

2020
embedded
VISION
summit®

Using an ISP for Real-time Data Augmentation

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Pony.ai
Sept 2020



Contents

Introduction to Pony.ai

Exposure in photography and computer vision

Experiments with exposure control

Conclusions and problem formulation

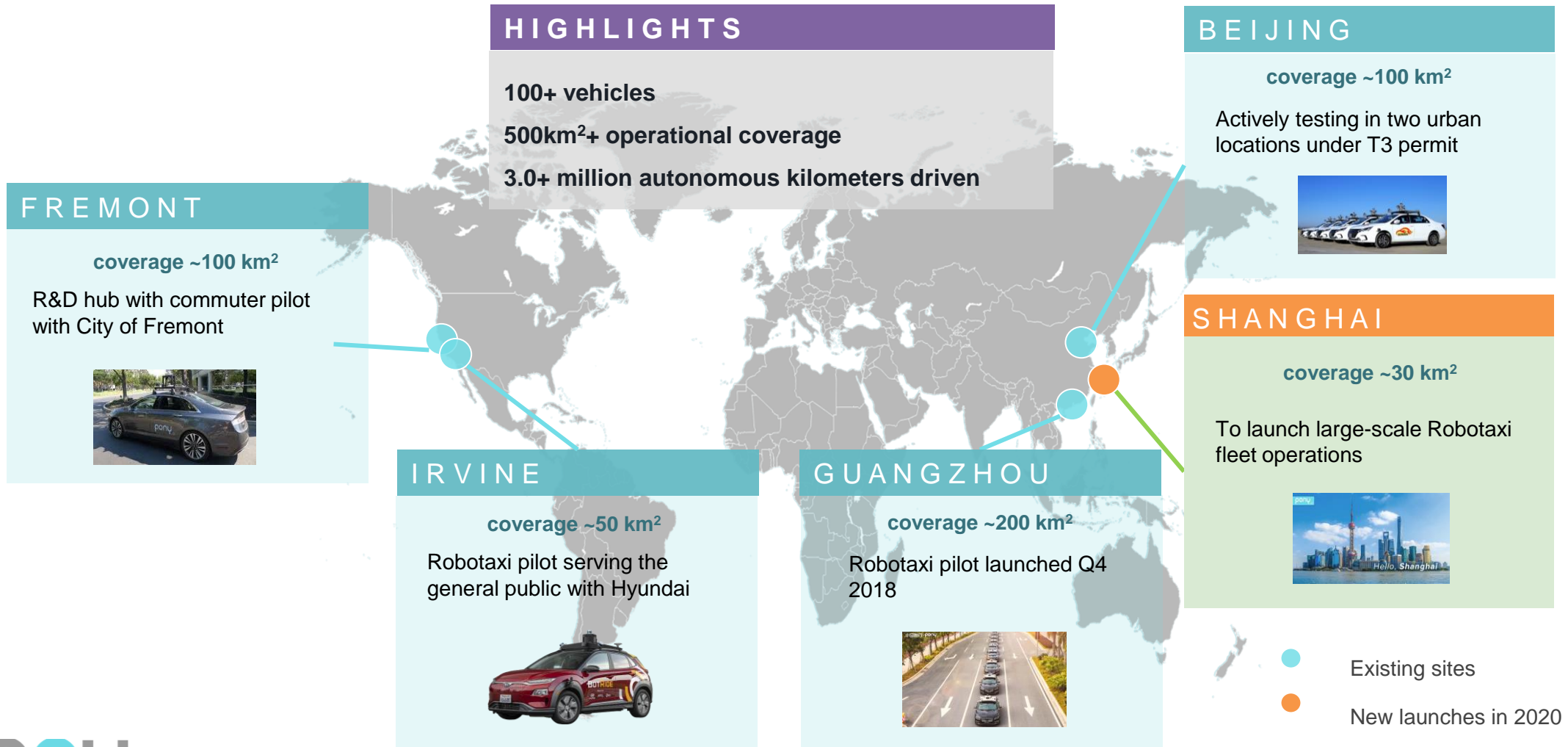
Proposed solution with automotive ISP



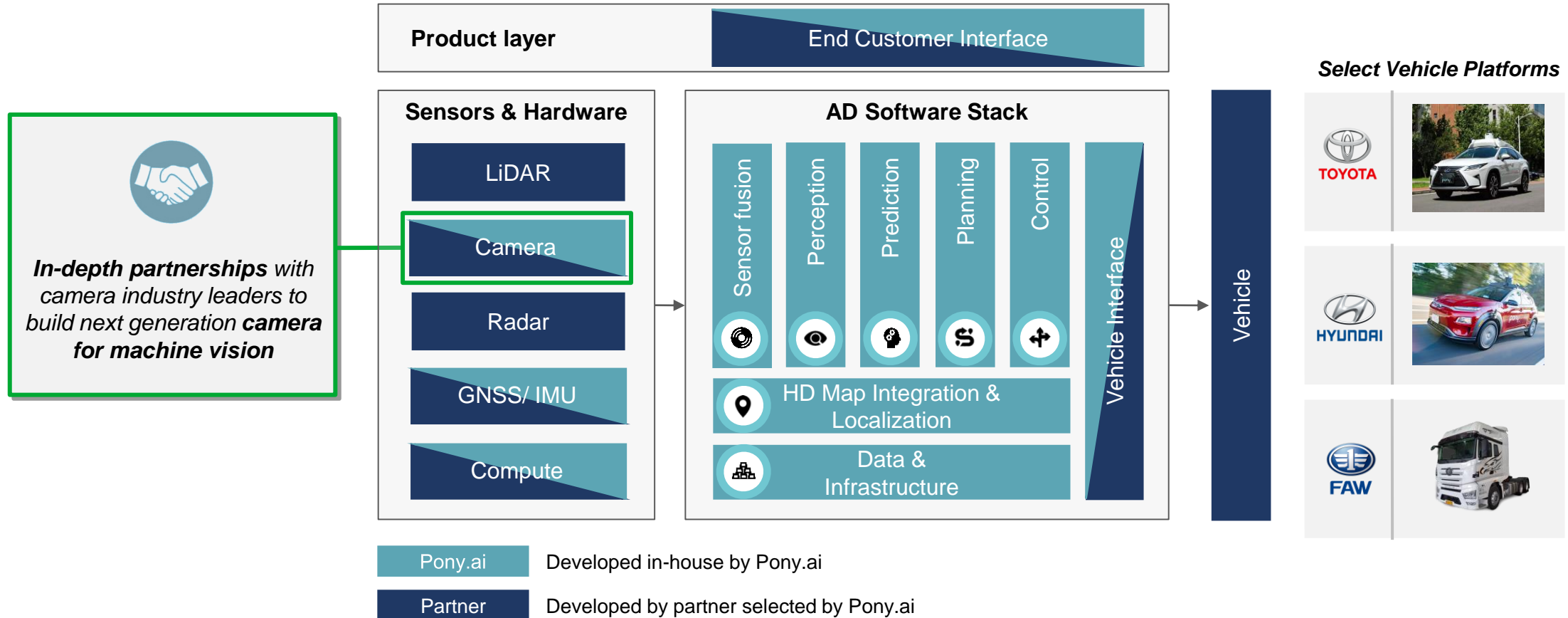
Pony.ai



Pony.ai – global autonomous technology leader



Core technology: vehicle-agnostic “virtual driver”



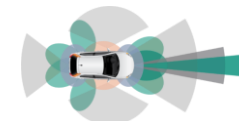
Collaboration between Pony.ai and ON Semiconductor



About ON Semiconductor

- Global semiconductor company with ~35,000 employees
- Broad automotive portfolio, segment contributing 33% of total revenue
 - #1 automotive image sensor provider globally

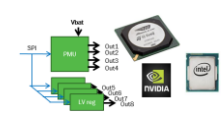
ONsemi Products Enabling AD



Active Safety
Solutions



In-Vehicle
Networking



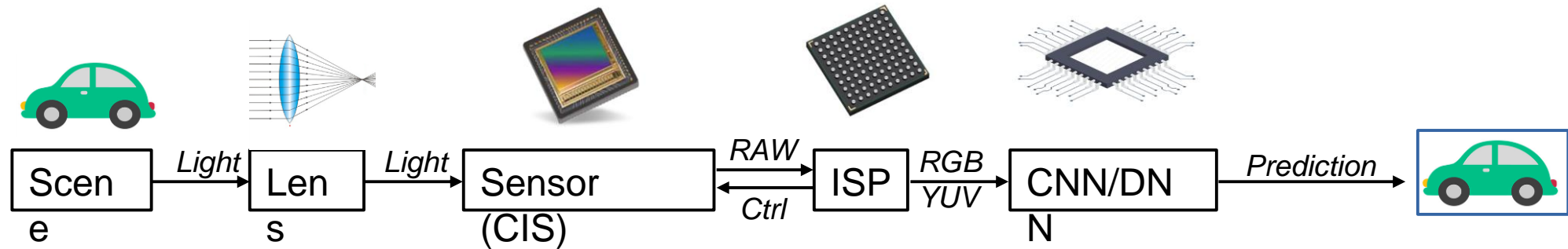
ASIL Compliant
Power Management

Scope of Collaboration

- Collaborate on developing next-generation image sensing and processing technologies for machine vision and AI
- Today's presentation covers one of the collaboration topics *"Using an ISP for real-time data augmentation"*

Exposure

Perceptual CV pipeline architecture



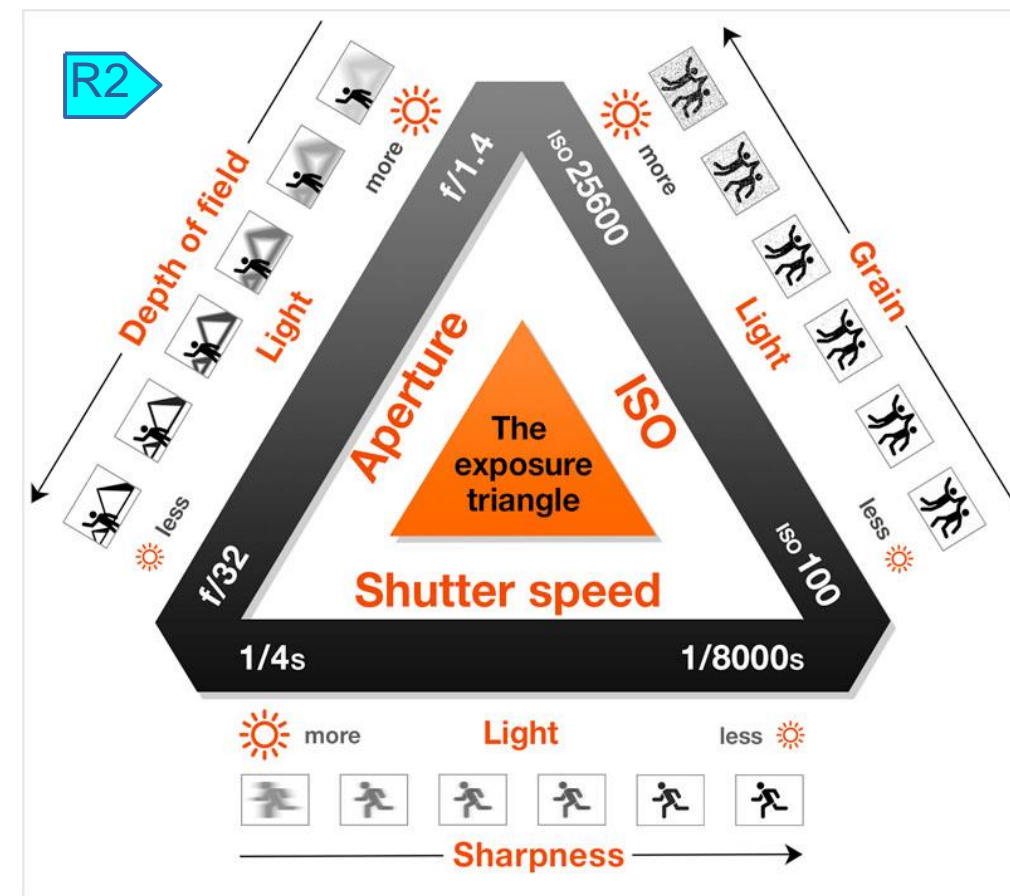
Exposure in photography

R1 Exposure is the amount of light per unit area reaching a frame of photographic film or the surface of an electronic image sensor

Exposure is determined by shutter speed, lens aperture, and scene luminance.

