



# Low-Complexity No-Reference PSNR Estimation for H.264/AVC Encoded Video

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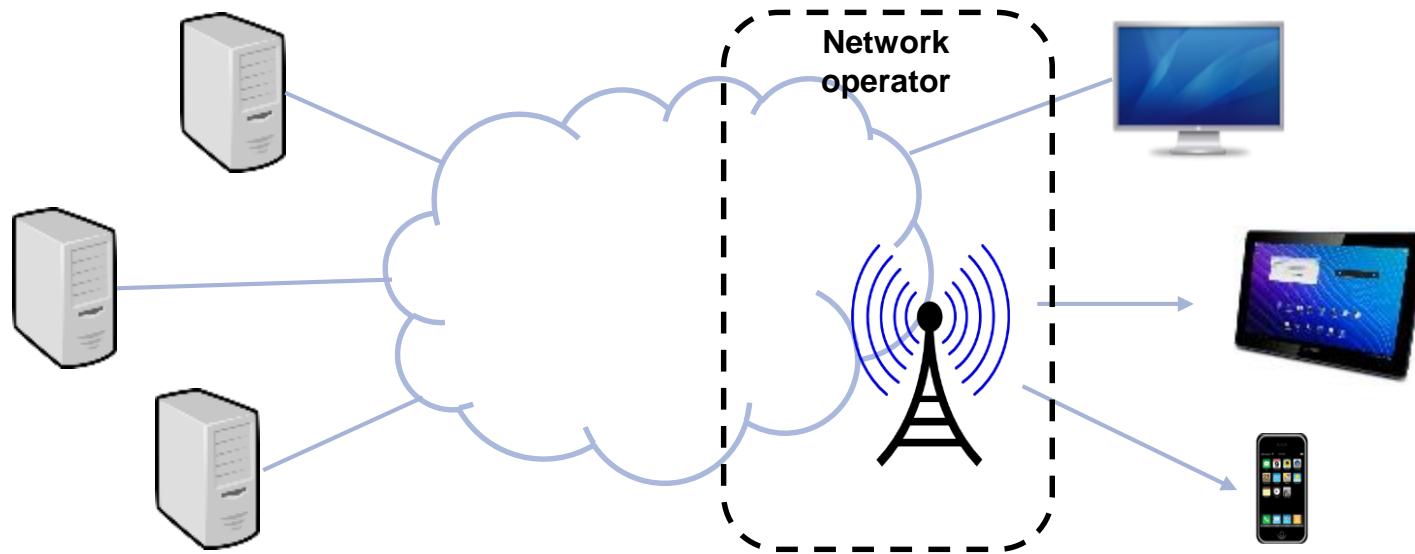
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# Agenda

- Motivation
- Related work
- Proposed No-Reference PSNR estimation
- Validation
- Example application
- Conclusion

## Motivation

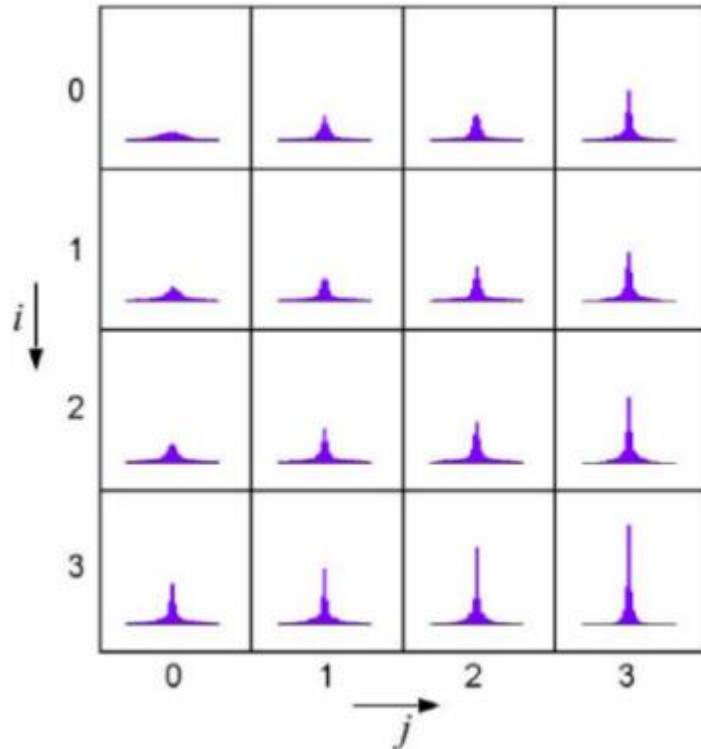


- Fast increasing video traffic to a huge number of clients
  - Need for scalable in-network quality monitoring
- Mostly HTTP/TCP based video streaming
  - No network-induced video distortion

