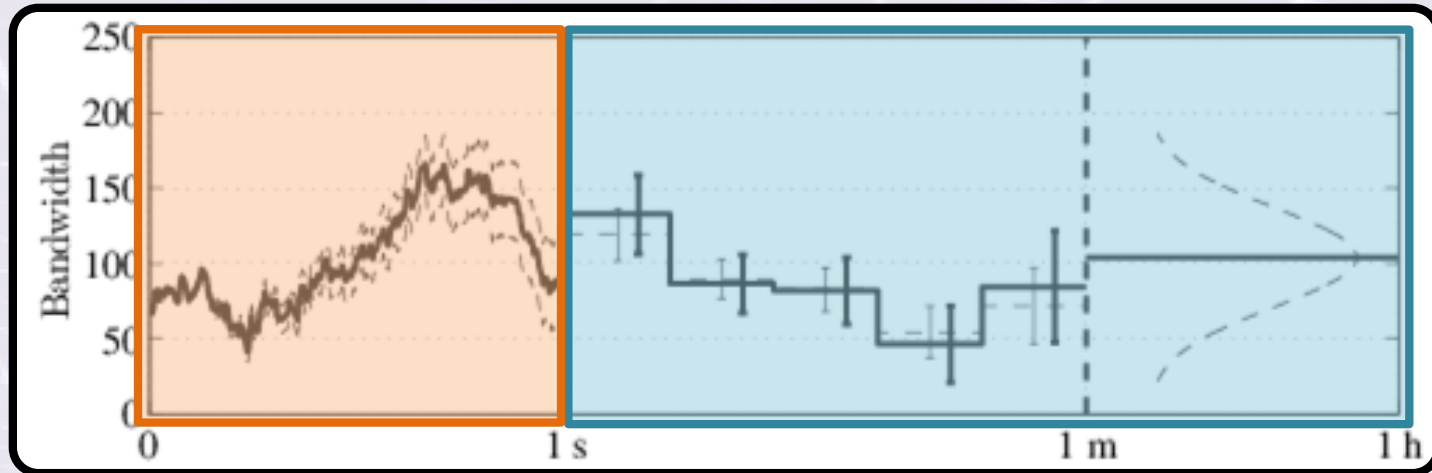
[1] Nicola Bui, Foivos Michelinakis, Joerg Widmer “A Model for Throughput Prediction for Mobile Users”, in European Wireless 2014

[2] Nicola Bui, Joerg Widmer “Mobile Network Resource Optimization under Imperfect Prediction”, in WoWMoM 2015

[3] Nicola Bui, Stefan Valentin, Joerg Widmer “Anticipatory Quality-Resource Allocation for Multi-User Mobile Video Streaming”, in CNTCV workshop 2015

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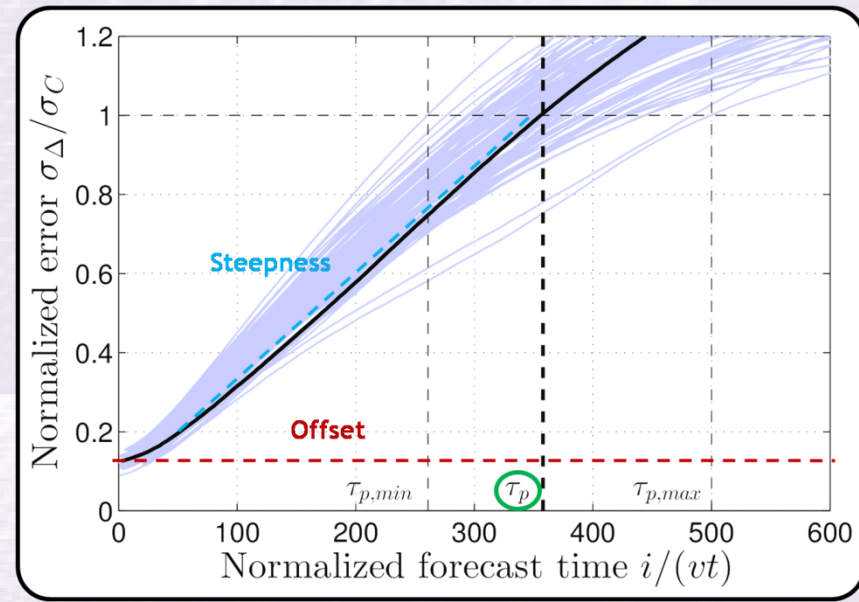


- Idea

- **Short term:** approximate the exact variation of the throughput time series as a sum of random variables using ARMA filters
- **Medium-long term:** approximate the statistic distribution of the throughput accounting for uncertainties (user position, cell congestion, fading, etc.)