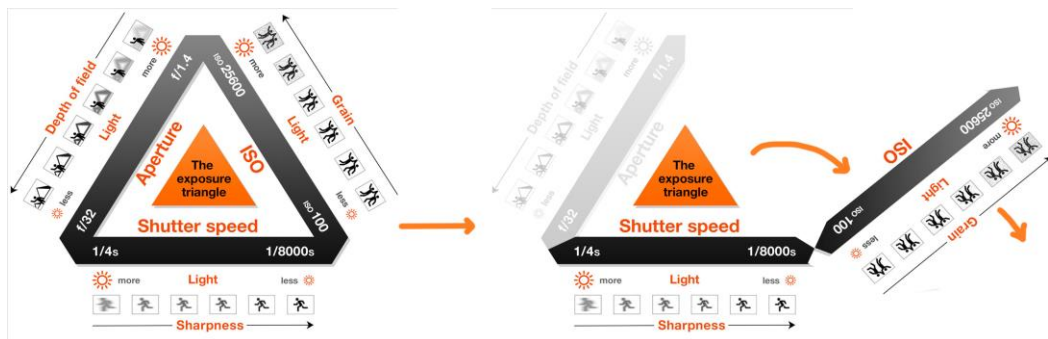
Exposure control is fundamental in photography. 

"Correct" exposure (in photography) may be defined as an exposure that achieves the effect the photographer intended.

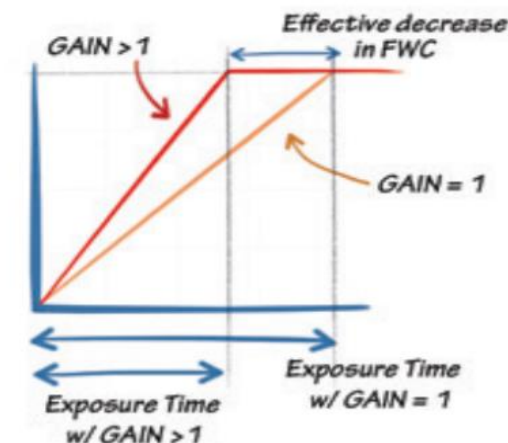


Exposure in CV and automotive

Lens aperture is fixed (no moving parts)



R1



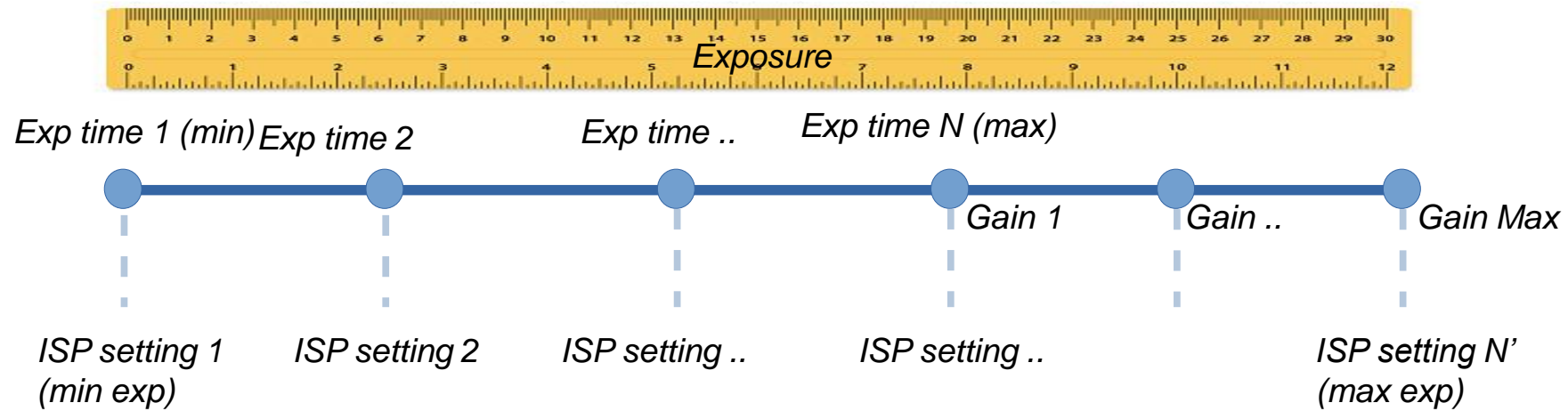
ISP (image signal processor) controls:

- Exposure time ET (integration time/shutter speed)
- Gain (ISO / film sensitivity)

based on scene light metering and history (statistics)

Exposure time is first priority. Gain compensates for dark image with short ET

Exposure in automotive and CV



Exposure time is first priority. Gain compensates for dark image with short ET

ISP usually has different settings for highest and lowest exposure points on the spectrum above (sometimes in between)

Affect Dynamic Range (DR), tone reproduction

Noise/SNR

Affects sharpness, motion blur,

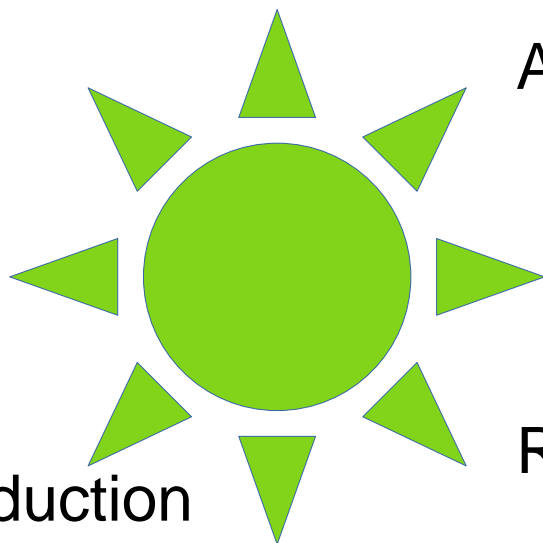
controls image sensor

Quantization

Color accuracy and reproduction

Responsible for under/over saturation

Drives ISP setting



Exposure control is crucial for object detection and safety of AV

Lab experiment 1: Automatic Exposure

Setup

Camera:

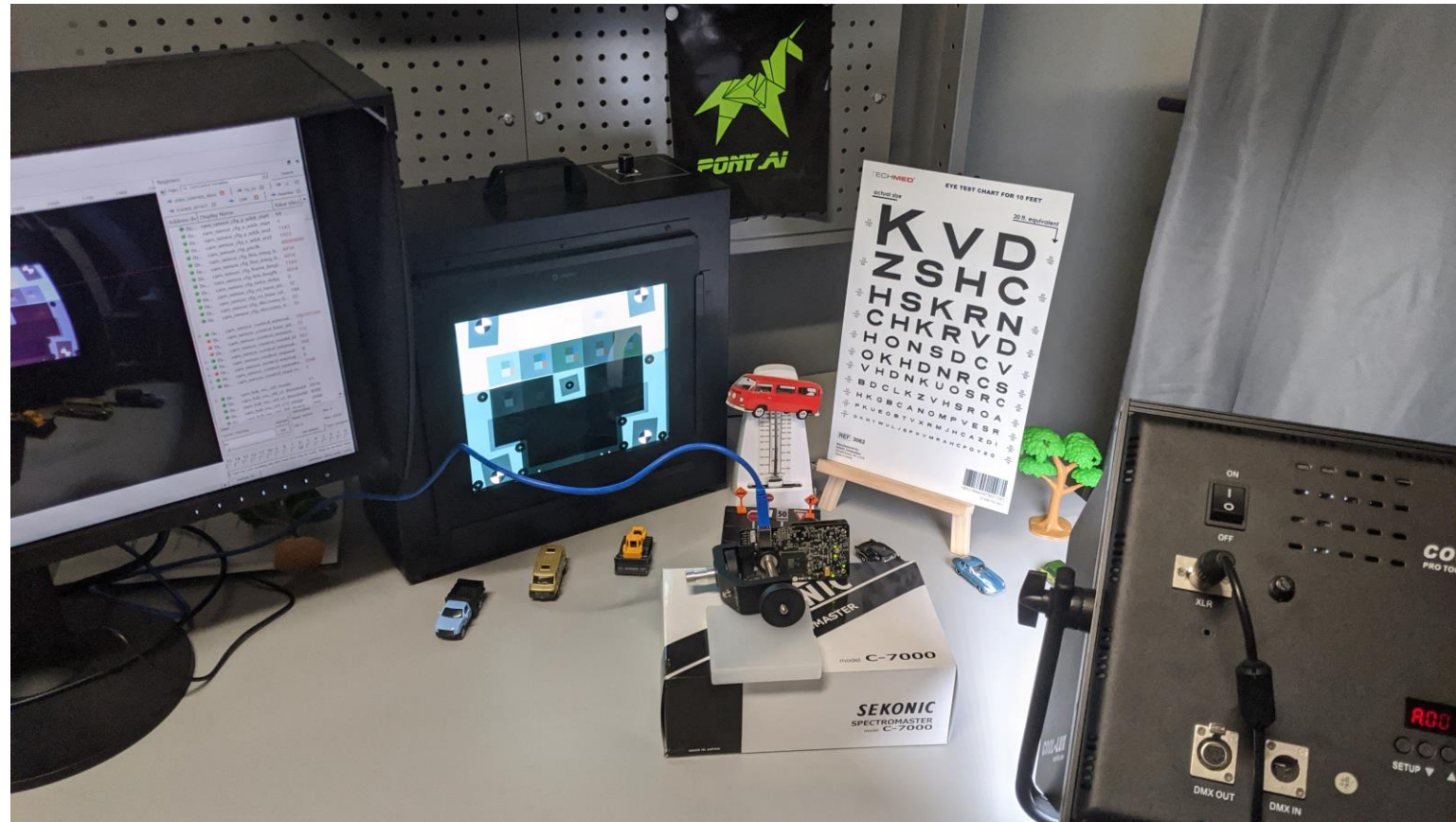
- Sensor in single exposure mode
- ISP: OnSemi AP200
- Lens: 180D ImmerVision

