



# Camera Networks Dimensioning and Scheduling with Quasi Worst-Case Transmission Time

Viktor Edpalm Axis Communications Alexandre Martins Axis Communications | Lund University Martina Maggio Lund University Karl-Erik Årzén Lund University



### Who are we ?

- Swedish company
- Since 1996 / Canon since 2015
- ~3000 employees & 179 countries
- Development offices:
  - Lund
  - Linköping
  - Paris
  - Shanghai
- ~835 million € sales revenue
- We do (mainly) **network cameras** !

(and we are the world's number 2 at it)







### Network dimensioning ?

#### **Dimensioning <> "long term"**

#### How to design a system of cameras?

#### How to estimate?

- Storage space ?
- Network bandwidth ?
- Peak capacity ?

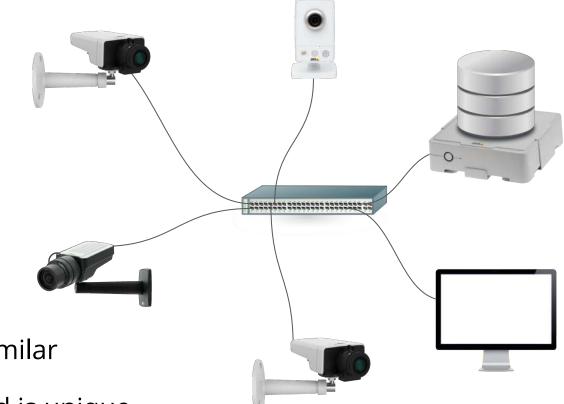
#### Knowing that:

• Each camera is unique

... but somewhat similar

• Each set-up/time period is unique

... but belongs to a limited category list



## Network scheduling ?

Scheduling <> "short term"

#### How to estimate?

- Maximum bandwidth ?
- Delay?
- Peak capacity ?

#### Knowing that:

• Each camera is unique

... but somewhat similar

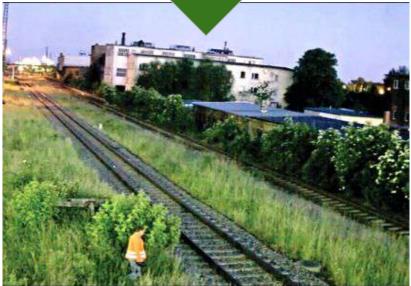
• Each set-up/time period is unique

... but belongs to a limited category list



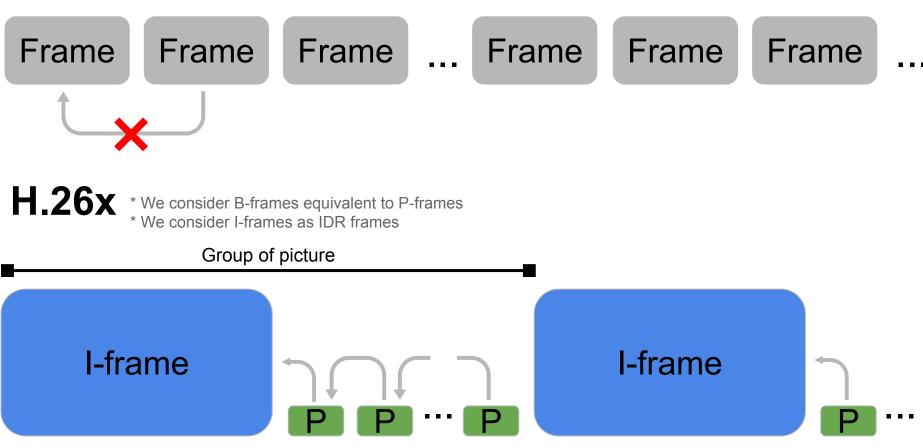




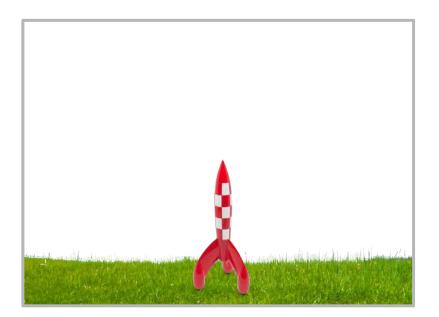




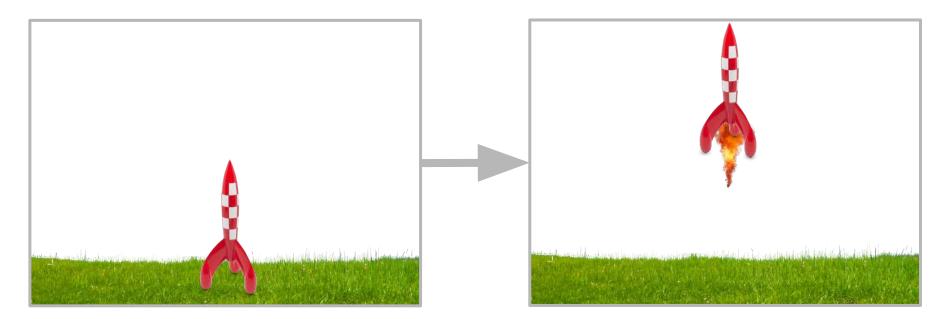
### MJPEG



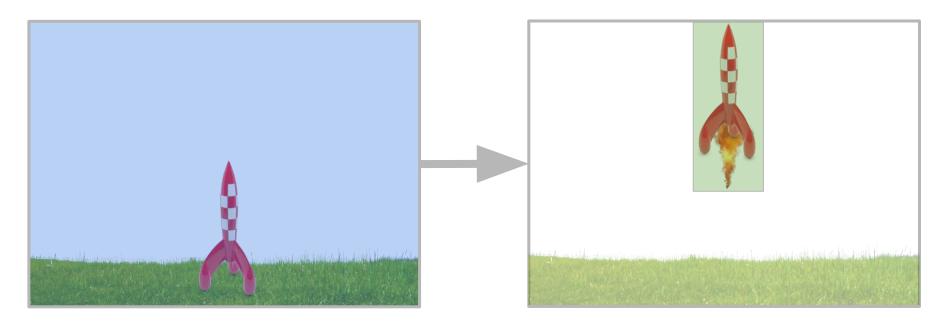


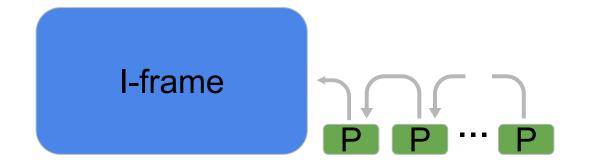




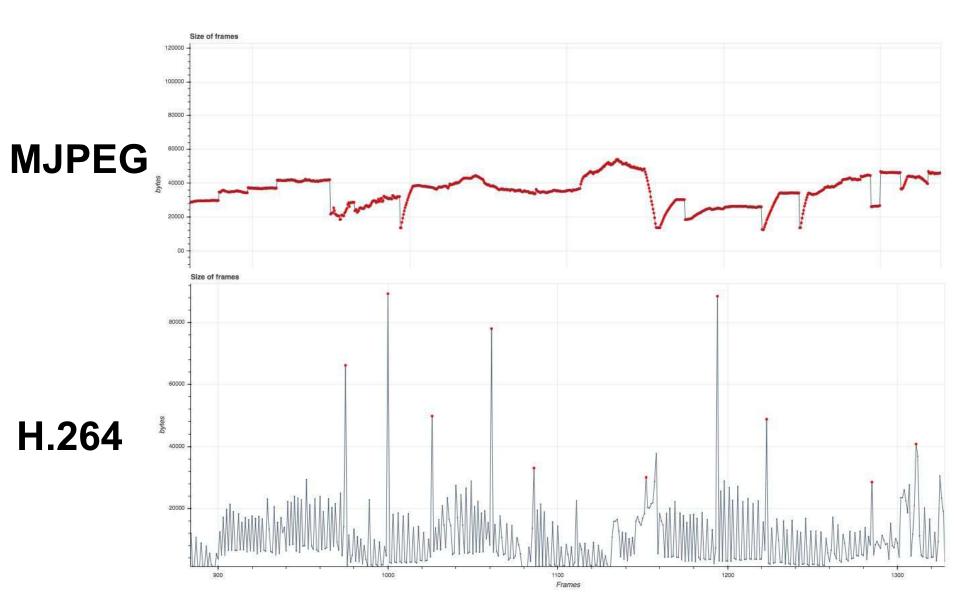










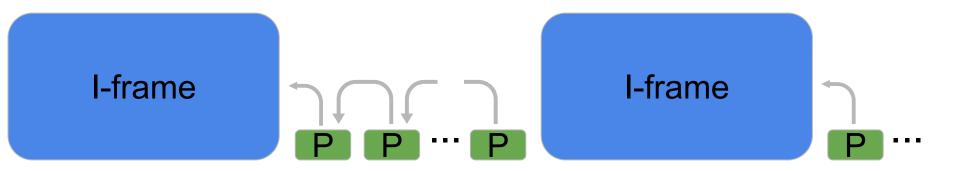


### H.264 and scheduling

Multiframe model + Many cameras:

- I-frames each generates a separate load
- B/P-frames

Paradigm: non-preemptive scheduling with multi-frame tasks

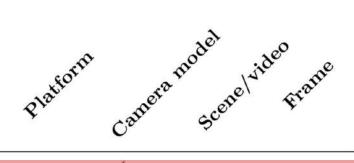


#### **Dimensioning Steps**

- Characterize a **reference camera** model: in lab, in real scenarios
- Characterize **other camera** models: in lab, (in real scenarios)
- Characterize **scene types**: in real scenarios for different cameras
- Gather scene/video parameters (motion, light level, zoom, GOP, fps...)
  - Measurable in real time
  - Estimated from usage/placement
- Extrapolate **expected frame sizes** and bandwidth