➤ Low-complexity No-Reference PSNR estimation

## Related work (1)

### Statistical properties of transform coefficients

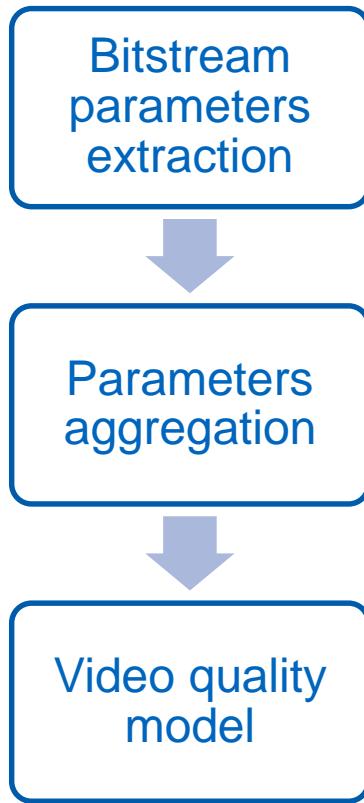


- Distribution  $f_X(x)$  estimated using the quantized coefficients  $X_k$ .
  - Local mean square error is estimated using:  $\varepsilon_k^2 = \int_{-\infty}^{+\infty} f_X(x|X_k)(X_k - x)^2 dx$
  - Estimated MSE:  $MSE_{est} = \frac{1}{N} \sum_{k=1}^N \varepsilon_k^2$
- High complexity

[1] T. Brandao, M. Queluz, „No-Reference Quality Assessment of H.264/AVC Encoded Video, IEEE Trans. CSVT, 2010

## Related work (2)

### Bitstream features based video quality



- 21 basic parameters (motion vectors, DCT coefficients, ...) are extracted from the bitstream
  - Aggregated to 44 picture level parameters
  - Different modules form a video quality model
- 
- High complexity
  - Targeted at RTP/UDP streaming (takes into account packet losses)

[2] ITU-T, Rec. P1202.1 „Parametric non-intrusive bitstream assessment of video media streaming quality“, 2012

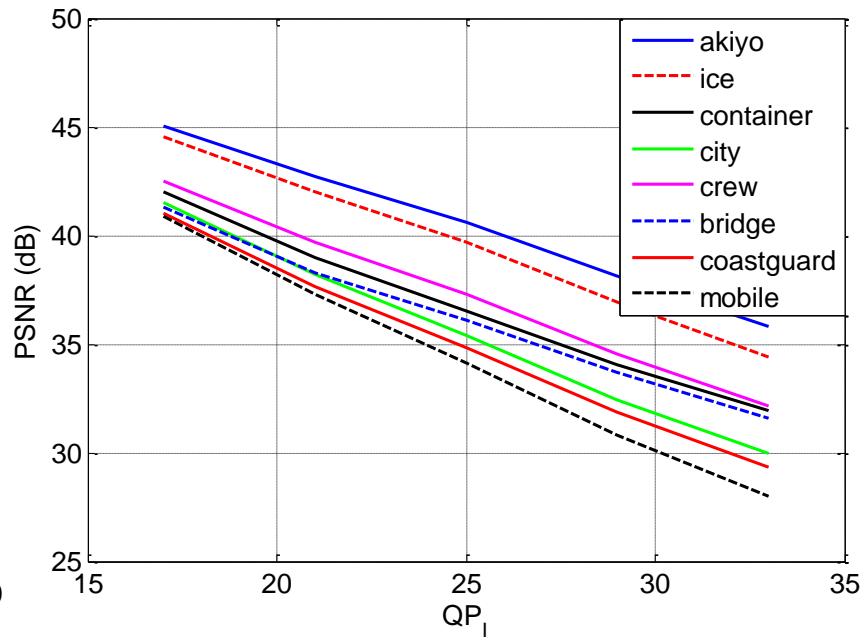
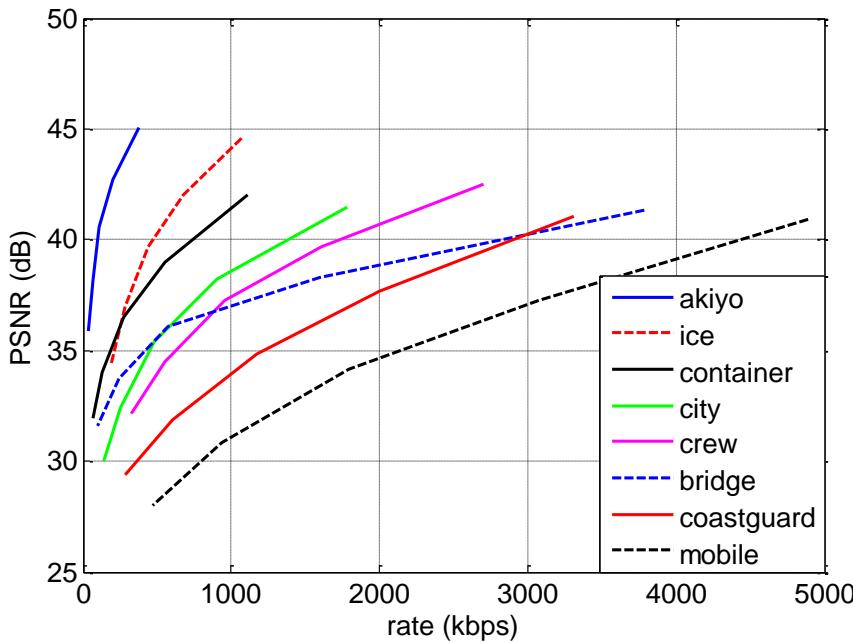