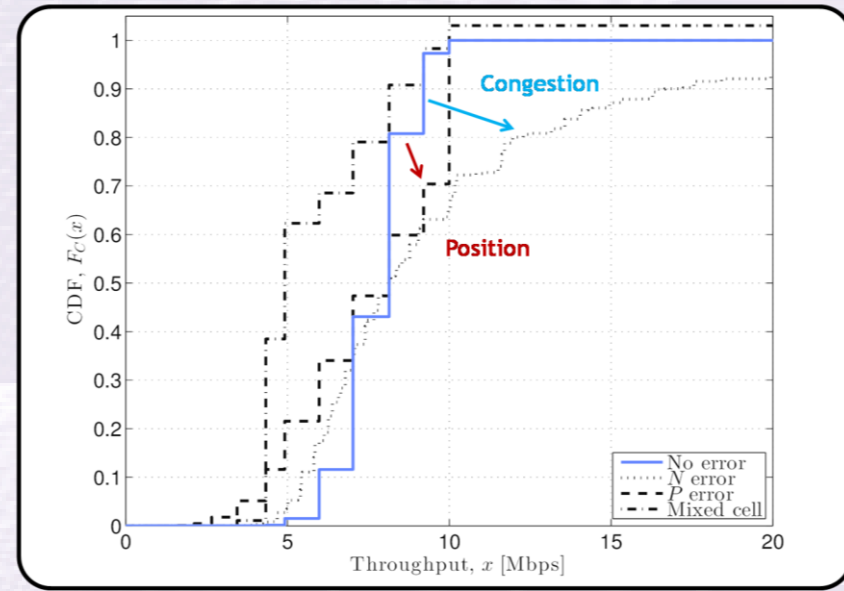
Short-term prediction

- ARMA filters to model achievable rates (throughput) fluctuations varying the sequence dynamics
- After a normalization over dynamics the prediction error shows a characteristic shape
- Two main components:
 - Offset dependent on the intrinsic noise
 - Steepness dependent on the dynamics speed
- Definition of reliability time



Medium-long term prediction

- The objective is to characterize the impact of imperfect knowledge on the achievable rate distribution
- **Congestion** -> the number of active users in a cell is not easily obtained
- **Position** -> in a given time frame the user position (both actual and measured) may vary





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Diffraction grating
 $d \sin \theta = m \lambda$ ($m=0, \pm 1, 2$) maxima lines
 slit separation
 $R = Nm$ resolving power
 $\Delta \theta_{hw} > \frac{\lambda}{Nd \cos \theta}$ half width
 aperture diameter

Plane mirrors
 $i = -p$
 Obj - S is pos if on same side as incoming light
 Img - S' is pos if on same side of outgoing light
 F - pos if on same side of outgoing light

spherical mirror
 $\frac{1}{f} = \frac{1}{p} + \frac{1}{q}$
 $m = -\frac{q}{p}$
 real, virtual, inverted, upright, magnified, diminished

$m = \frac{25cm}{f}$

System model: inputs

- Users' achievable rates are predicted:

$$R = \{r_{i,j} \in [0, r_M], i \in \mathcal{N}, j \in \mathcal{T}\}$$

- Can be assigned to users

$$A = \{a_{i,j} \in [0, 1], i \in \mathcal{N}, j \in \mathcal{T}\}$$

- The can buffer data

$$b_{i,j+1} = b_{i,j} + a_{i,j}r_{i,j} - \underline{d_{i,j}} - \underline{u_{i,j}}$$