Platform Camera model frame **1.** Parameters Camera detail properties  $(D_C)$ Camera noise  $(N_C)$ Compression/QP Dynamic range factor (DR)Frame rate (FPS) Group of pictures (GOP)h, w, resolution Motion encoder efficiency  $(M_{EC})$  $\checkmark$ Motion level (ML)Nature factor  $(N_F)$ Scene detail level  $(D_S)$ Scene illumination (L)Size of Average Object (SAO)

#### 1. Parameters

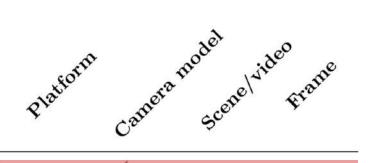


Camera detail properties $(D_C)$		$\checkmark$		
Camera noise $(N_C)$		$\checkmark$		
Compression/QP				$\checkmark$
Dynamic range factor $(DR)$			$\checkmark$	
Frame rate (FPS)			$\checkmark$	$\checkmark$
Group of pictures $(GOP)$			$\checkmark$	$\checkmark$
h, w, resolution			$\checkmark$	
Motion encoder efficiency $(M_{EC})$	$\checkmark$			
Motion level $(ML)$				$\checkmark$
Nature factor $(N_F)$			$\checkmark$	
Scene detail level $(D_S)$			$\checkmark$	
Scene illumination $(L)$			$\checkmark$	$\checkmark$
Size of Average Object $(SAO)$			$\checkmark$	$\checkmark$

measured in laboratory

measured at runtime

#### 1. Parameters

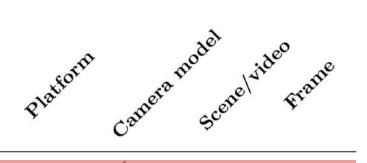


Camera detail properties $(D_C)$		$\checkmark$			-
Camera noise $(N_C)$		$\checkmark$			
Compression/QP				$\checkmark$	
Dynamic range factor $(DR)$			$\checkmark$		
Frame rate (FPS)			$\checkmark$	$\checkmark$	– Camera
Group of pictures $(GOP)$			$\checkmark$	$\checkmark$	
h, w, resolution			$\checkmark$		
Motion encoder efficiency $(M_{EC})$	$\checkmark$				
Motion level $(ML)$				$\checkmark$	
Nature factor $(N_F)$			$\checkmark$		
Scene detail level $(D_S)$			$\checkmark$		
Scene illumination $(L)$			$\checkmark$	$\checkmark$	
Size of Average Object $(SAO)$			$\checkmark$	$\checkmark$	

measured in laboratory

measured at runtime

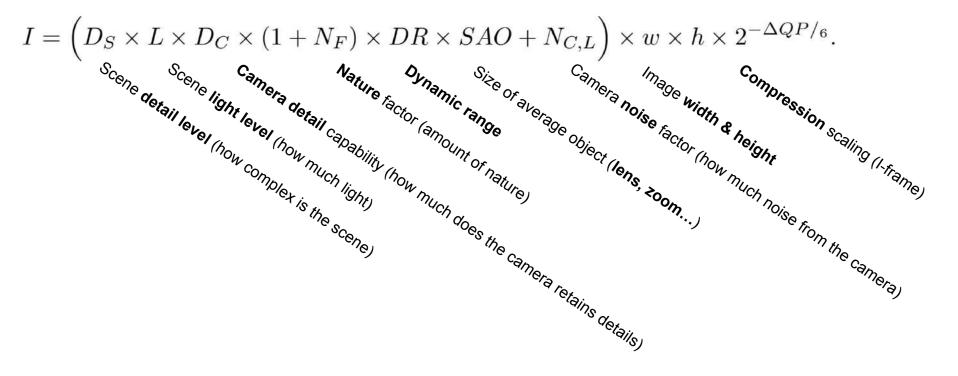
#### 1. Parameters

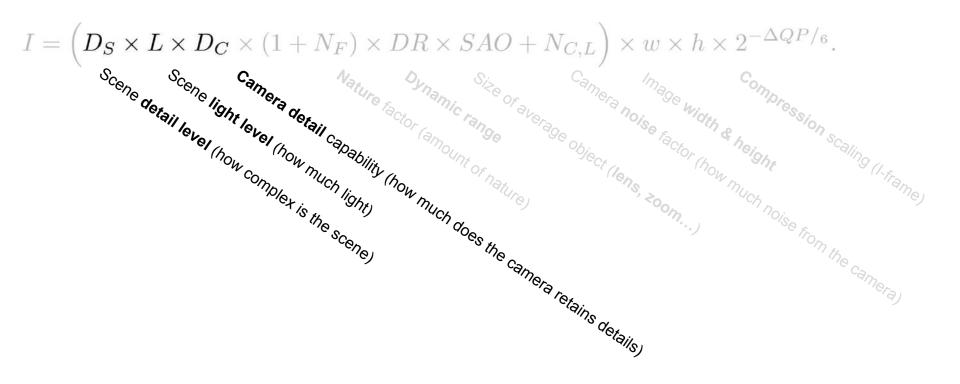


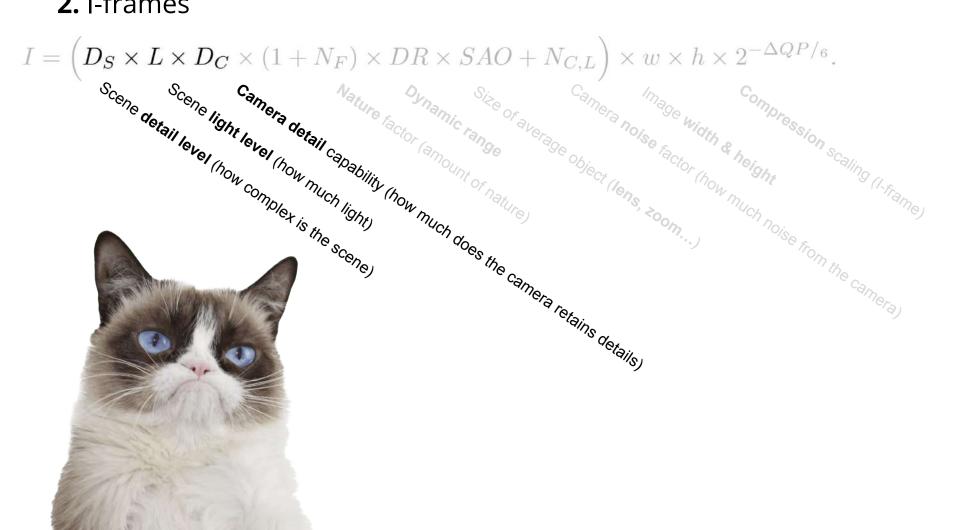
Camera detail properties $(D_C)$		$\checkmark$			-
Camera noise $(N_C)$		$\checkmark$			
Compression/QP				$\checkmark$	
Dynamic range factor $(DR)$			$\checkmark$		
Frame rate (FPS)			$\checkmark$	$\checkmark$	– Camera
Group of pictures $(GOP)$			$\checkmark$	$\checkmark$	
h, w, resolution			$\checkmark$		
Motion encoder efficiency $(M_{EC})$	$\checkmark$				
Motion level $(ML)$				$\checkmark$	
Nature factor $(N_F)$			$\checkmark$		Casina
Scene detail level $(D_S)$			$\checkmark$		Scene
Scene illumination $(L)$			$\checkmark$	$\checkmark$	
Size of Average Object $(SAO)$			$\checkmark$	$\checkmark$	