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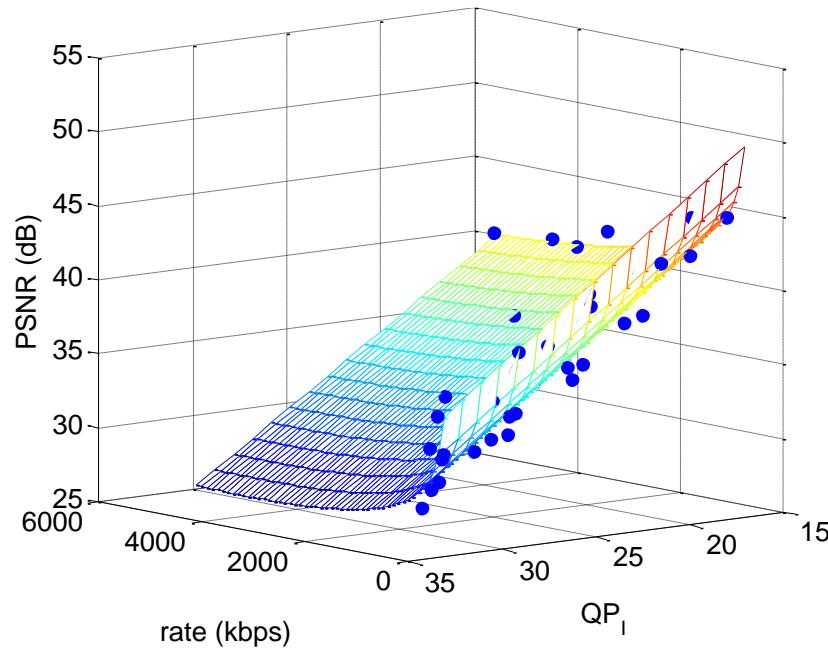
# Proposed No-Reference PSNR estimation

- H.264 baseline, fixed QP
- $QP_I$  denotes mean QP over I-frames



$$\text{PSNR}_{\text{EST}} = b_1 + b_2 * \log(\text{rate}) + b_3 * QP_I + b_4 * \text{rate} * QP_I$$