- To stream the content at and bitrate between

Minimum and **Maximum**

Optimization problem

- Optimization priorities
 1. Minimize total interruption time
 2. Minimize total missing extra quality
- $K \gg 1$ enforce priorities

$$\text{minimize}_A \quad \sum_{k \in \mathcal{N}} \sum_{m \in \mathcal{T}} \left(\underline{Kl_{k,m}/d_{k,m}} + \underline{e_{k,m}/u_{k,m}} \right)$$

$$\text{subject to:} \quad a_{i,j} \geq 0; \quad \sum_{k \in \mathcal{N}} a_{k,j} \leq 1$$

$$\forall i \in \mathcal{N}; j \in \mathcal{T}$$

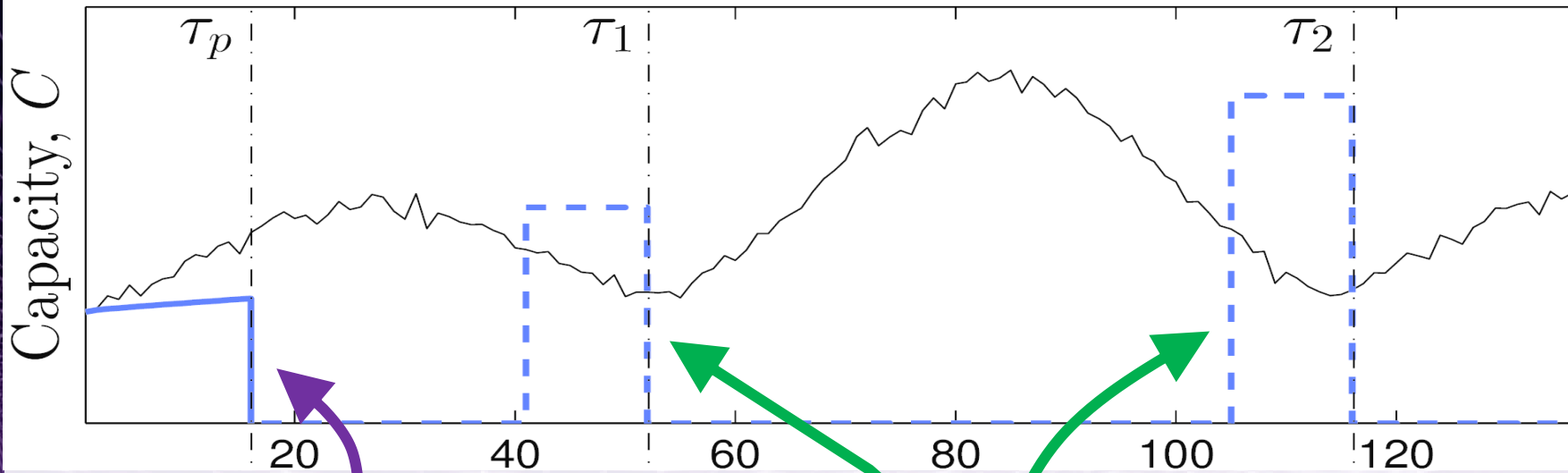


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Optimization with imperfect prediction

- Using predictors and statistic to create a predicted achievable rate sequence
- Solve the optimization problem
- Adopt the solution for a few time slot
- Update the prediction and repeat