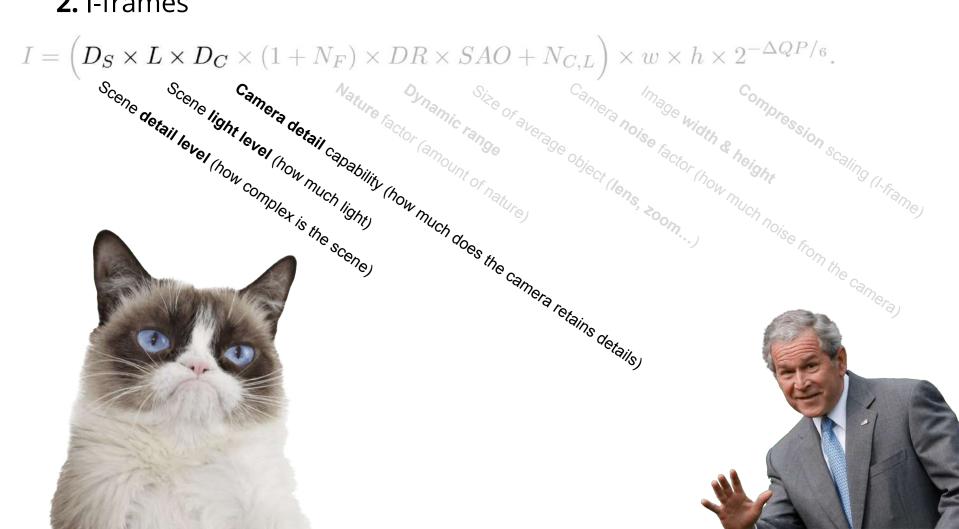
measured in laboratory

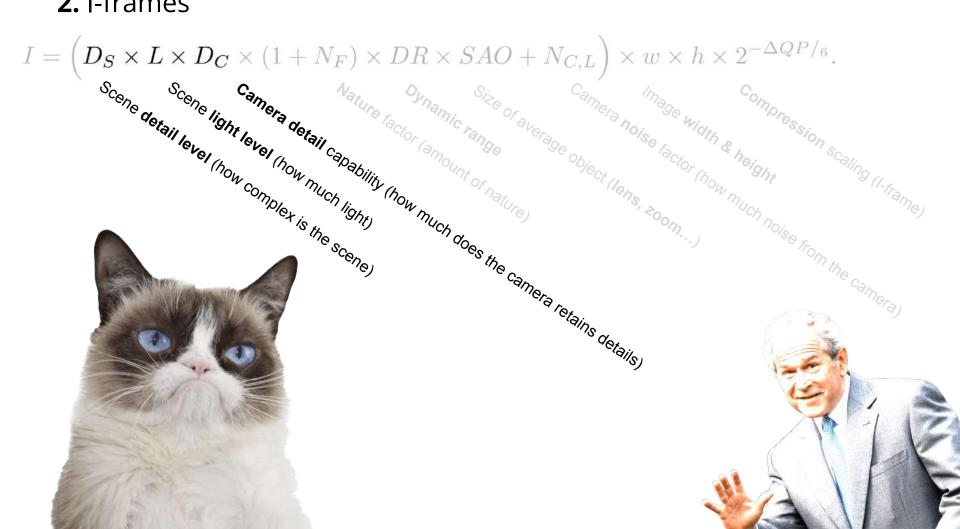
measured at runtime

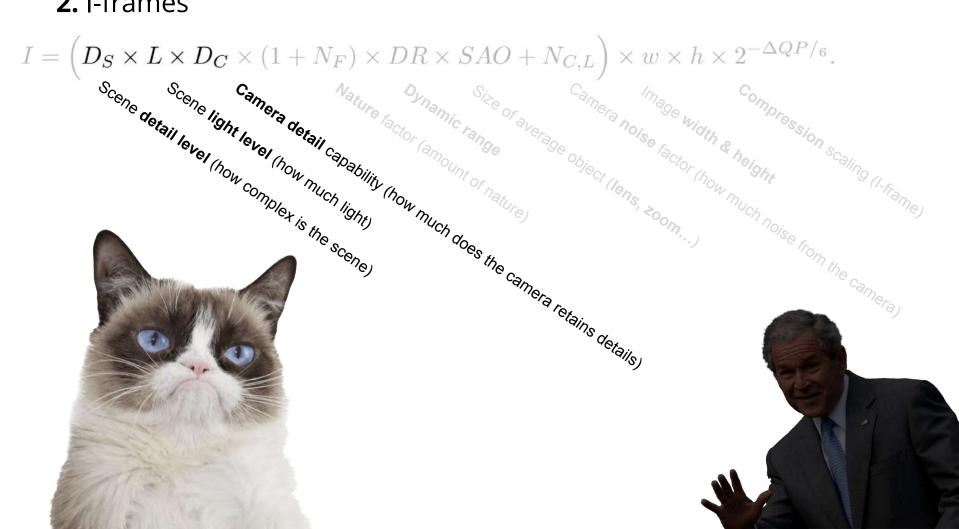


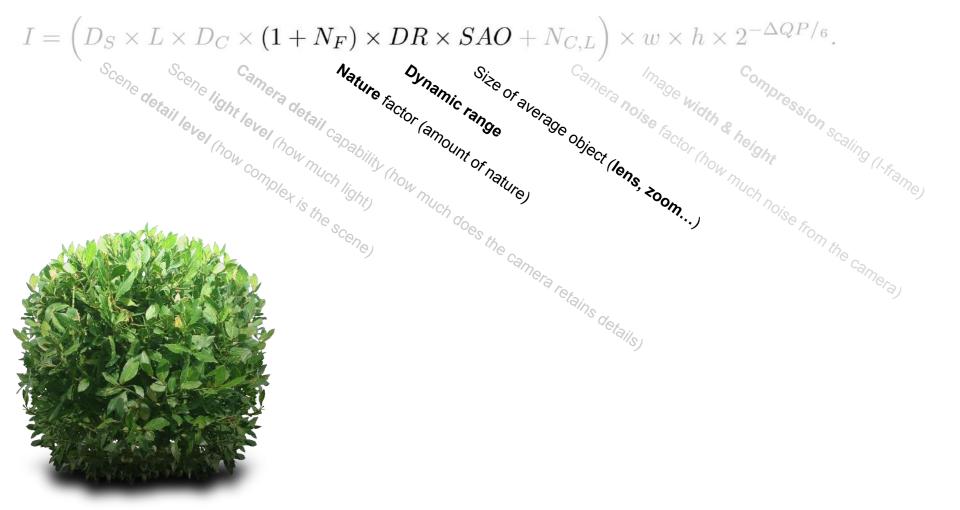


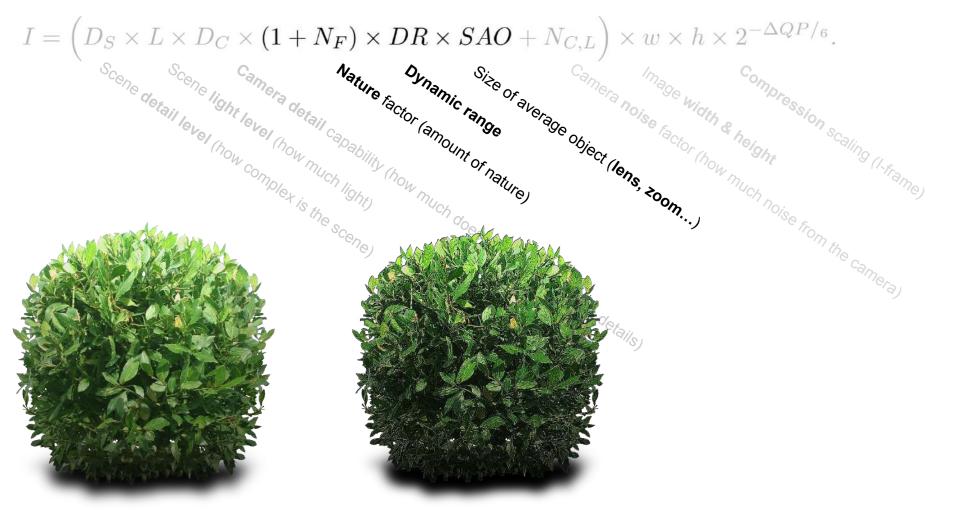


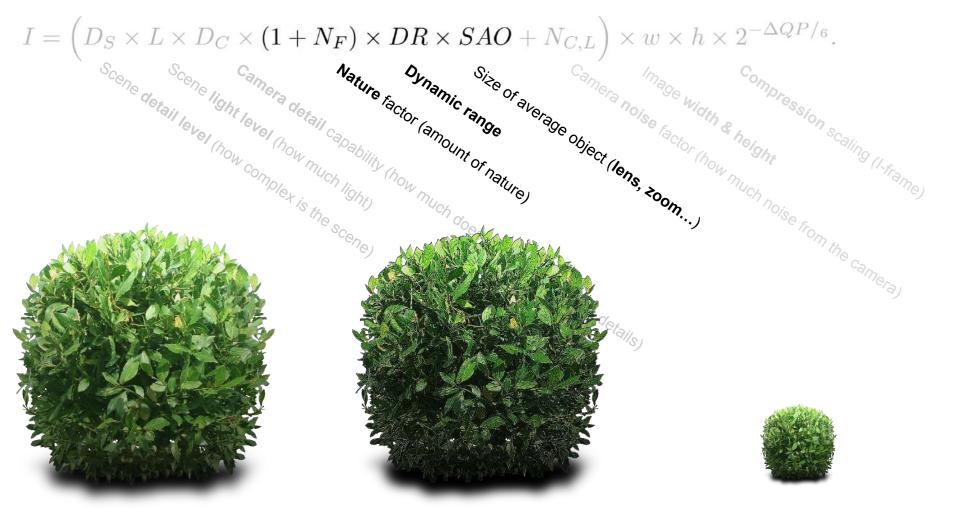


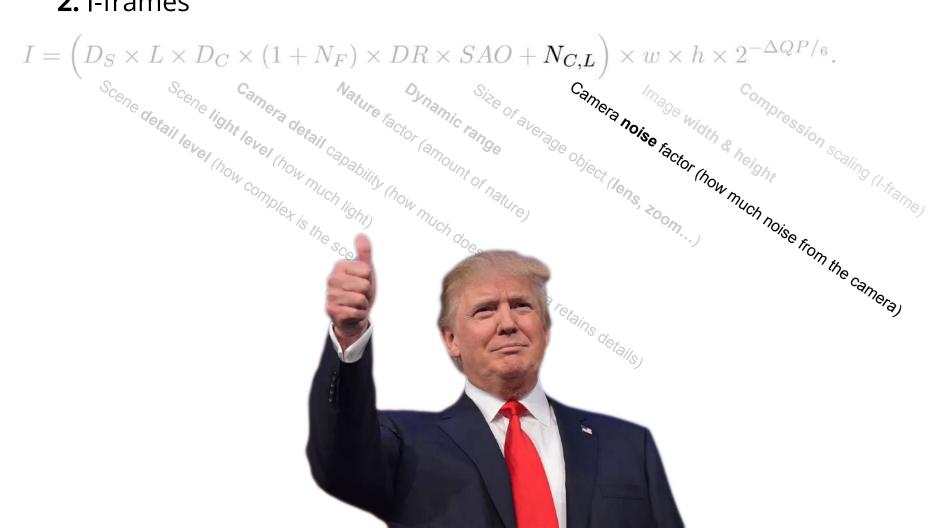


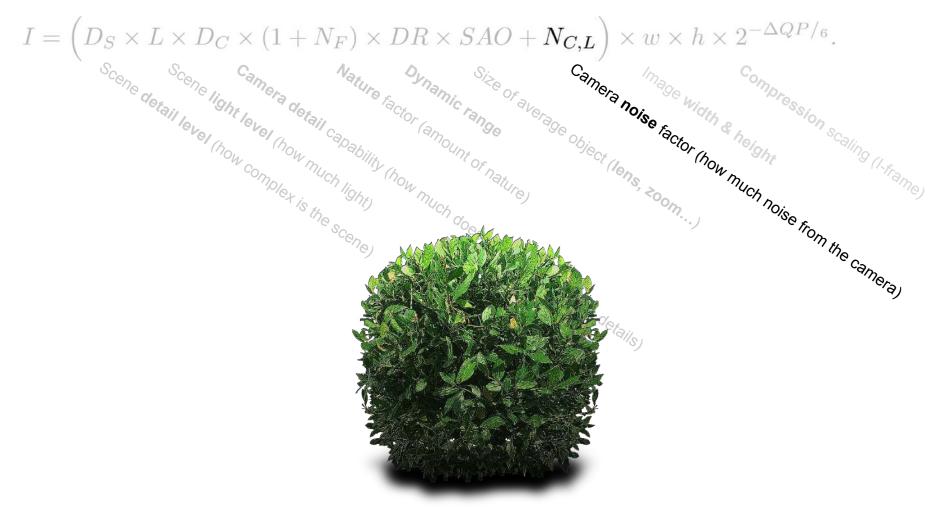


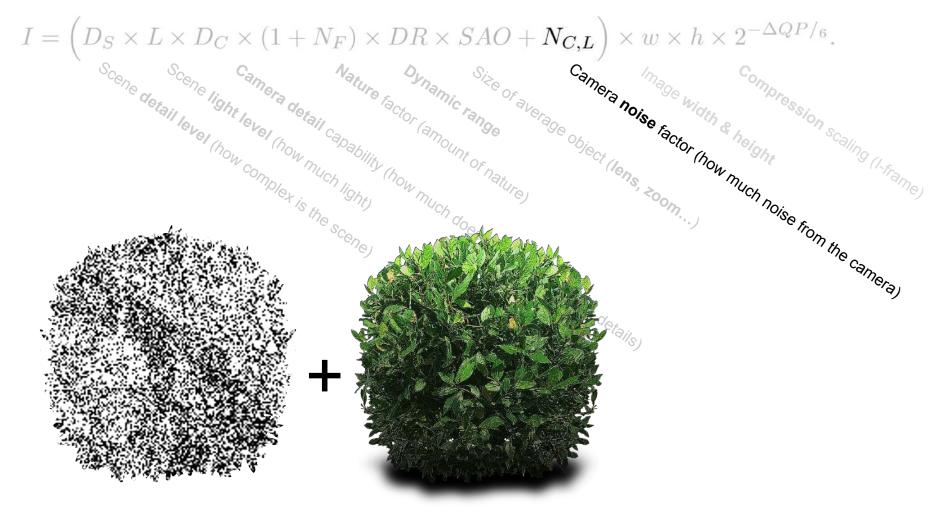


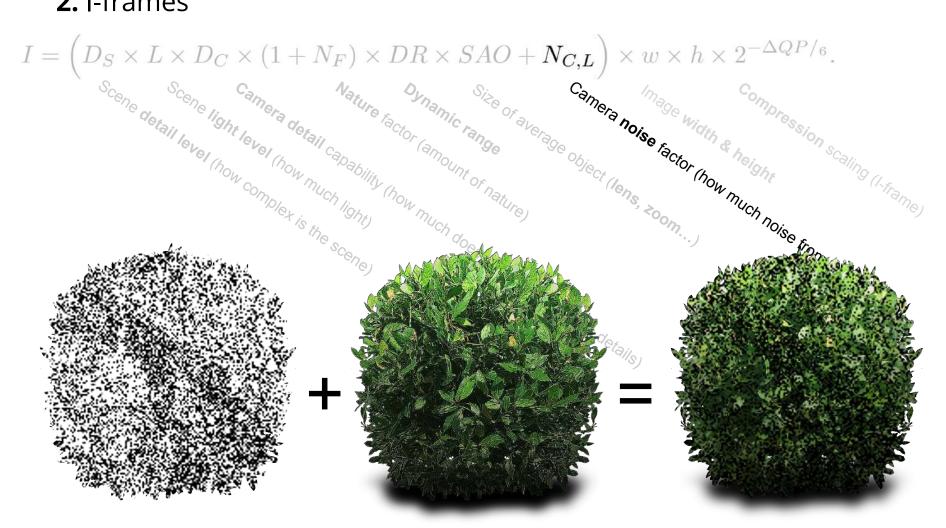


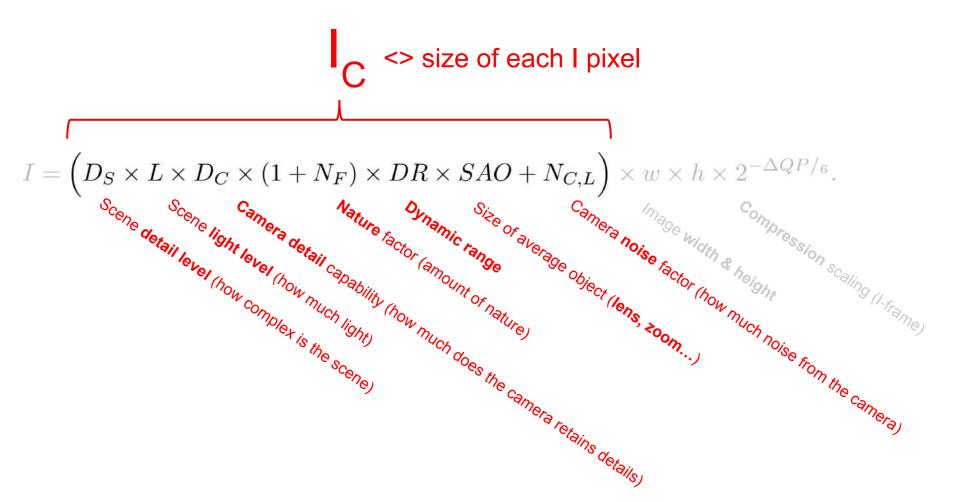


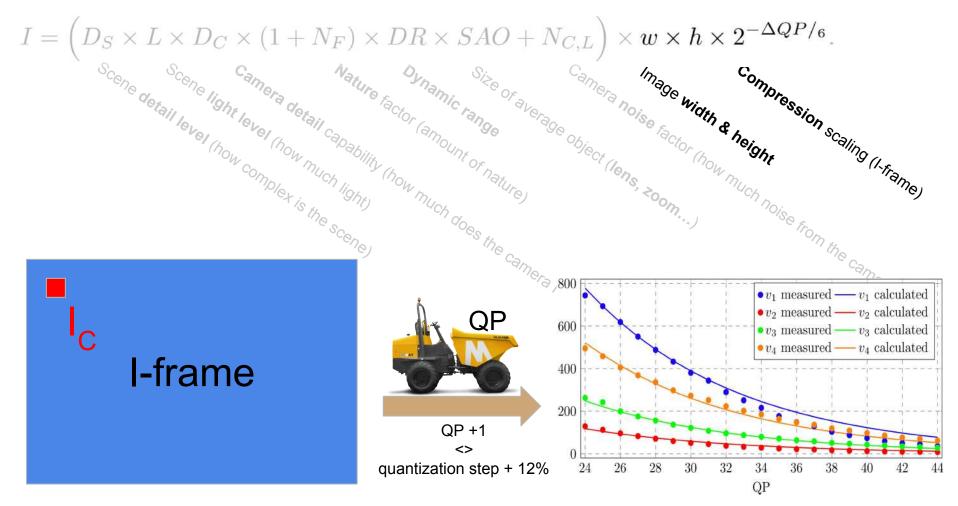


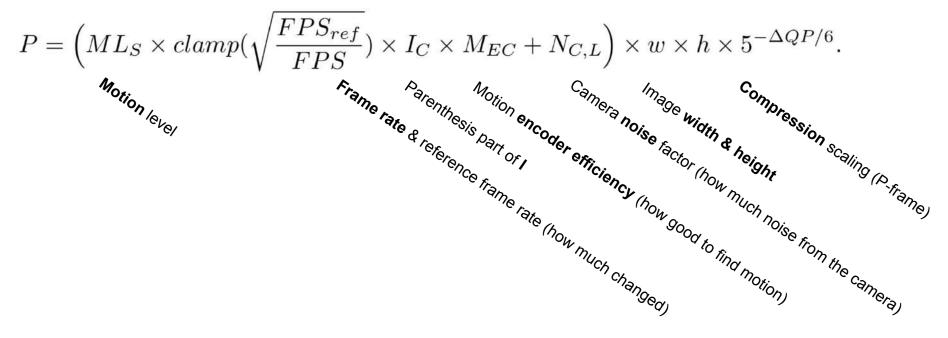


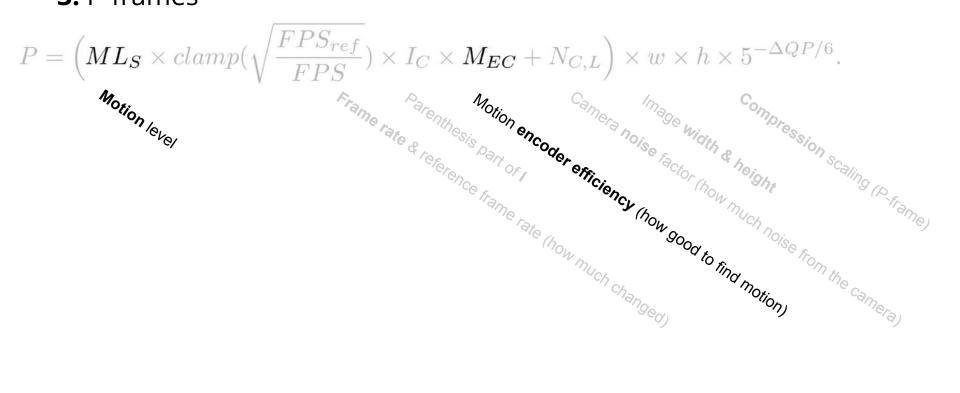