# Training

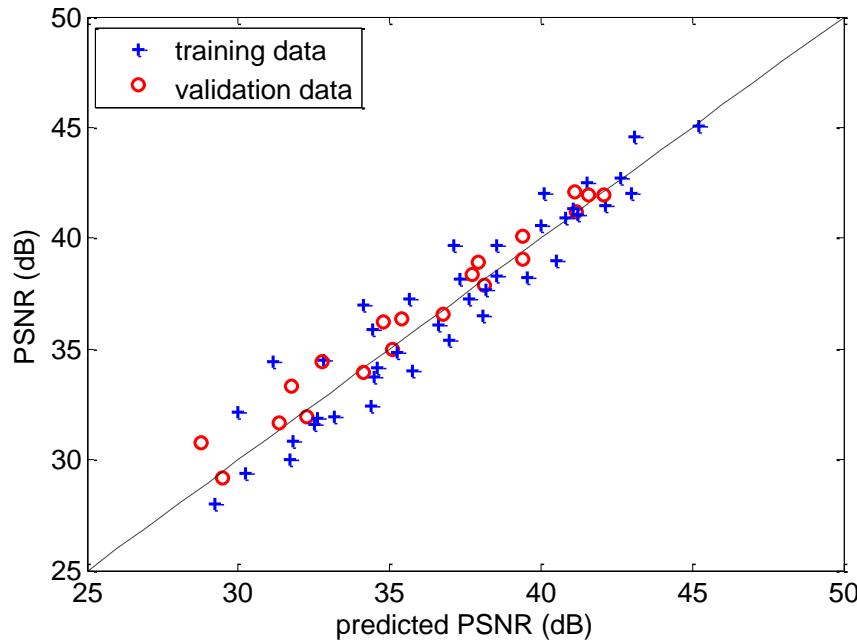


- 8 CIF videos, 30 fps
- Baseline profile
- x264 with constant QP [17, 21, 25, 29, 33]
- Multiple linear regression

$$\text{PSNR}_{\text{EST}} = b_1 + b_2 * \log(\text{rate}) + b_3 * \text{QP}_I + b_4 * \text{rate} * \text{QP}_I$$

Model parameters	<b>b</b> <sub>1</sub>	<b>b</b> <sub>2</sub>	<b>b</b> <sub>3</sub>	<b>b</b> <sub>4</sub>
Value	74.791	-2.215	-0.975	1.71 * 10 <sup>-5</sup>

# Validation (1)

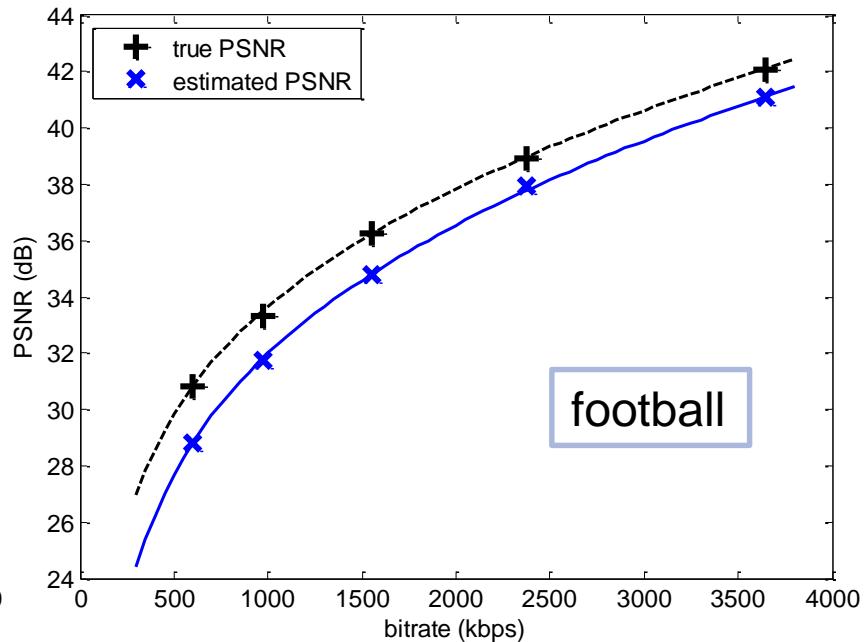
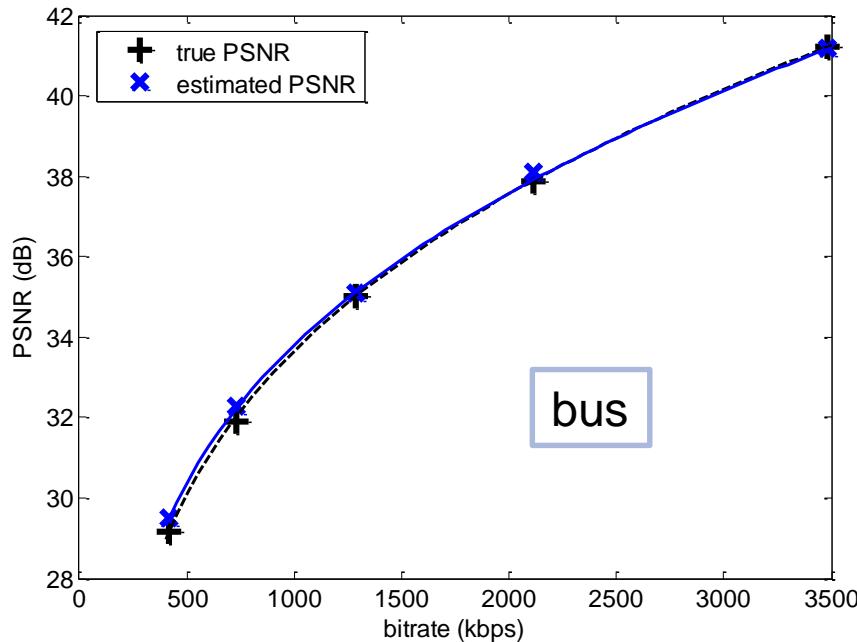