Light:

- Cool-lux LED (1/25/50/75 strength)
- Imatest Light-box

Scene:

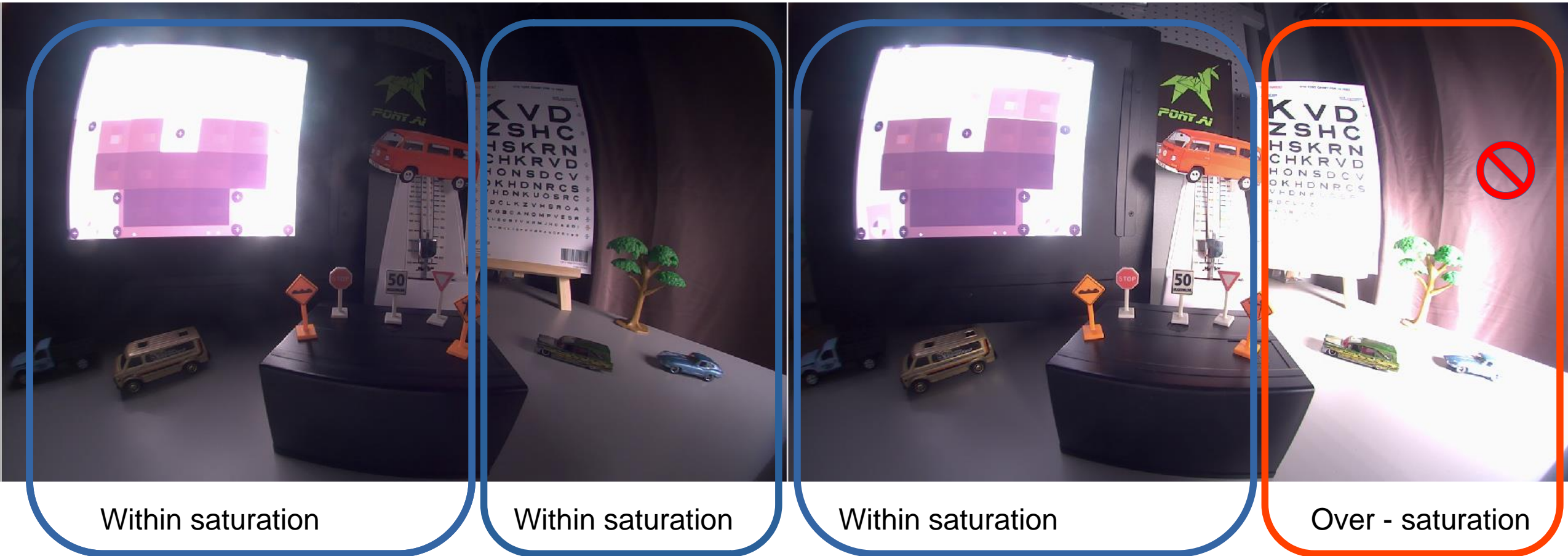
- Imatest Contrast resolution chart
- Small car models (4x)
- Human Vision test chart
- Moving object



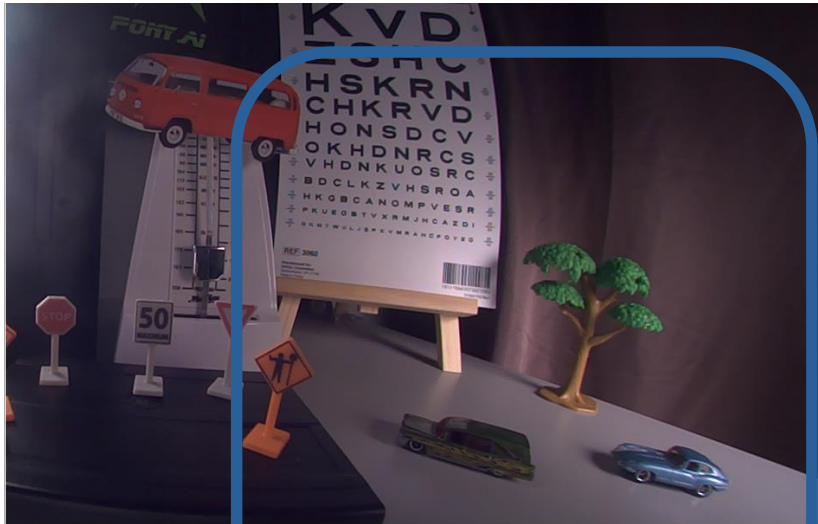
Test 1: AE + changed light

Weak light (<100lux)

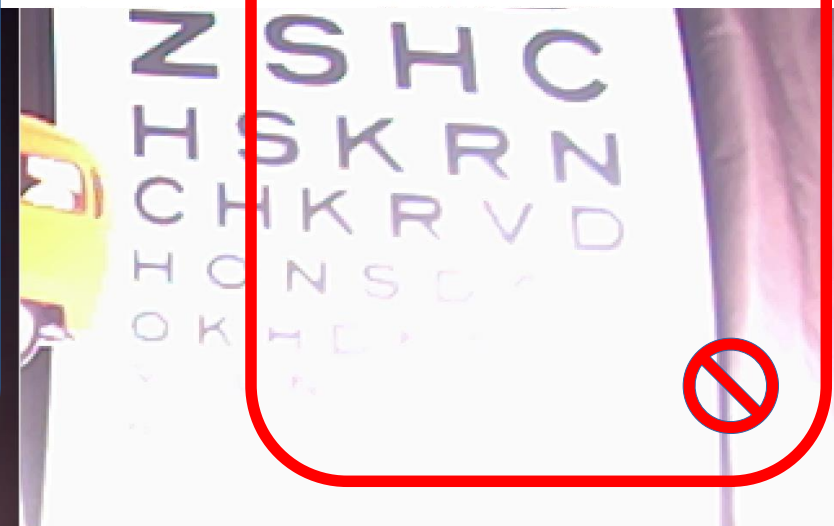
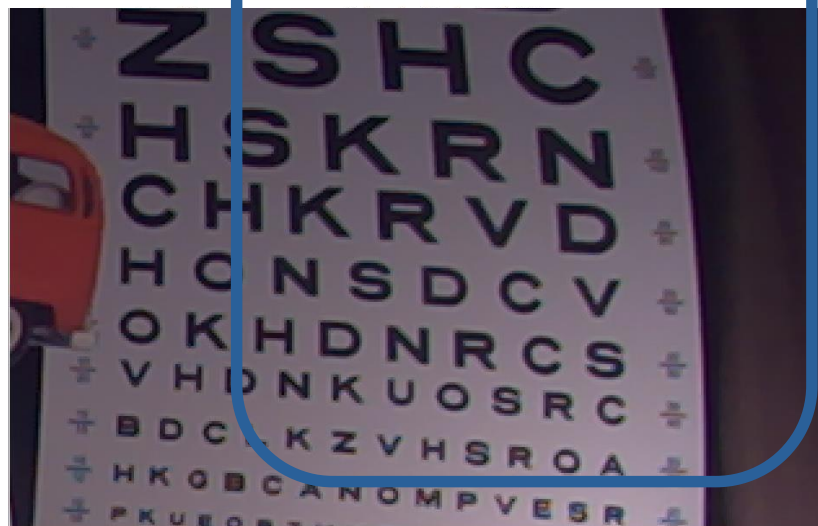
Strong light (~500lux)



Test 1: Automatic Exposure / Changed light

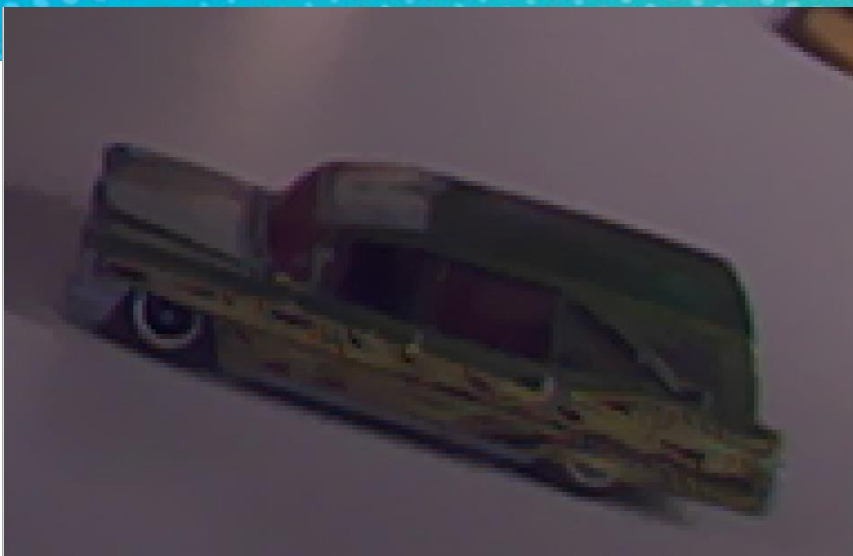


Lost visibility

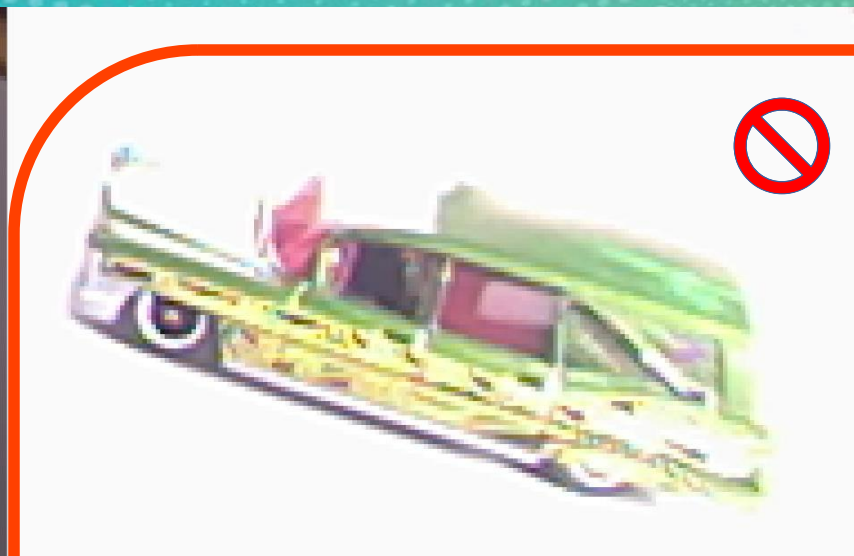


Lost visibility

Test 1: AE / Changed light



Automatic Exposure (dark light, 10)