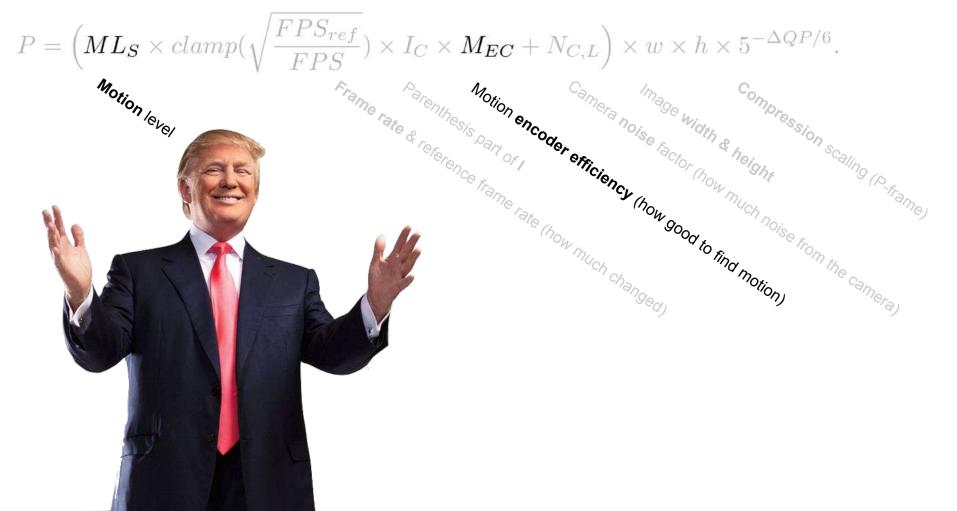


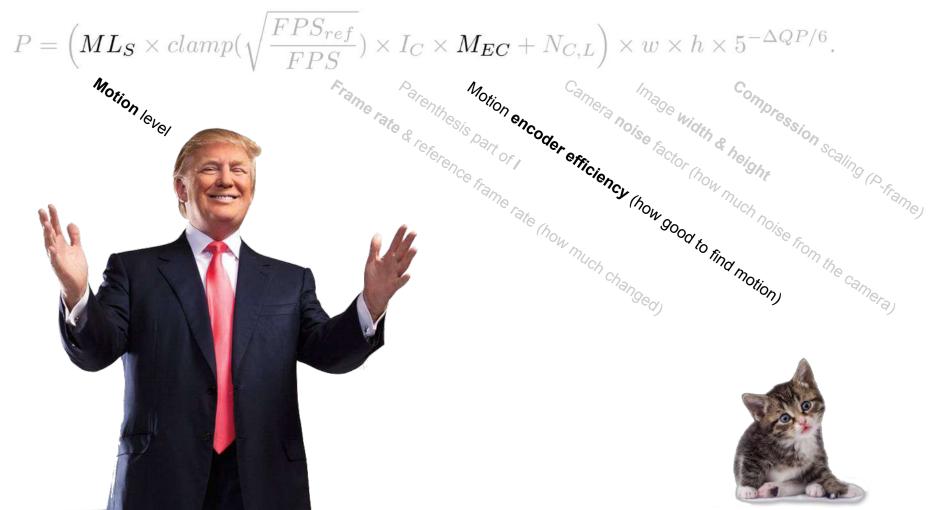


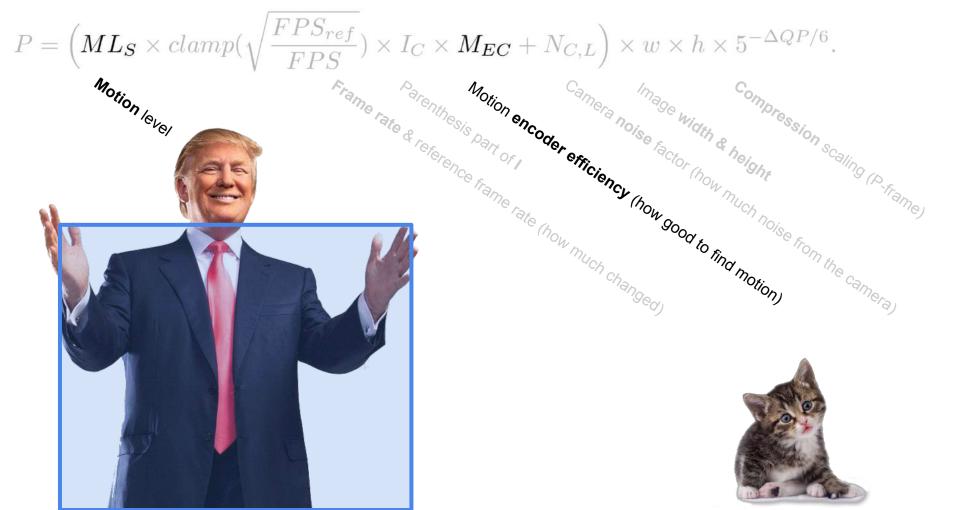


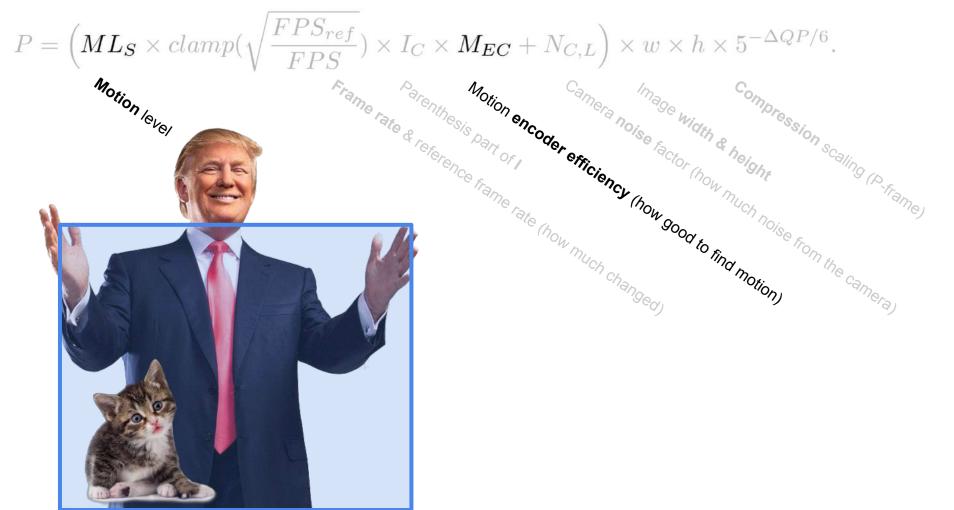


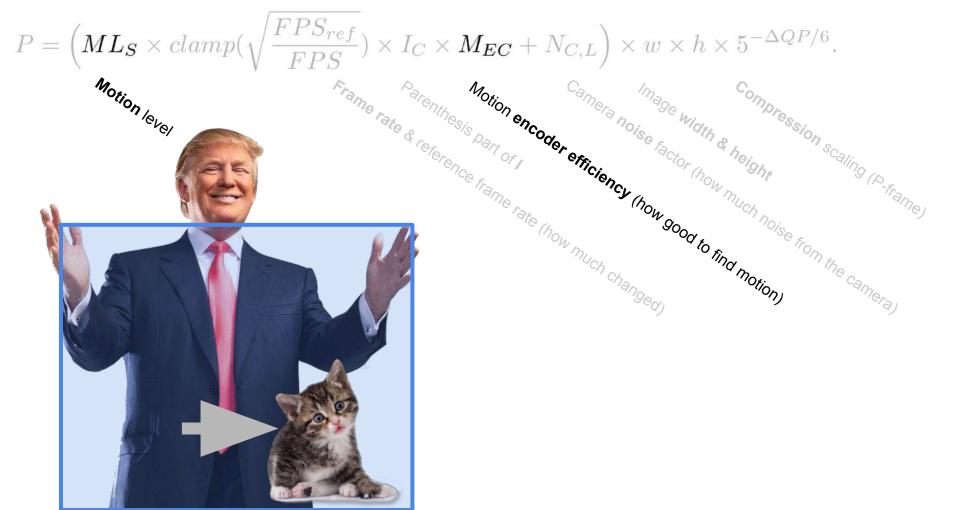


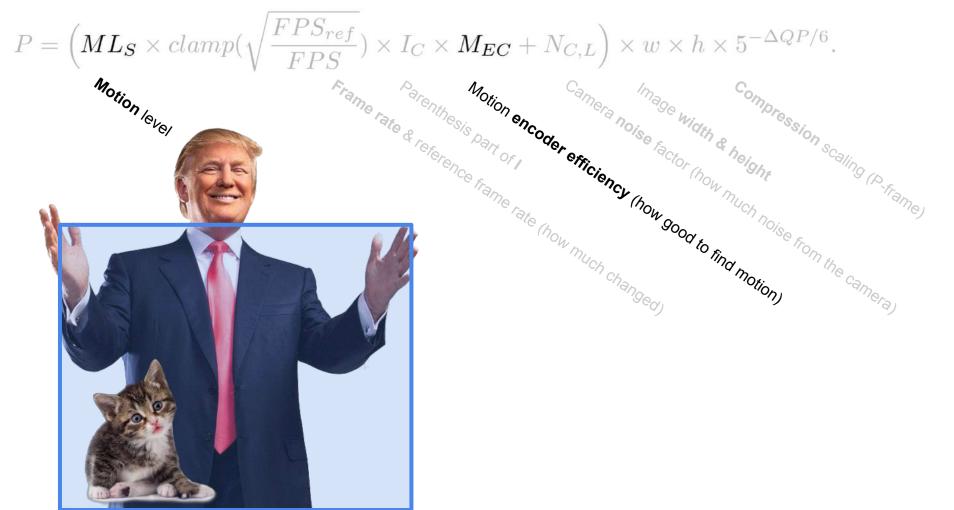


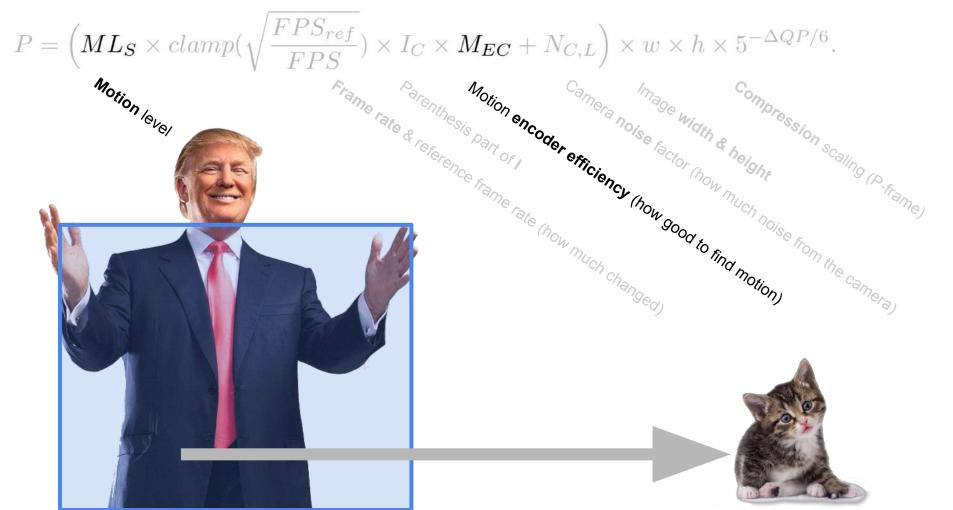


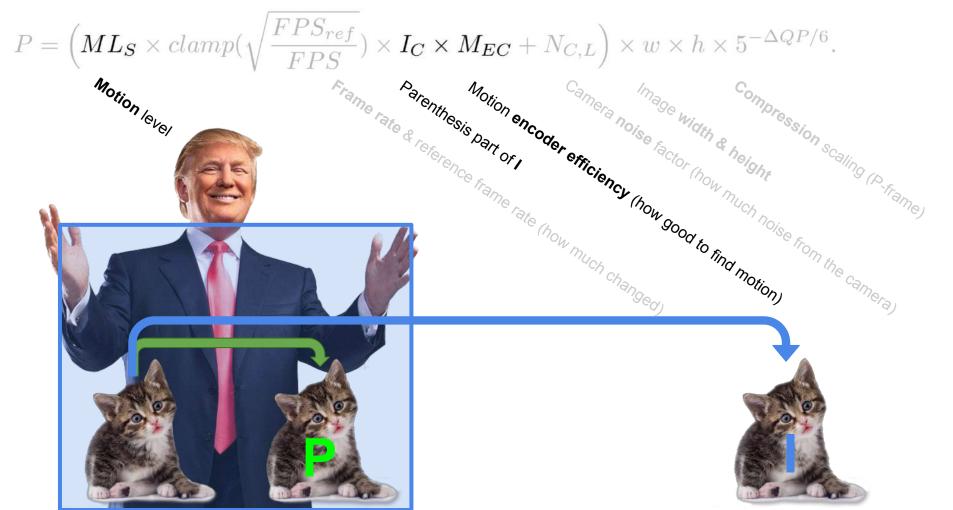


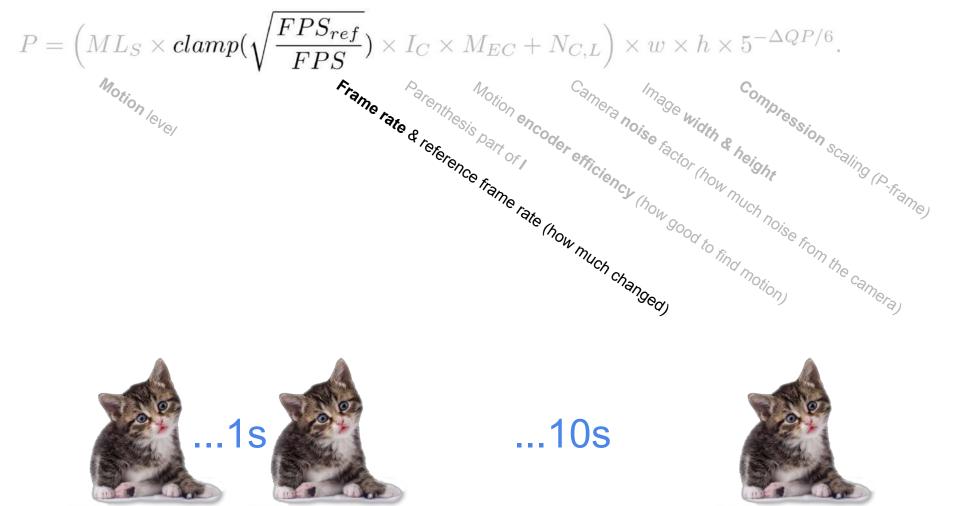






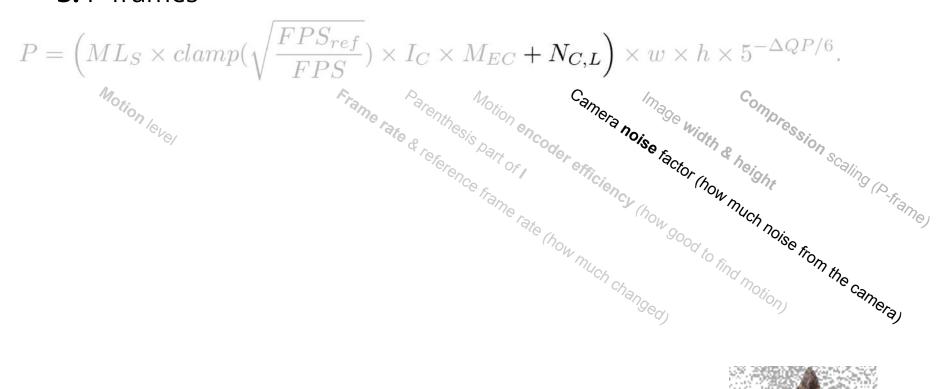




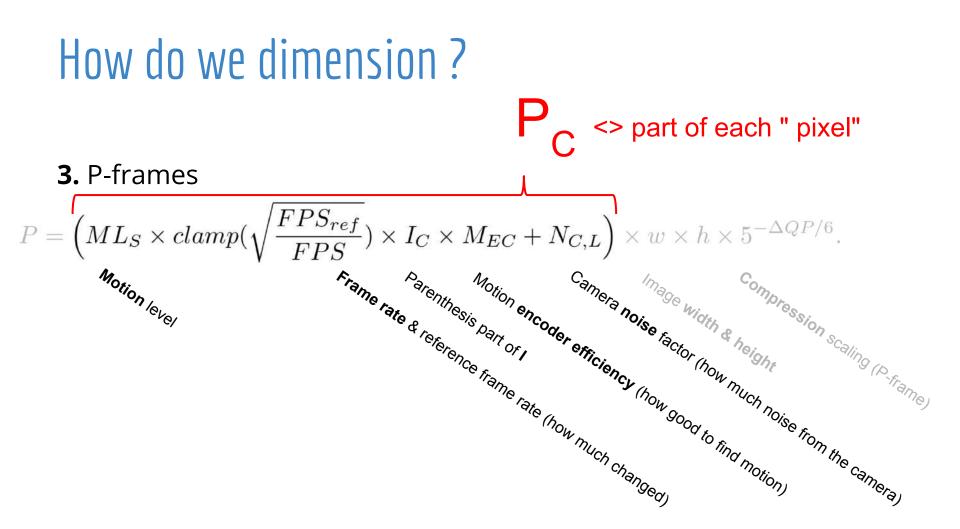


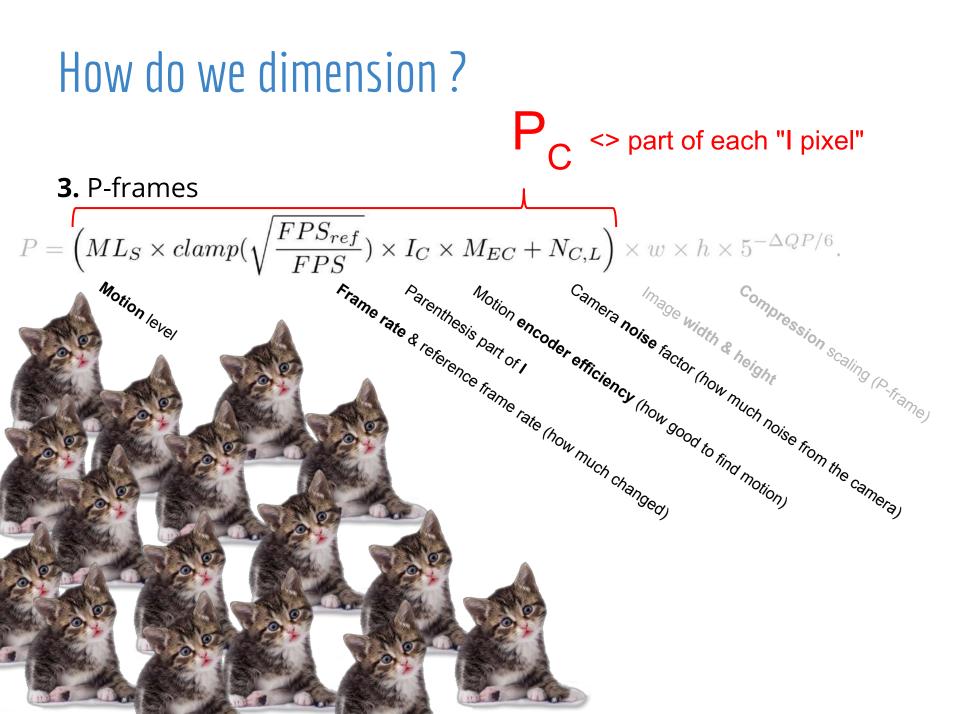
#### 3. P-frames

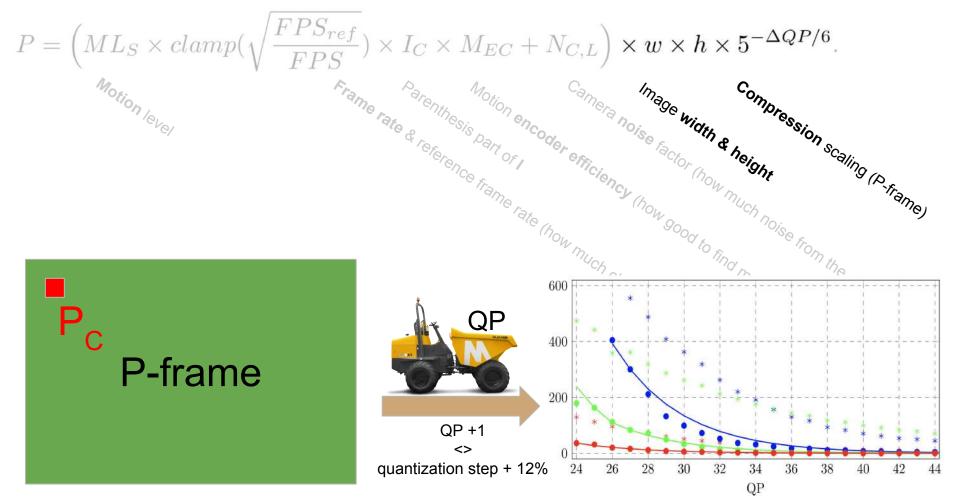