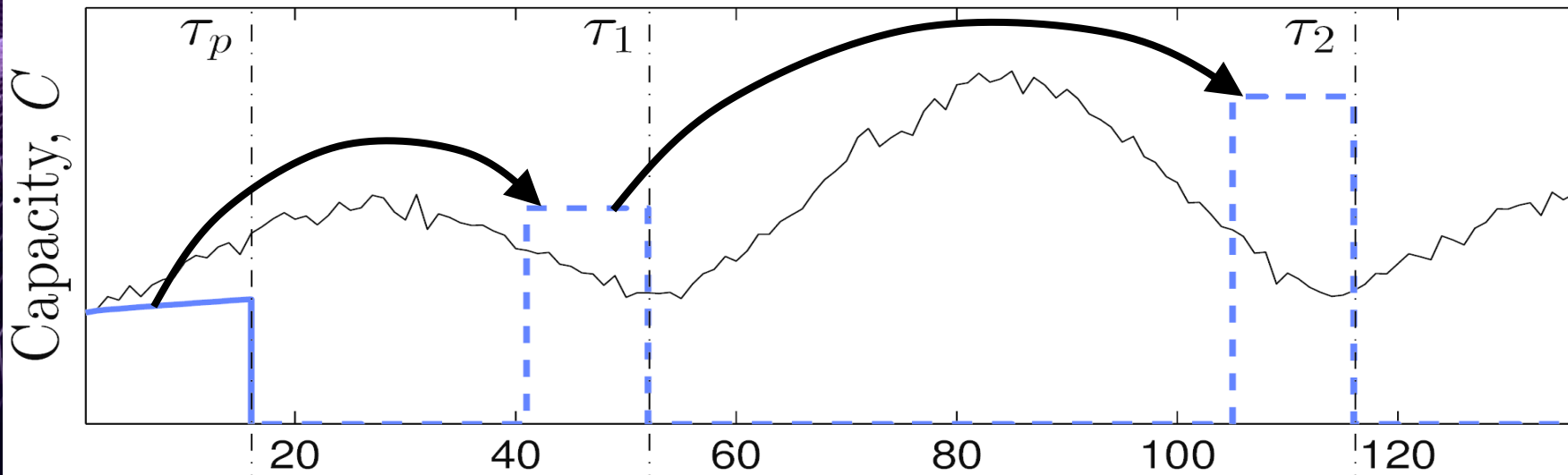
Combined prediction example



- Short term prediction
- Worst case scenario from the medium-long term models

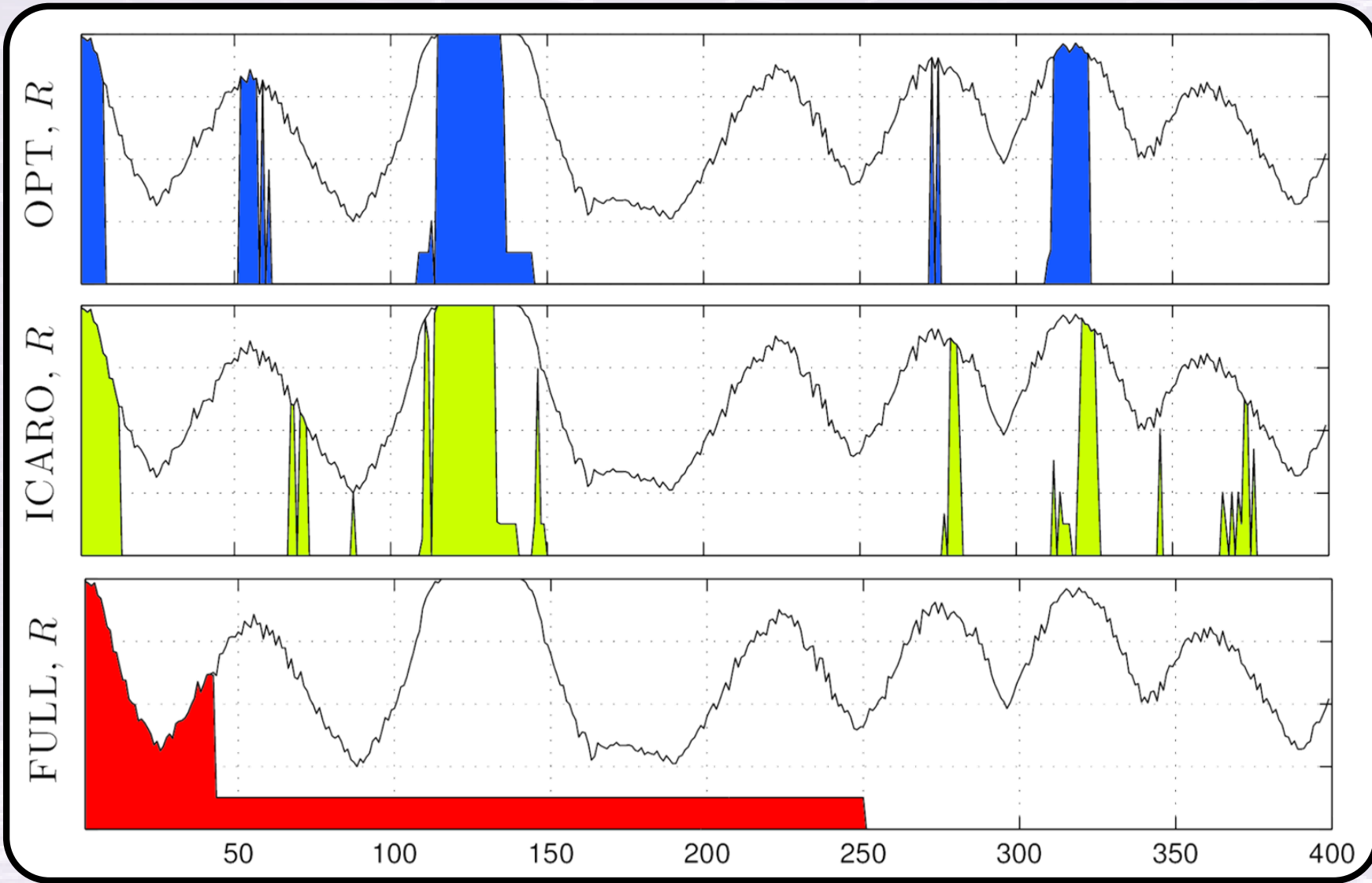
ICARO: concept

- Imperfect Capacity prediction-Aware Resource Optimization
- Idea: to **iteratively decide** how much of the **current rate** to use according to the **Split and Sort** algorithm applied to the **combined prediction sequence**
- The current slot is not taken if the buffer allows to wait for cheaper slots