Automatic Exposure (bright light, 250)

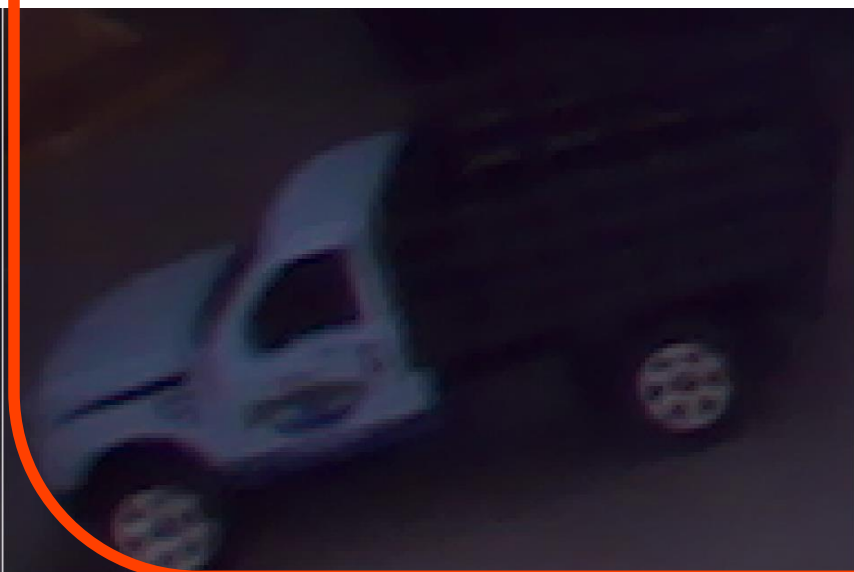
Same
Image

Can't
Be
Properly
exposed

Too dark



Automatic Exposure (dark light, 10)



Automatic Exposure (bright light, 250)

Lab Experiment 2: Manual Exposure

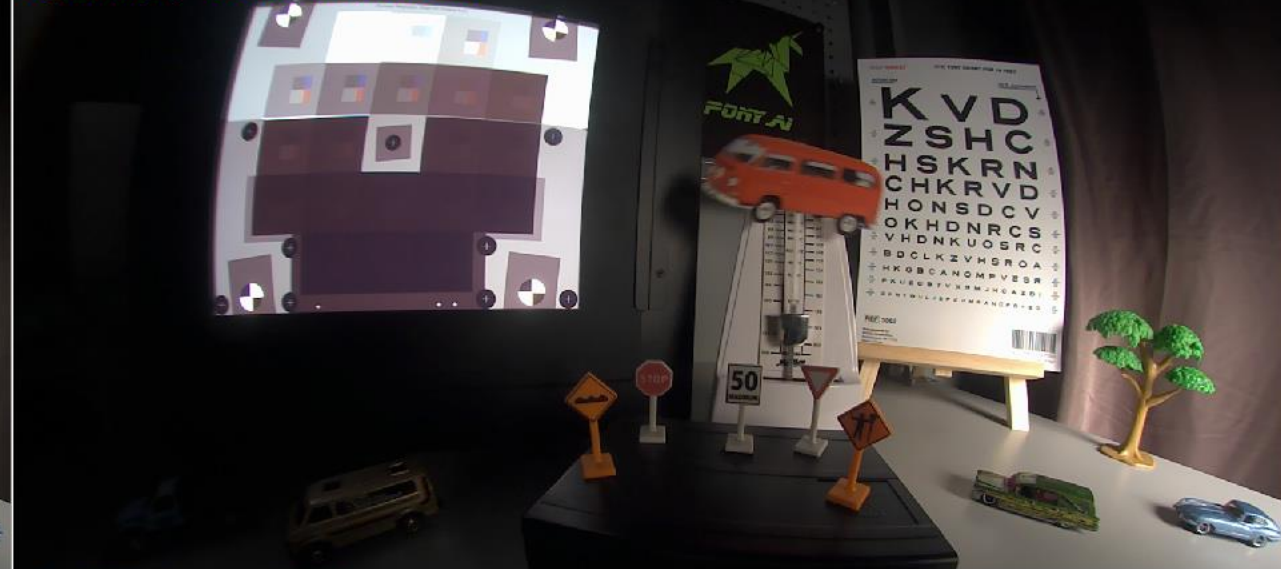
Manual Exposure : Full Scene

3_00.bmp , 1920x1080 , 5.93MB



3 MS, Light (99)

12_00.bmp , 1920x1080 , 5.93MB



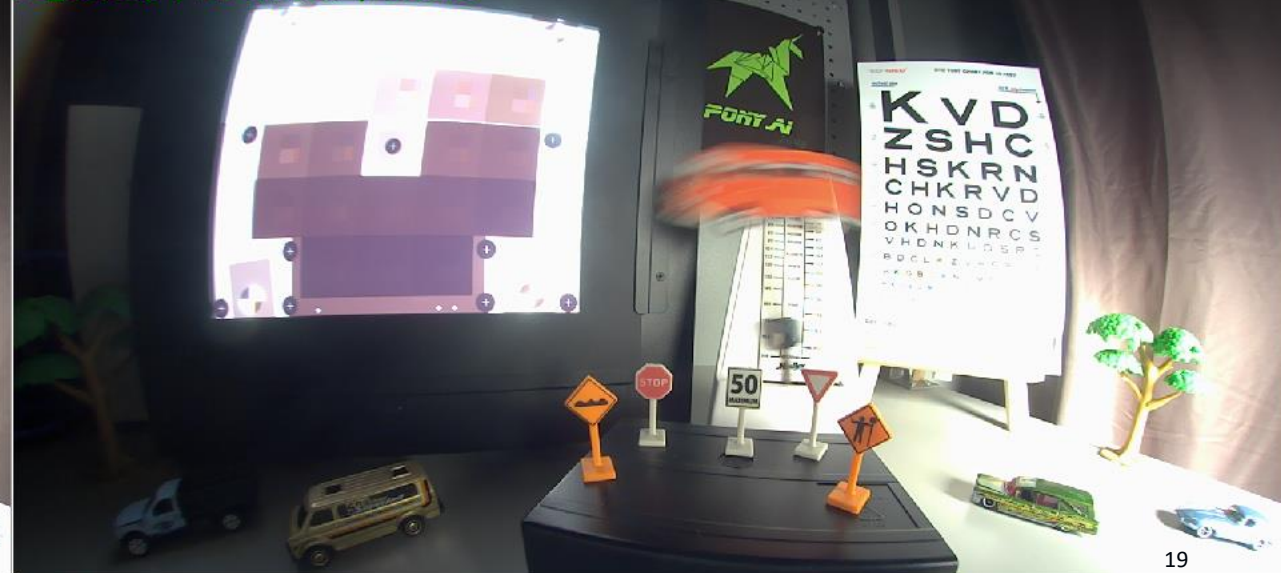
12 MS, Light (99)