 $P = \left(ML_S \times clamp(\sqrt{\frac{FPS_{ref}}{FPS}}) \times I_C \times M_{EC} + N_{C,L}\right) \times w \times h \times 5^{-\Delta QP/6}.$ Camera noise ractor (how much noise from the camera) Frame rate & reference frame rate (how much changed) Motion encoder efficiency (how good to find motion) Parenthesis part of 1 Compression Scaling (P.frame) Motion level Keep "close" to reference: limit to [2 ... 0.5] 7 ... 30 ... 120 fps ...1s ....10s











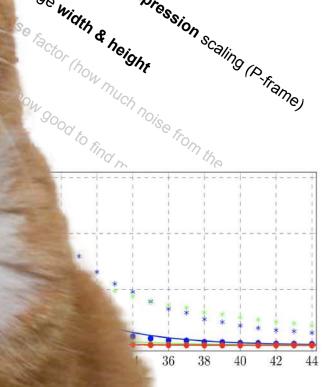
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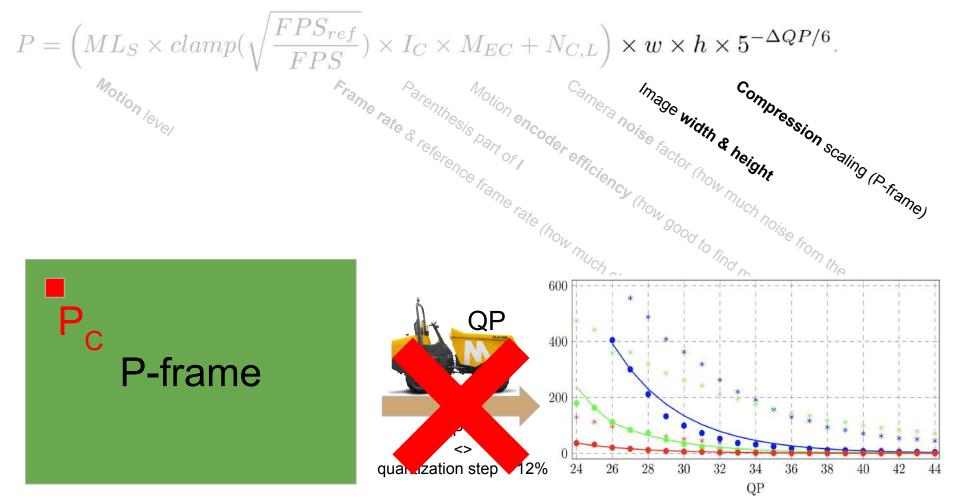
 $\times w \times h \times 5^{-\Delta QP/6}$ .