- 4 test videos: *bus, football, foreman, highway*
- Same encoding parameters as training videos

	<b>4 validation videos</b>	<b>all 12 videos</b>
RMSE [dB]	0.891	1.043
Pearson correlation	0.983	0.968

# Validation (2)

## Individual videos



	bus	football	foreman	highway
RMSE [dB]	0.246	1.447	0.244	0.981
PC	0.999	0.999	0.999	0.998

# Influence of the encoder configuration

3-way ANOVA on the PSNR

Source	Mean Sq.	df	F-value	p-value
$QP_I$	378.4	1	321.51	<0.0001
bitrate	8.06	1	6.85	0.0101
mode	0.69	2	0.59	0.5569
$QP_I$ : bitrate	18.43	1	15.65	0.0001
Error	1.177			

- Constant QP [17 ... 33]
- Rate control [500 kbps ... 2.5 Mbps]
- crf [19 ... 27]
- $8 \times 3 \times 5 = 120$  videos

	rate control	crf
RMSE [dB]	1.078	1.147
Pearson correlation	0.979	0.950

# Influence of the encoding profile

3-way ANOVA on the PSNR

Source	Mean Sq.	df	F-value	p-value
$QP_I$	1240.19	1	542.2	<0.0001
bitrate	61.87	1	26.65	<0.0001
profile	0.97	2	0.49	0.8116
$QP_I$ : bitrate	118.04	1	50.85	<0.0001
Error	2.32			

- x264 constant QP
  - Baseline
  - Main
  - High
- $12 \times 3 \times 5 = 180$  videos

	main	high
RMSE [dB]	1.163	1.168
Pearson correlation	0.964	0.963

# Influence of the encoder

3-way ANOVA on the PSNR

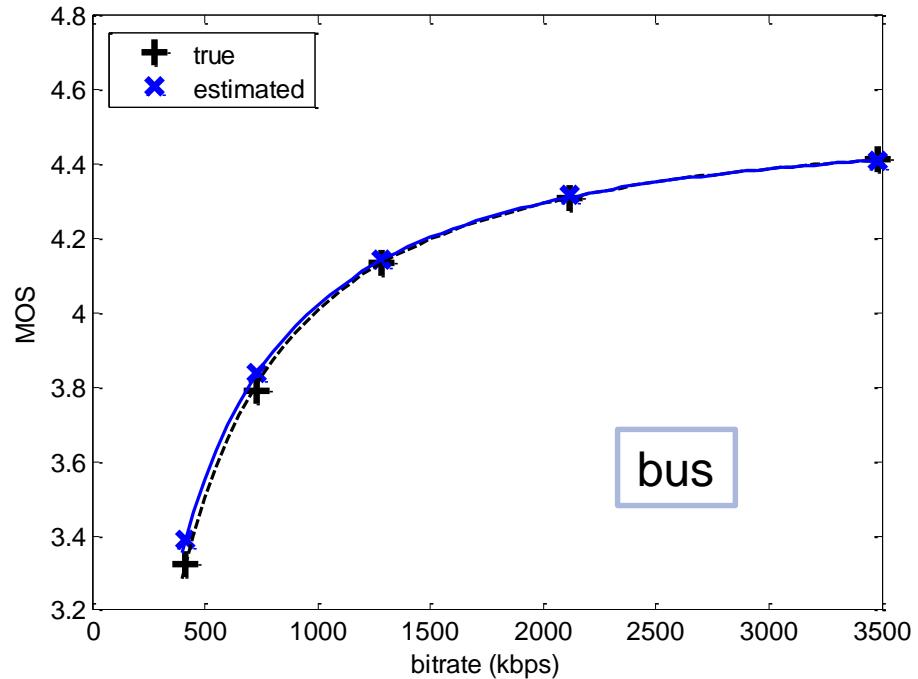
Source	Mean Sq.	df	F-value	p-value
$QP_I$	910.16	1	390.62	<0.0001
bitrate	43.33	1	18.6	<0.0001
encoder	3.18	1	1.36	0.2455
$QP_I$ : bitrate	74.44	1	31.95	<0.0001
Error	2.33			