36_00.bmp , 1920x1080 , 5.93MB



36 MS, Light (99)

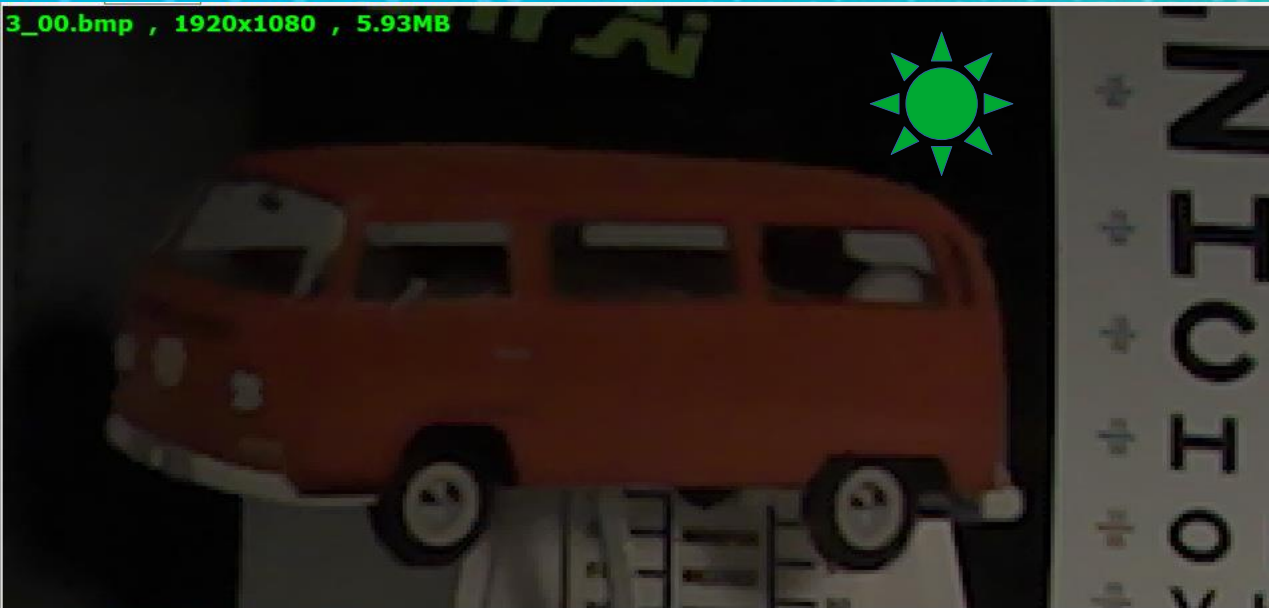
72_00.bmp , 1920x1080 , 5.93MB



72 MS, Light (99)

Manual Exposure: Moving object

3_00.bmp , 1920x1080 , 5.93MB



3 MS, Light (99)

12_00.bmp , 1920x1080 , 5.93MB



12 MS, Light (99)

36_00.bmp , 1920x1080 , 5.93MB

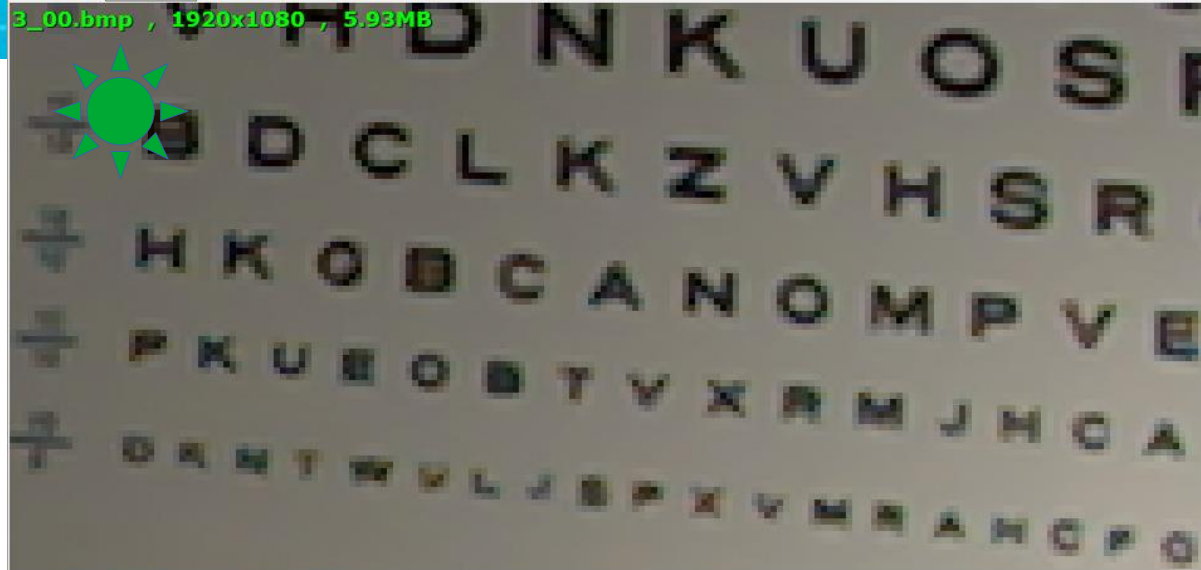


72_00.bmp , 1920x1080 , 5.93MB



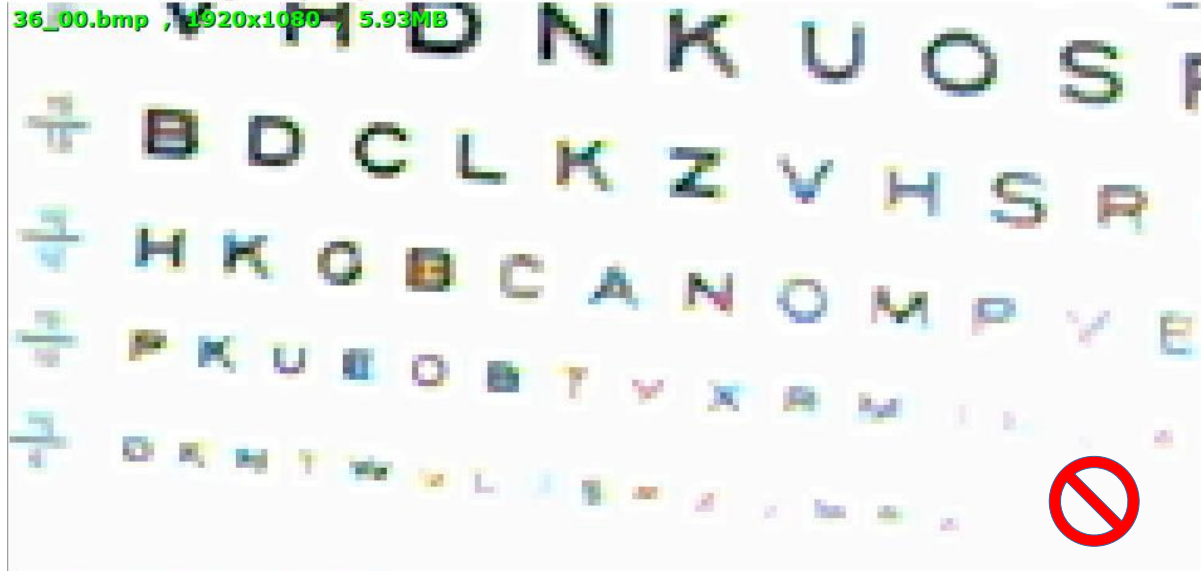
Manual Exposure: Human Vision Chart

3_00.bmp , 1920x1080 , 5.93MB



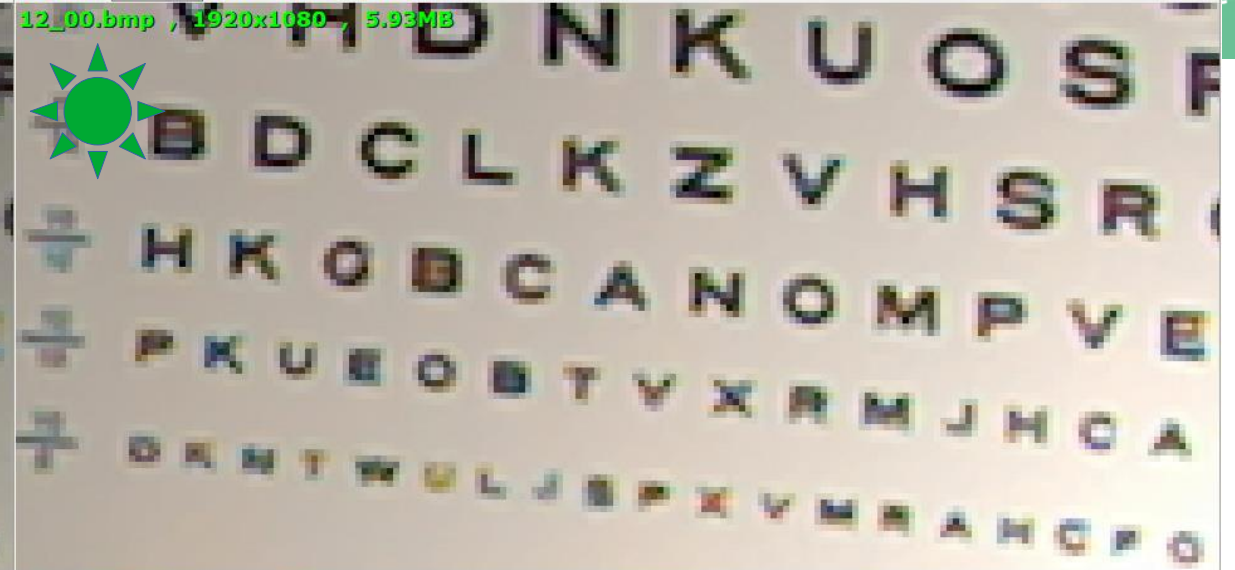
3 MS, Light (99)

36_00.bmp , 1920x1080 , 5.93MB



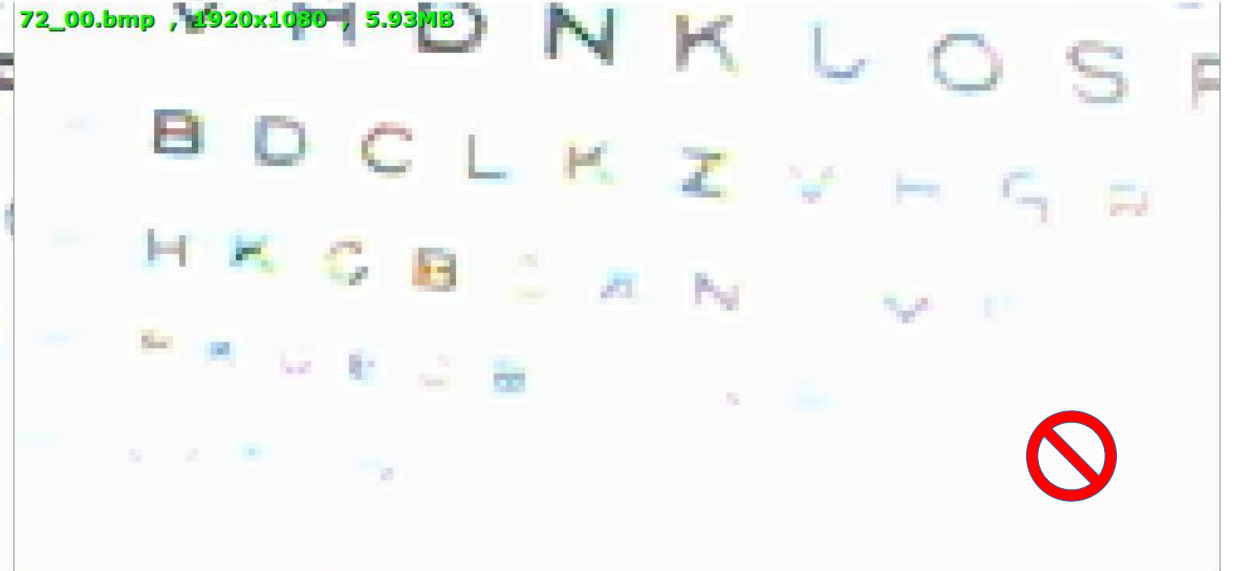
36 MS, Light (99)