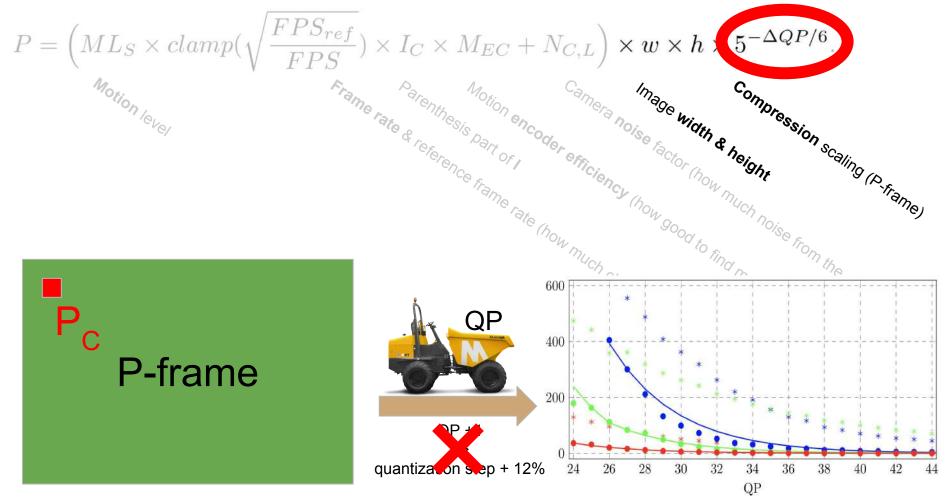
Inage width & height

P<sub>c</sub> P-frame



Compression scaling (P-frame)





$$I = \left(D_S \times L \times D_C \times (1 + N_F) \times DR \times SAO + N_{C,L}\right) \times w \times h \times 2^{-\Delta QP/6}.$$

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$$I + (GOP - 1) \times P$$

$$F = \frac{I + (GOP - 1) \times P}{GOP}$$

#### 4. Video bandwidth

$$I = \left(D_S \times L \times D_C \times (1 + N_F) \times DR \times SAO + N_{C,L}\right) \times w \times h \times 2^{-\Delta QP/6}.$$
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$$F = \frac{I + (GOP - 1) \times P}{GOP}$$

bandwidth =frame sizes  $\times$  video frame rate.

$$B = F \times FPS.$$

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bandwidth =frame sizes  $\times$  video frame rate.

$$B = F \times FPS.$$



## What is already out there ?

Two documented algorithms:

- **RQM**: Ding, W. and B. Liu (1996). "Rate control of mpeg video coding and recording by rate-quantization modeling"
- **SOTA**: Seetanadi, G. N., L. Oliveira, L. Almeida, K.-E. Arzen, and M. Maggio (2017). "Game-theoretic network bandwidth distribution for self-adaptive cameras.
- + multiple closed source security industry solutions (H.264?)