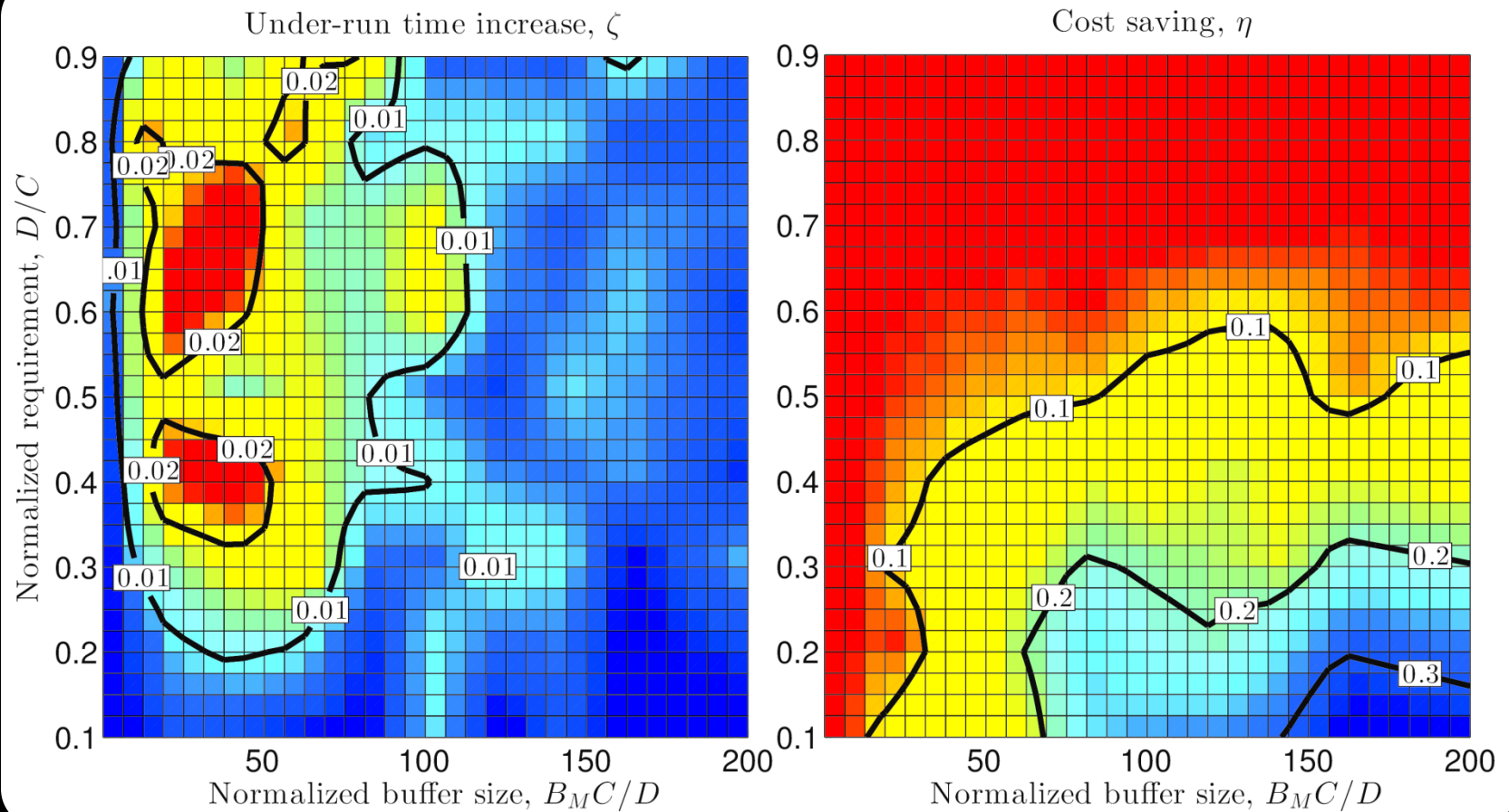


ICARO: result example



ICARO: overall results

- Almost optimal interruption time
- Up to 30% resource savings



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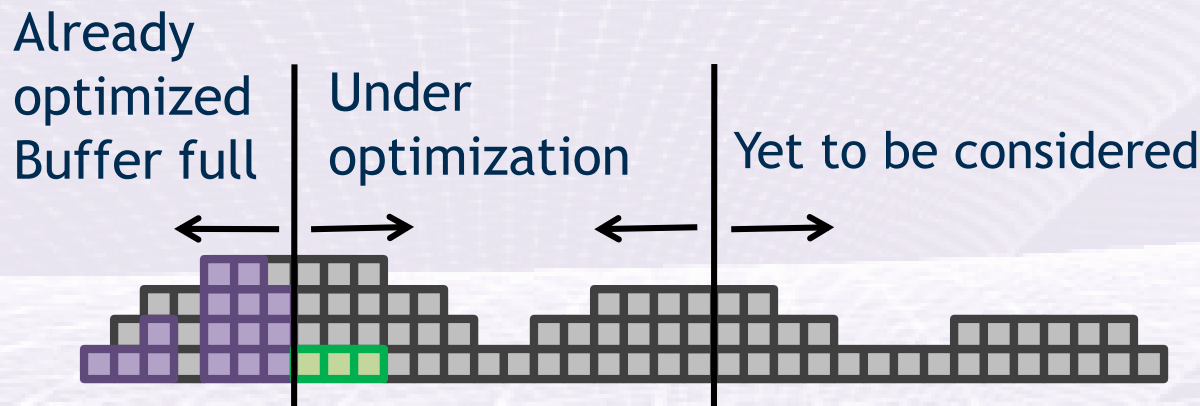
Multiple users optimization

- Heuristic algorithm to solve the continuous streaming and quality maximization problem
- Problem extension to support guaranteed quality of service
- Problem approximation to allow for a fast computation of the solution