- Baseline constant QP
- JM encoder
- $12*2*5 = 120$  videos

	JM encoder
RMSE [dB]	1.203
Pearson correlation	0.972

## Example application (1)

- Estimation of utility curves
  - Applicable to DASH videos, with multiple representations
  - PSNR to MOS using [3]
- High PC for individual videos leads to good shape estimation



[3] S. Wolf and M. Pinson, „Video quality measurement techniques“, NTIA, TR-02-392, 2002

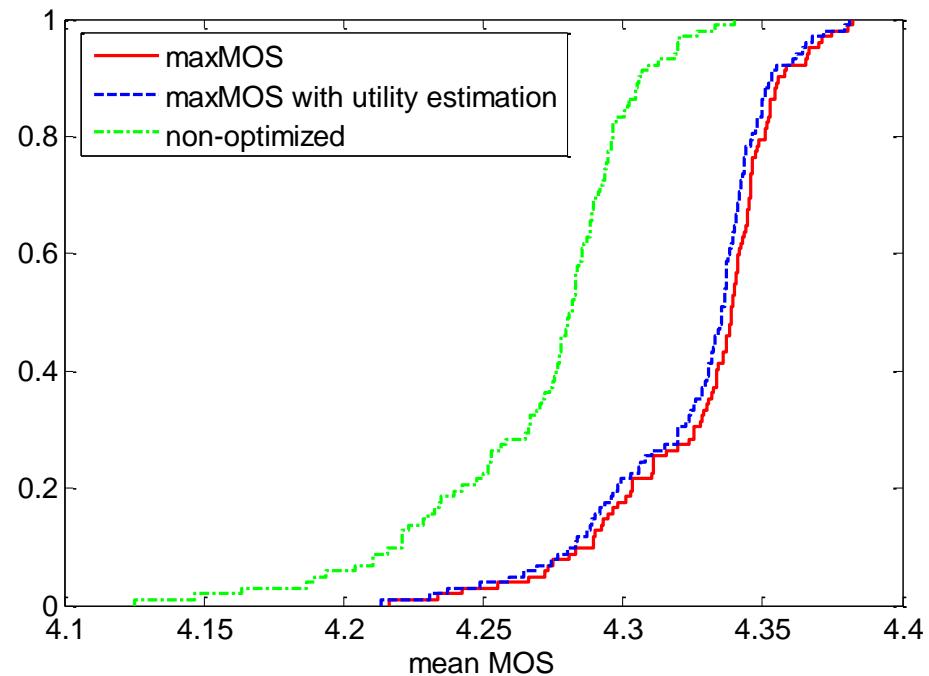
## Example application (2)

Simulated LTE cell

- Link layer LTE model
- 5 MHz bandwidth
- $K = 10$  CIF streaming users
- Radio resources allocated using utility curves as:

$$\arg \max_{(\alpha_1, \dots, \alpha_K)} \sum_{k=1}^K U_k(\alpha_k)$$

$$s.t. \sum a_k = 1$$



[4] A. El Essaili et al. "Quality-of-Experience driven Adaptive HTTP Media Delivery", IEEE ICC 2013

# Conclusion

- Novel No-Reference PSNR estimation method
- Only 2 parameters needed for estimation:
  - Bitrate of the video stream
  - Mean QP of the I-frames
- Above 0.95 Pearson correlation and around 1dB RMSE for various encoding configurations.
  
- Further Work:
  - Reduce the bias with additional information about the video content.
  - Evaluate the effect of different resolutions and different framerates.