RQM (MPEG)

$$s(q_r) = \alpha + \beta \cdot 1 / q_r \gamma$$

s: frame size q<sub>r</sub>: compression level (1..31) α: constant (overhead bits)

β: constant (resolution & motion) γ: constant (frametype)

**SOTA** (MJPEG)

$$s(q_l) = q_l \cdot s_{max}$$

s: frame size  $q_i$ : compression level (0.01..1) ;  $s_{max}$ : "raw" max frame size

### **Bandwidth prediction**

#### Tests

- 1. Verification of the prediction using recorded material available at Axis communications
- 2. On-site test in a hotel complex

#### **Bandwidth prediction**

#### Relative error %

-	Scenario	MODEL	SIMP	SOTALIN	SOTAEXP	
Parking lot no motion	n <b>1</b>	2.88%	12.96%	18102.64%	1853.12%	
Parking lot low motion	on <b>2</b>	3.70%	9.36%	11731.72%	1169.53%	
Parking lot high moti	ion <b>3</b>	14.49%	26.55%	5815.86%	534.77%	
Parking lot no motion	n <b>4</b>	1.17%	16.57%	13762.27%	1164.47%	
Parking lot low motion	<sup>on</sup> 5	8.13%	0.98%	10373.71%	855.38%	
Parking lot high moti	ion <b>6</b>	3.60%	8.43%	6184.23%	473.23%	
Highway	7	3.57%	30.76%	1793.07%	103.12%	
Highway	8	17.86%	37.51%	6660.98%	625.45%	
Fence	9	6.17%	22.18%	1038.00%	133.95%	
Fence	<b>10</b>	7.14%	69.19%	247.72%	28.51%	
4k street	11	9.48%	157.25%	1984.79%	180.55%	
2k fence	<b>12</b>	3.93%	3.03%	14076.62%	1193.15%	
Road crossing	13	18.57%	48.22%	5272.21%	354.30%	

Scenario numbers come from H.264 Video Frame Size Estimation: http://bit.ly/2LvcWtS

### **Bandwidth prediction**

Industrial field test at a hotel complex, against 5 commercial bitrate estimations (obfuscated).

11 scenarios, ~5 days recordings.

-	Scenario	MODEL	SIMP	EXT 1	EXT 2	EXT 3	EXT 4	EXT 5
Reception	14	1.48%	7.06%	54.78%	40.92%	134.02%	44.39%	170.03%
Exit door	15	167.48%	806.24%	1678.17%	1518.85%	2588.49%	1558.73%	3002.18%
Office	<b>16</b>	47.67%	255.90%	470.10%	419.02%	761.96%	431.81%	894.59%
Street corr	ner <b>17</b>	10.91%	81.79%	106.88%	88.34%	212.79%	92.98%	260.92%
Reception	18	75.24%	104.66%	169.61%	145.46%	307.64%	151.50%	370.37%
Mall	<b>19</b>	51.40%	9.14%	84.40%	67.88%	178.81%	72.02%	221.71%
Elevator	20	11.37%	70.20%	50.19%	36.74%	127.08%	40.10%	162.02%
Exit door	<b>21</b>	207.49%	745.77%	1254.80%	1133.41%	1948.37%	1163.79%	2263.57%
Parking lot	t <b>22</b>	9.56%	15.40%	51.00%	70.22%	82.89%	83.87%	83.17%
Parking lot	t <b>23</b>	17.33%	49.91%	62.27%	47.73%	145.35%	51.37%	183.10%
Parking lot	t <b>24</b>	17.43%	15.22%	105.64%	131.82%	149.07%	150.42%	149.46%

#### Relative error %

Scenario numbers come from H.264 Video Frame Size Estimation: http://bit.ly/2LvcWtS

#### **Bandwidth prediction**

#### **Average relative errors**

Our model: 29%

SOTA: 2100%

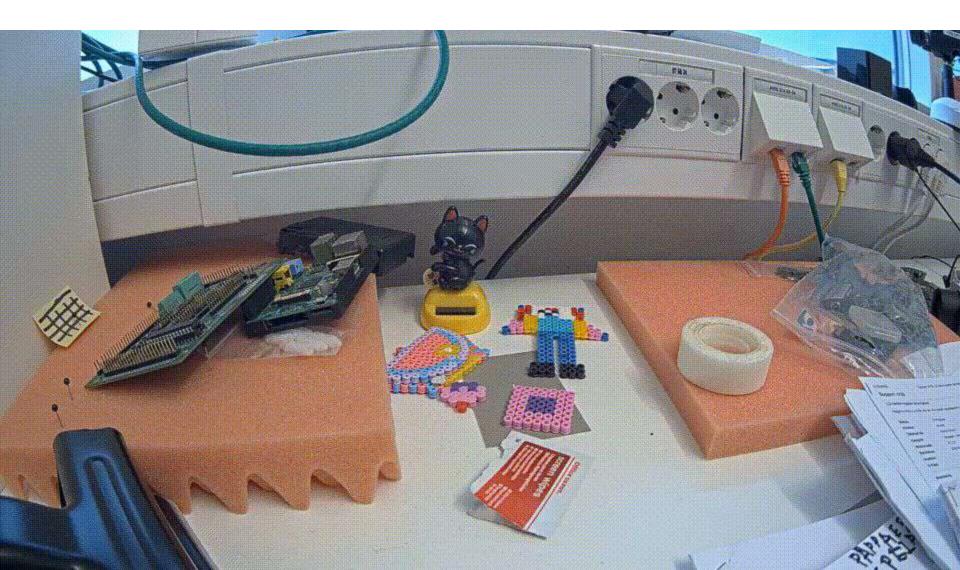
Commercial models: 336%

### **Scheduling prediction**

#### Tests

1. Verification using recorded material with triggered motion amount

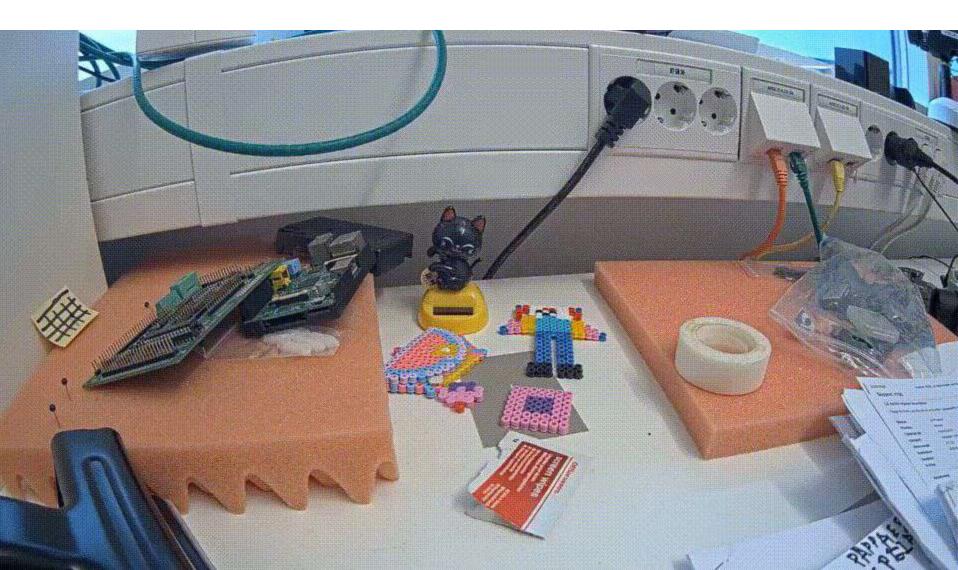
### How do we perform ? Low Motion



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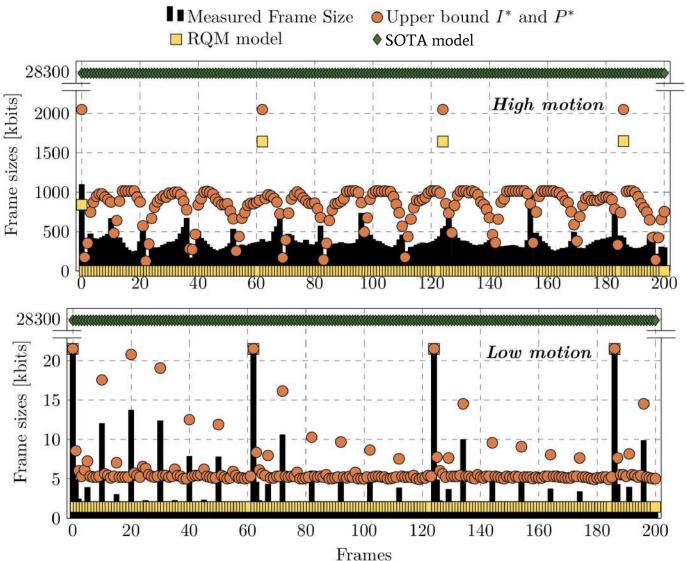


### How do we perform ? High Motion



### How do we perform ? High Motion





### **Scheduling prediction**

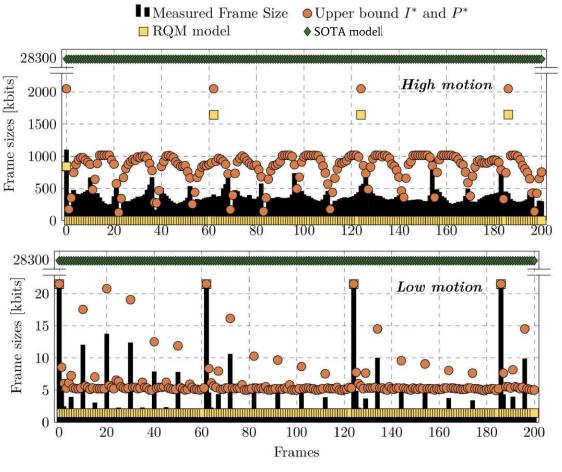
RQM performs sometimes better

... but is a posteriori

... underestimates sometimes

#### SOTA over-estimates A LOT

(as expected)



# Take away



- Model set of cameras with a taskset model.
  - allow reuse scheduling results from multiframe model.
  - better model for worst case transmission time.
  - simple enough to be run on camera.

- Perform better (by a magnitude) than competitors.
  - both academic and industrial.
  - tested in real life scenario.
  - deployed and running for year(s).



Did you find the cookie monster?