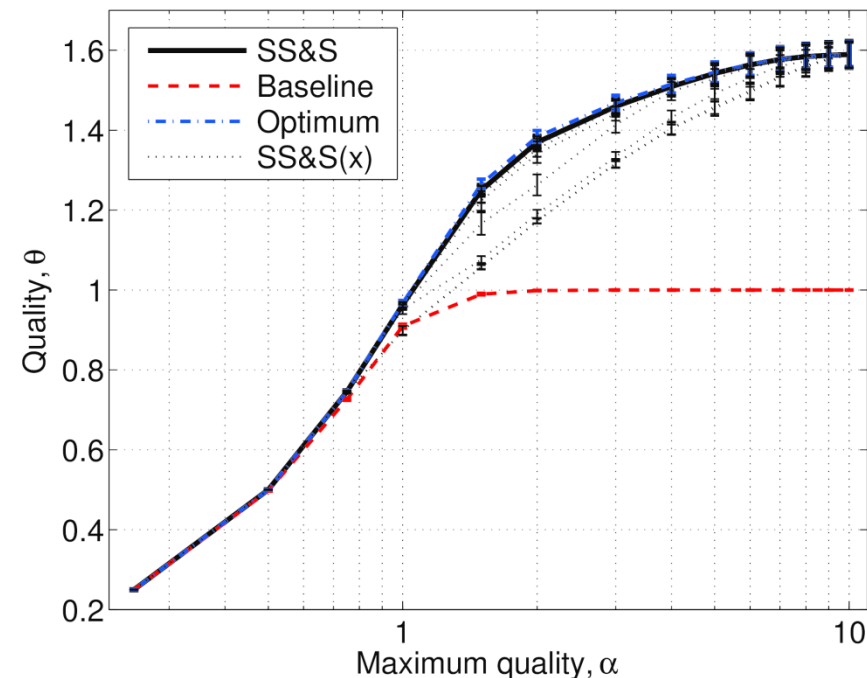
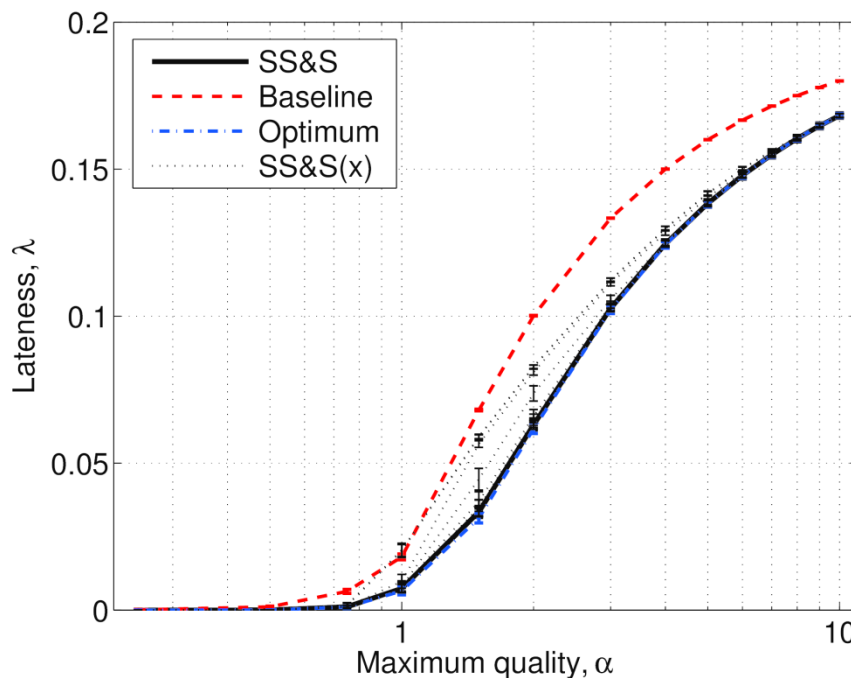
Algorithm

- The Split, Sort & Swap algorithm iterates the following steps:
 - Split & Sort:** greedily optimize a region until no more resource can be allocated due to either complete assignment of the best slot and users or buffer violation.
 - Swap:** within each of the split windows resources are swapped to improve the objective function.



Results: multi user improvements

- SS&S is achieving quasi optimal performance both in terms of saved interruption time (lateness on the left) and maximum quality (right)
- A higher number of iterations is required when $1 < \alpha < 3$: in this region the greedy approach is less likely to obtain an optimal allocation



- MILP formulation

$$\text{maximize}_{A,B,L,E,S} \quad \sum_{k \in \mathcal{N}} (K(\lambda_k + \underline{s_k}) + \theta_k)$$

$$\text{subject to:} \quad a_{i,j} \geq 0; \quad \sum_{k \in \mathcal{N}} a_{k,j} \leq 1$$

$$\lambda_i \geq \underline{\lambda_i^* s_i}; \quad \theta_i \geq \underline{\theta_i^* s_i}$$