12_00.bmp , 1920x1080 , 5.93MB



12 MS, Light (99)

72_00.bmp , 1920x1080 , 5.93MB

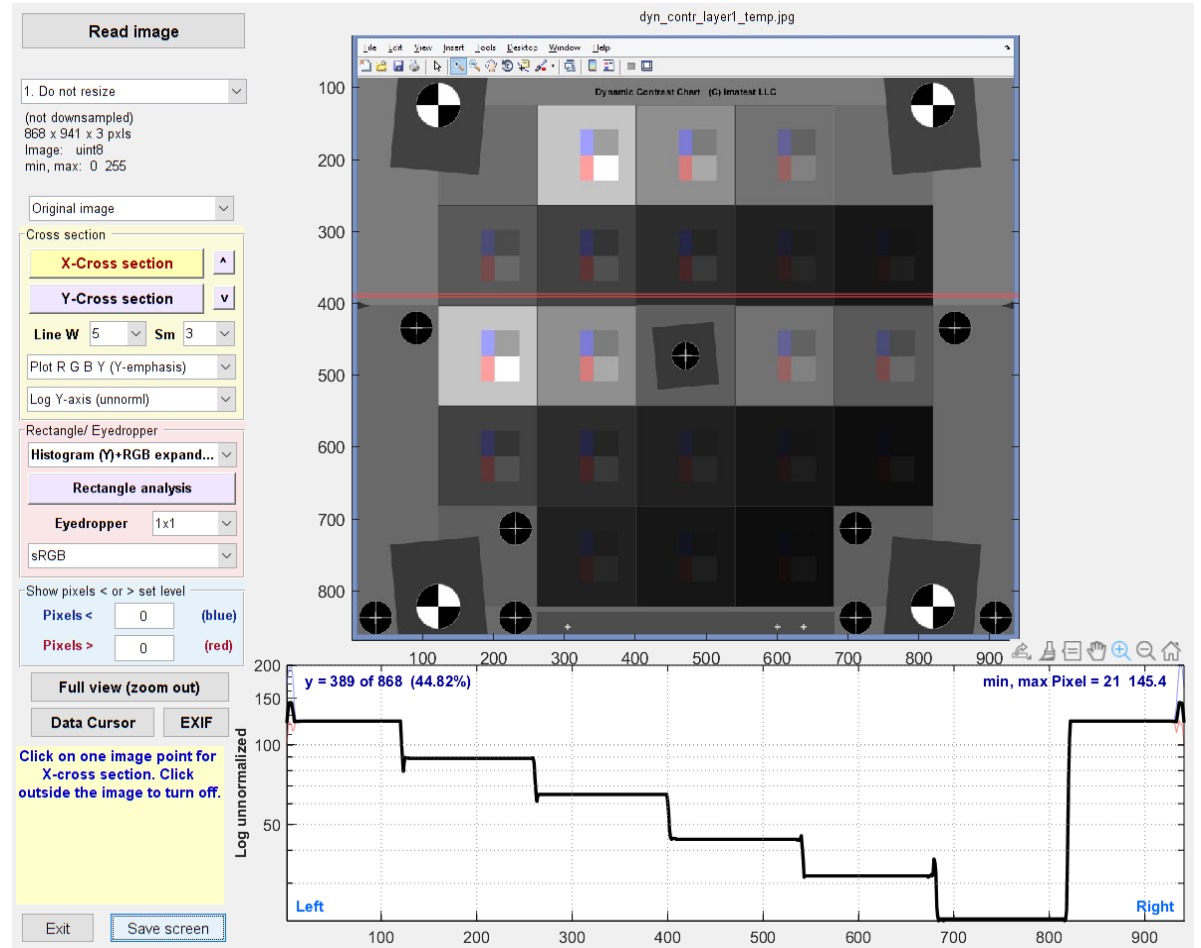
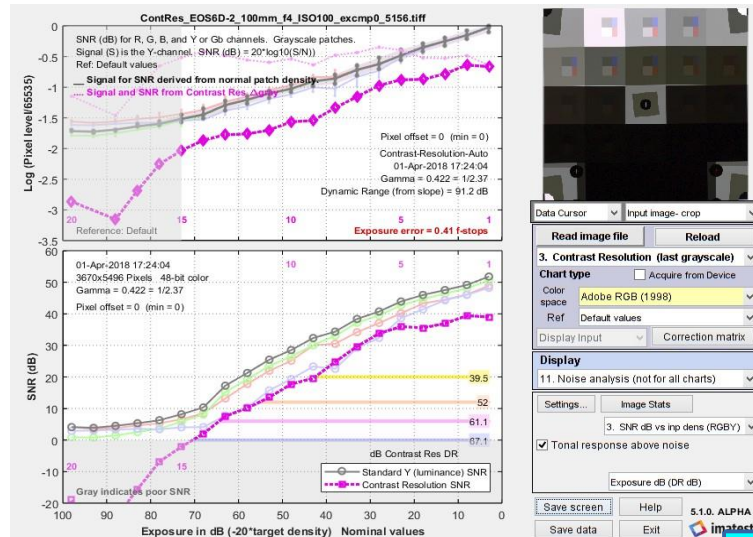
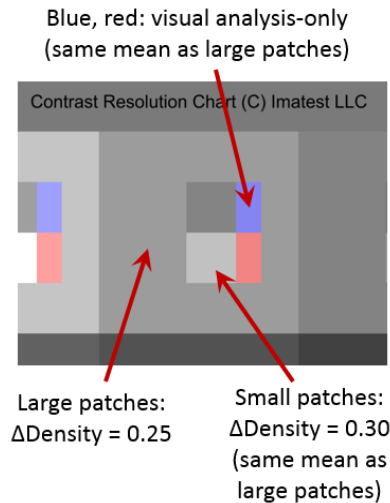


72 MS, Light (99)

Imatest Contrast Resolution

Measures the visibility of low contrast objects in larger fields over a wide range of brightness

The range of tones a camera responds to with good contrast and Signal-to-Noise Ratio (SNR)) is a function of the sensor and lens (and to some extent, the signal processing)



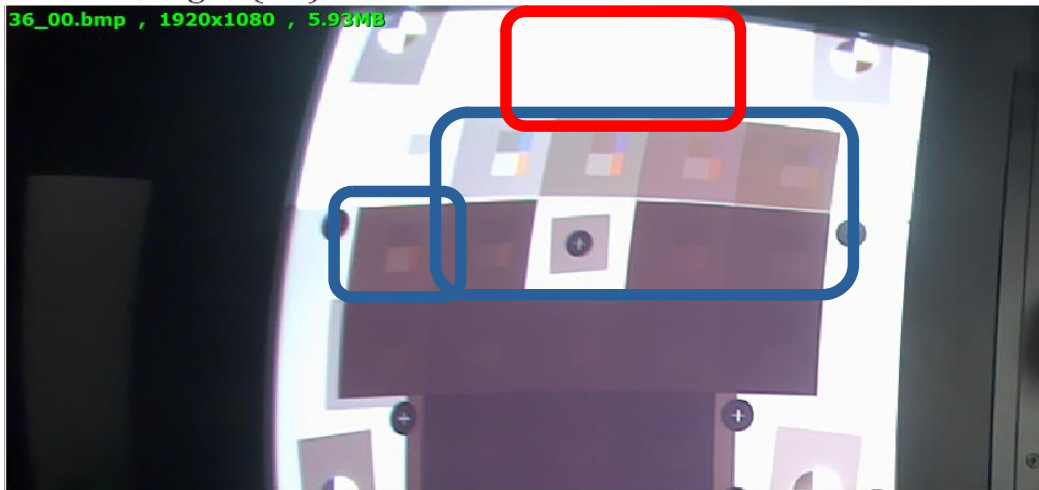
Manual Exposure: Imatest Contrast Resolution



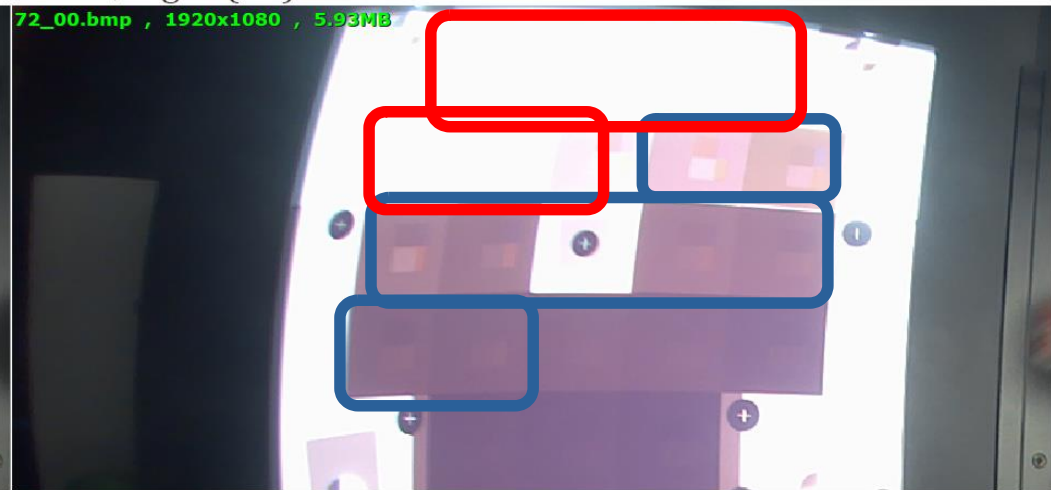
3 MS, Light (99)



12 MS, Light (99)



36 MS, Light (99)



72 MS, Light (99)

Manual Exposure: Shadows

3_00.bmp , 1920x1080 , 5.93MB



3 MS, Light (99)

12_00.bmp , 1920x1080 , 5.93MB



12 MS, Light (99)

36_00.bmp , 1920x1080 , 5.93MB



36 MS, Light (99)

72_00.bmp , 1920x1080 , 5.93MB



72 MS, Light (99)

Manual Exposure: Highlights



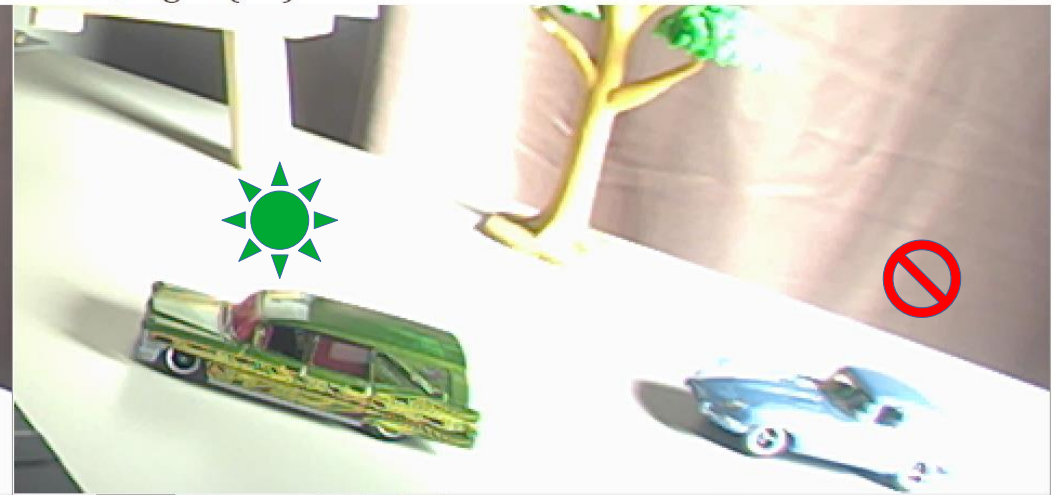
3 MS, Light (99)



12 MS, Light (99)



36 MS, Light (99)



72 MS, Light (99)

Field experiment, conclusions

Exposure time difference

mAP depends:

- ISP settings
- Training / Test set / Object sizes
- Model architecture

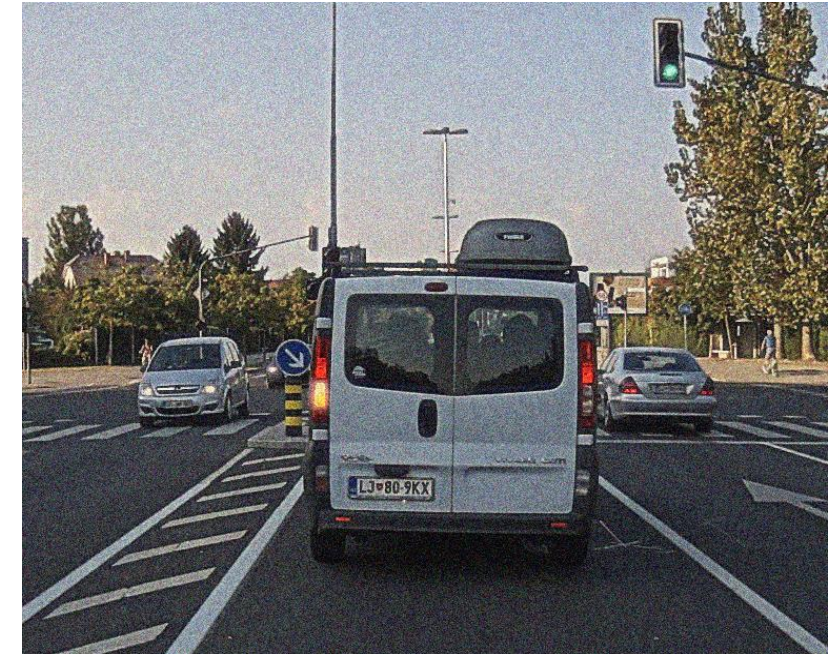
Changing one of the the above can change detection accuracy?