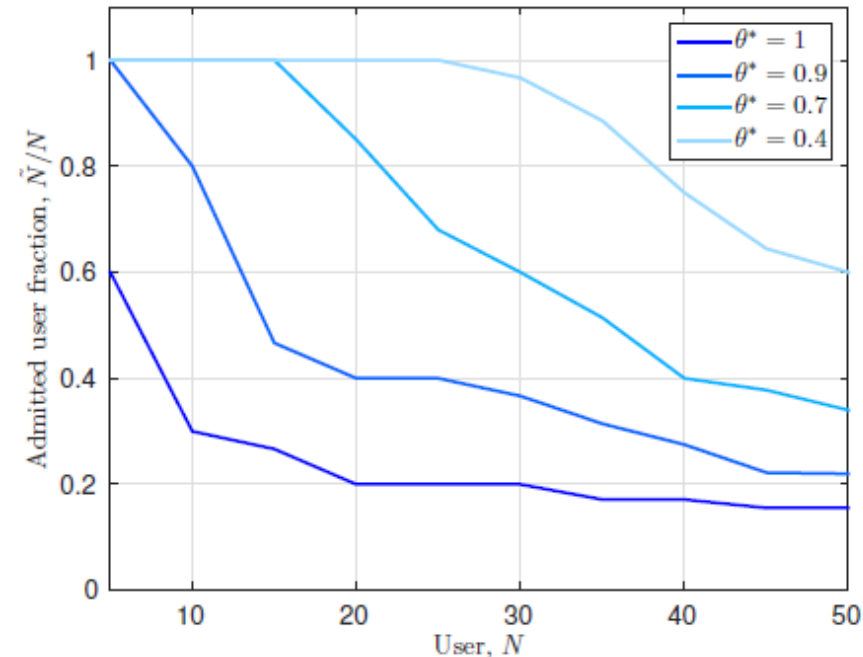
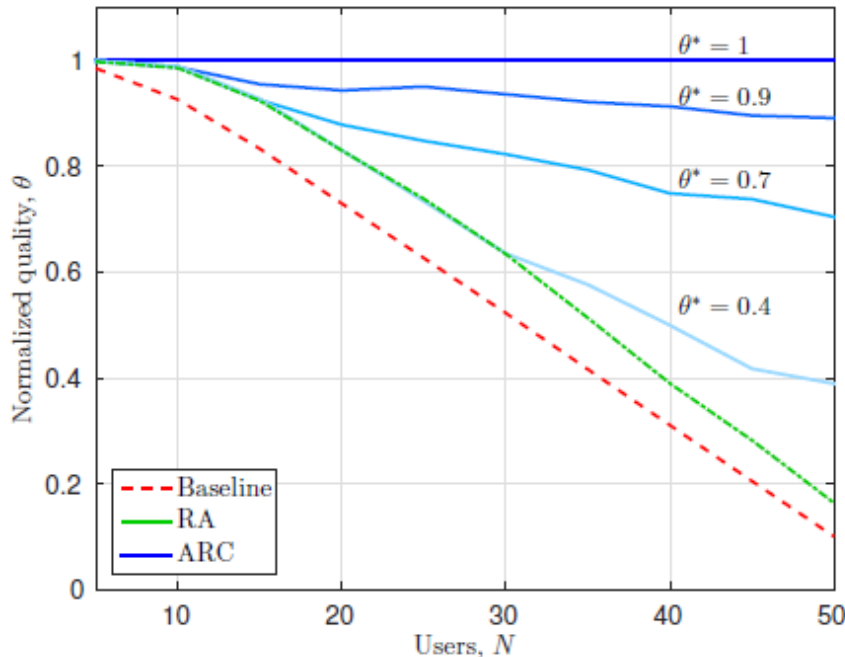
- New integer binary variable for **Scheduled users**
- **QoS constraints** for scheduled users

Approximation

- We rank users according to the problem objective function
- We compute the largest set of admitted users
 - Running a dichotomic search over the set size
 - For each size checking whether the problem is feasible with our SS&S algorithm or an LP solver
 - The last iteration gives also the solution

Results: guaranteed QoS

- The approximated problem is effective in guaranteeing QoS
- Quality (left) is consequently traded with the number of admitted users (right)



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- Prediction model
 - Short term based on filters
 - Medium/long term statistic model
- Single user optimization
 - Accounting for uncertainties
 - Quasi optimal interruption time
 - 30% resource saving
- Multi-user optimization
 - Accounting for variable quality
 - Guaranteed QoS

Thanks

Prediction-based Optimization
in Mobile Networks

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Stefan Valentin, Ilaria Malanchini

Questions?