Common question for ISP tuning: **A or B?**

Answer: **A and B.**



A: Long exposure time: motion blur, less noise (low gain)



B: Short exposure time: sharp image, high noise (high gain)

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Conclusions

It is impossible to choose the exposure parameters which will be optimal for all objects in the scene.

Same object appears differently depending on light, exposure control and ISP parameters.

Training set never has all possible variations of lighting conditions for any object.

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Augmentation

Three ways to improve data

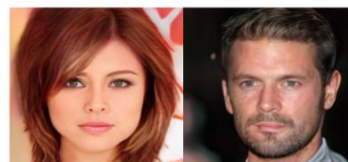
R7

1 - Collect more



- expensive
- requires manual labor

2 - Synthesize



- complicated
- might not truly represent the real data

3 - Augment



- simple
- but finding a good augmentation strategy takes lots of trial & error (**=time of AI engineers**)

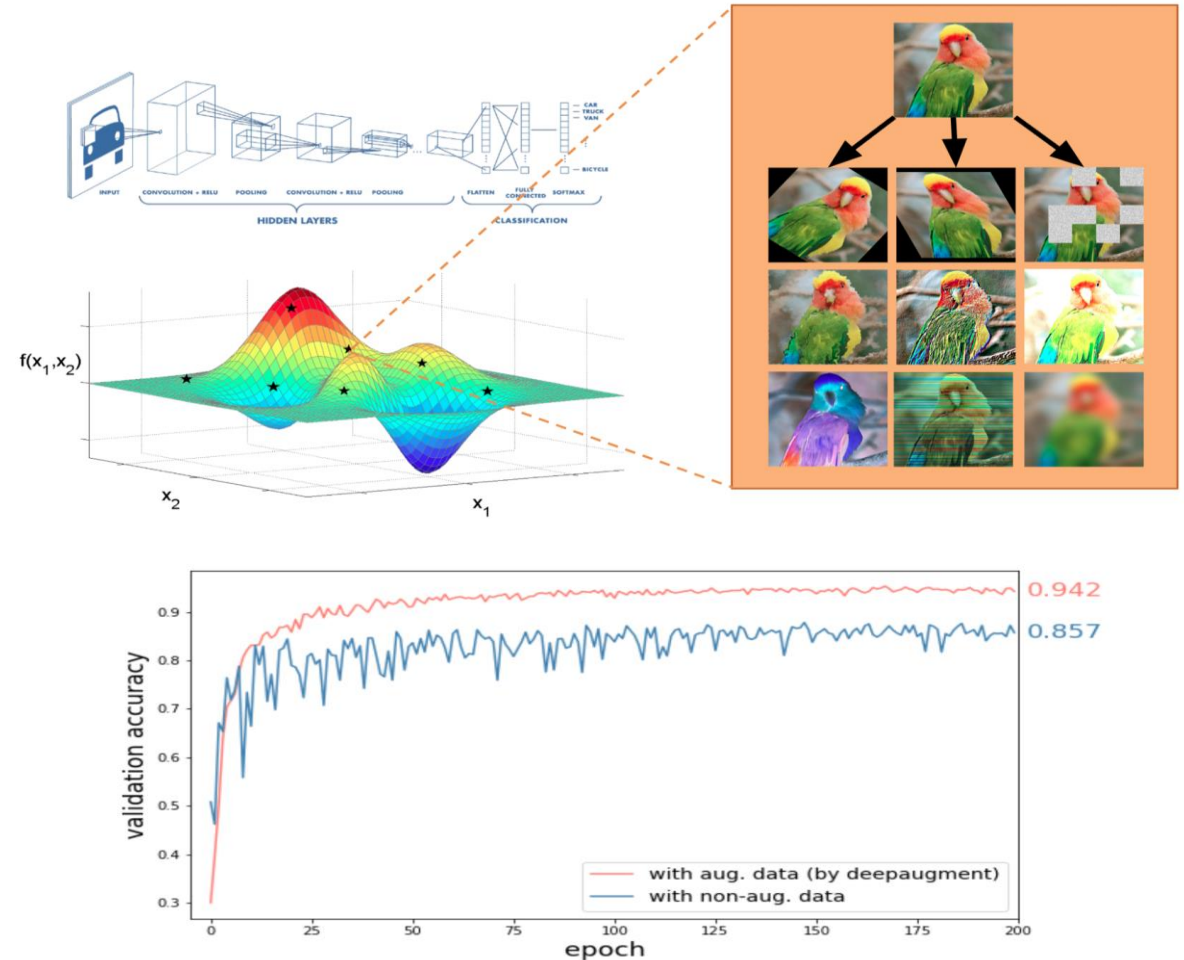


Augmentation (Auto ML)

R7

Data is the most critical piece of AI applications. Not having enough labeled data often leads to overfitting, which means the model will not be able to generalize to unseen examples.

This can be mitigated by data augmentation, which effectively increases the amount and diversity of data seen by the network. It is done by artificially producing new data by applying transformations on an original dataset such as rotation, cropping, occlusion, etc.



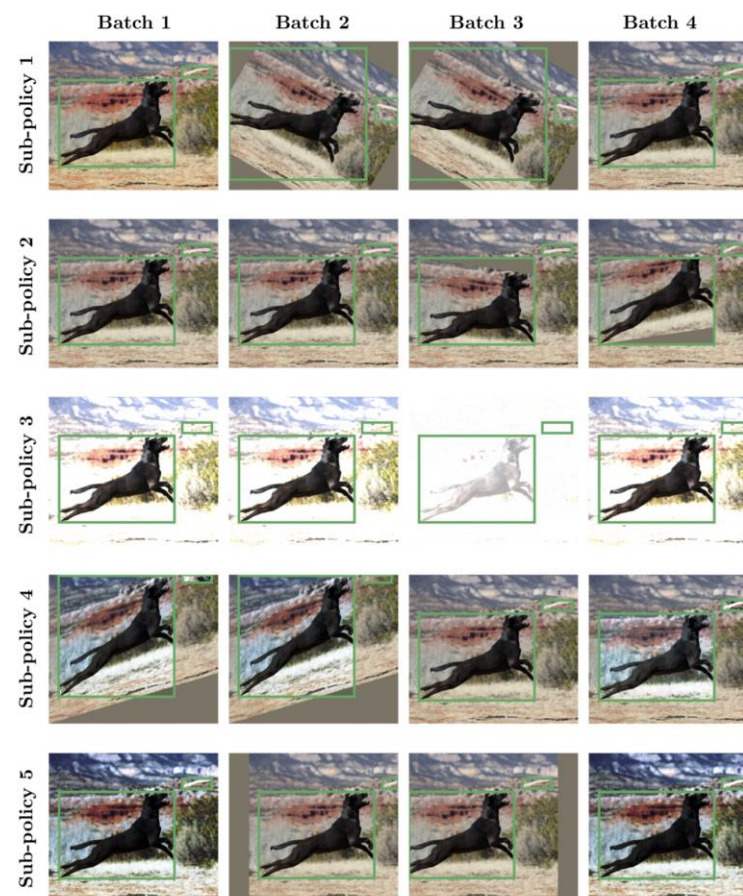
Comparison of validation accuracies of WideResNet-28-10 CNN model with CIFAR10 images when they are augmented by policies found by DeepAugment, and when they are not augmented. Validation accuracy is increased by 8.5%, equivalent to 60% reduction in error.

Learning Data Augmentation Strategies for Object Detection

Learned augmentation policy
Systematically improves object
detection

Learned data augmentation
improves model regularization

R8



Sub-policy 1. (Color, 0.2, 8), (Rotate, 0.8, 10)
Sub-policy 2. (BBBoxOnly.ShearY, 0.8, 5)
Sub-policy 3. (SolarizeAdd, 0.6, 8), (Brightness, 0.8, 10)
Sub-policy 4. (ShearY, 0.6, 10), (BBBoxOnly.Equalize, 0.6, 8)
Sub-policy 5. (Equalize, 0.6, 10), (TranslateX, 0.2, 2)

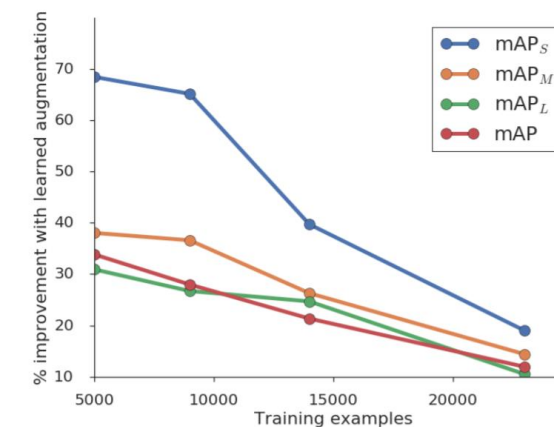
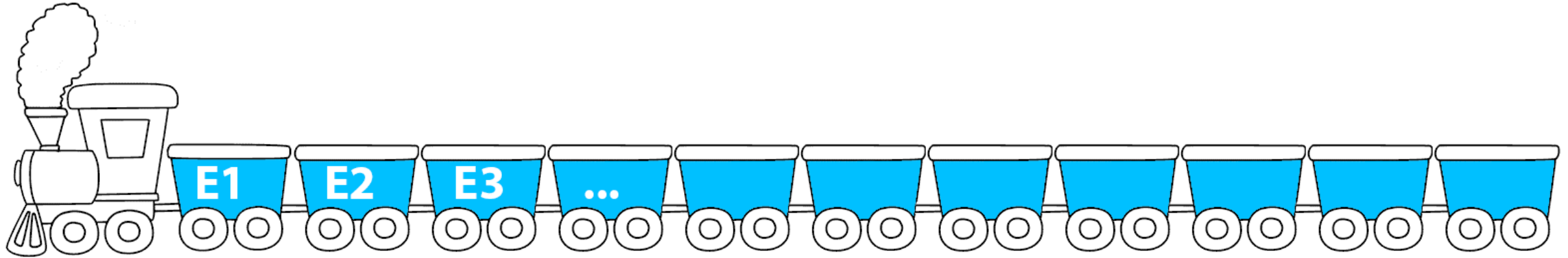


Figure 3: Percentage improvement in mAP for objects of different sizes due to the learned augmentation policy.

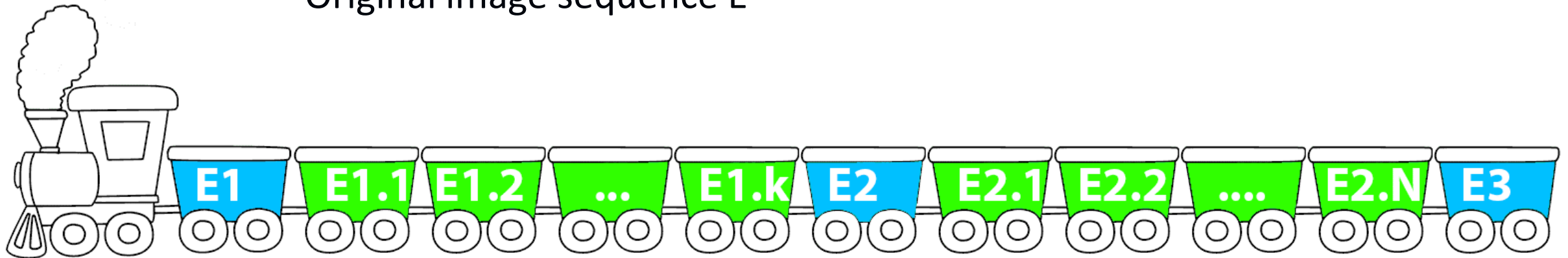
training set size	Baseline				Our results			
	mAP _S	mAP _M	mAP _L	mAP	mAP _S	mAP _M	mAP _L	mAP
5000	1.9	7.1	9.7	6.5	3.2	9.8	12.7	8.7
9000	4.3	12.3	17.6	11.8	7.1	16.8	22.3	15.1
14000	6.8	17.5	23.9	16.4	9.5	22.1	29.8	19.9
23000	10.0	24.3	33.3	22.6	11.9	27.8	36.8	25.3

Resnet-101/COCO (Small, Med, Large)

Augmented train

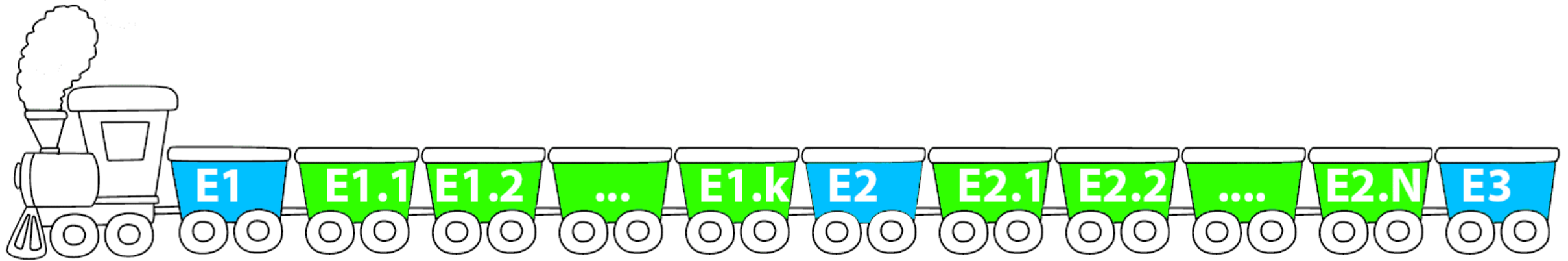


- Original image sequence E



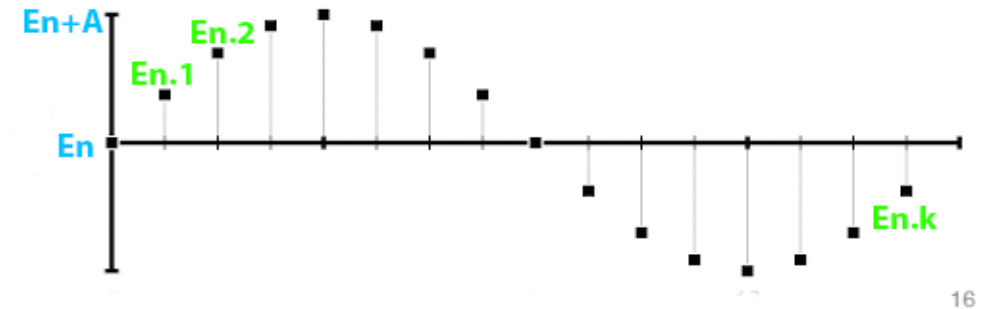
- Augmented sequence: E – anchor frames,
- k augmented frames between each two anchors

Augmented train for training networks



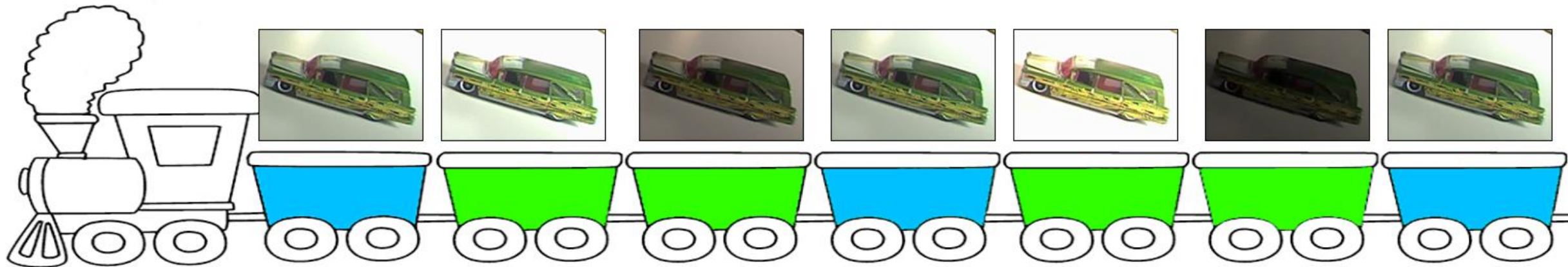
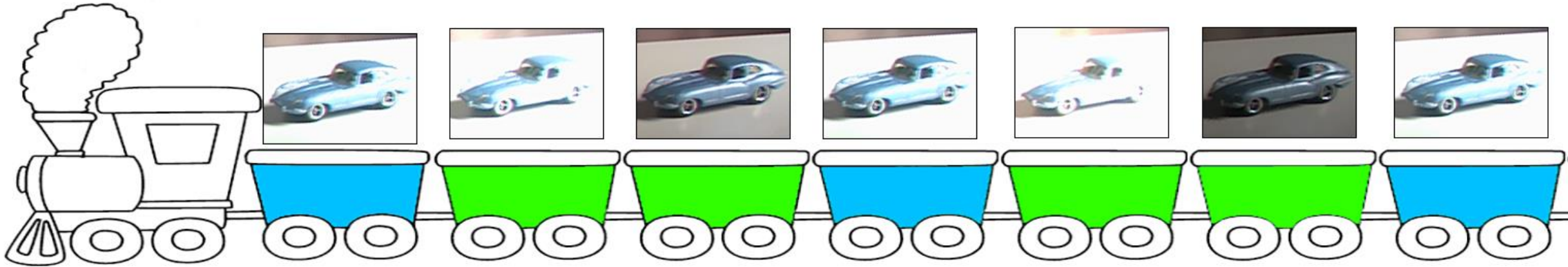
- $(E_n)'$ – exposure time of E_n
- $(E_{n.i})' = (E_n)' + A \cdot \sin(i \cdot 2\pi / (k+1))$,
- A – bracketing / augmentation strength

If $(k == 1)$, $(E_{n.i})' = \text{rand}(A) \cdot (-1)^n$



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Example of augmented exposure trains

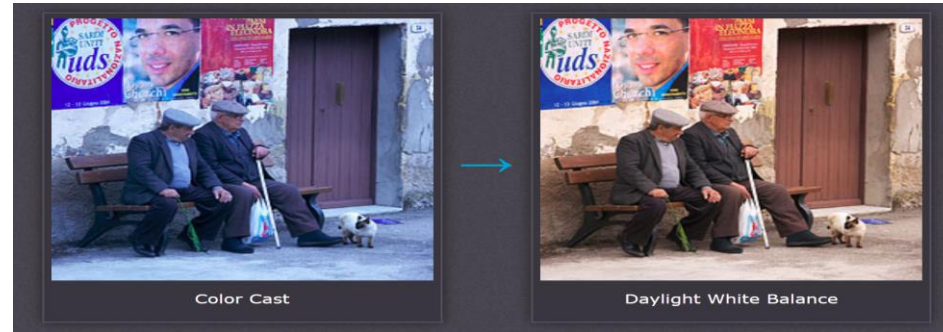


White Balance (WB) augmentation

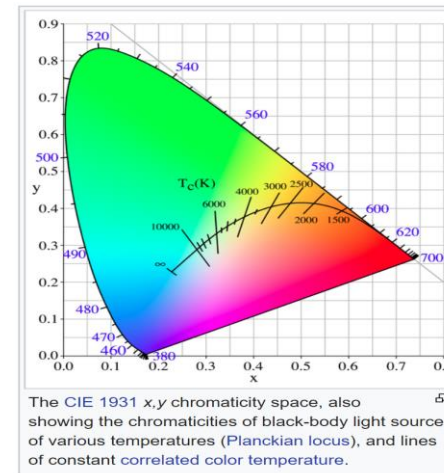
White balance (WB) is the process of removing unrealistic color casts, so that objects which appear white in person are rendered white in your photo

Proper camera white balance has to take into account the "color temperature" of a light source, which refers to the relative warmth or coolness of white light

R6



R5



Raw
Image

STAT
S

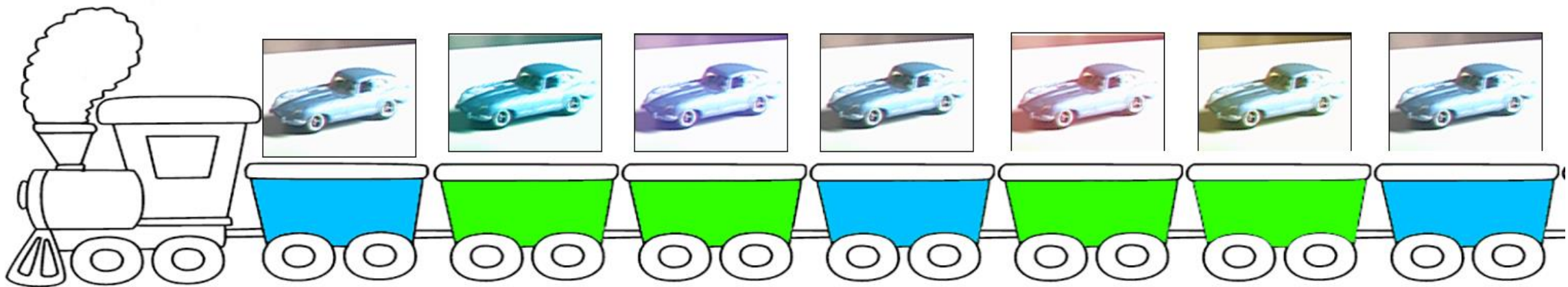
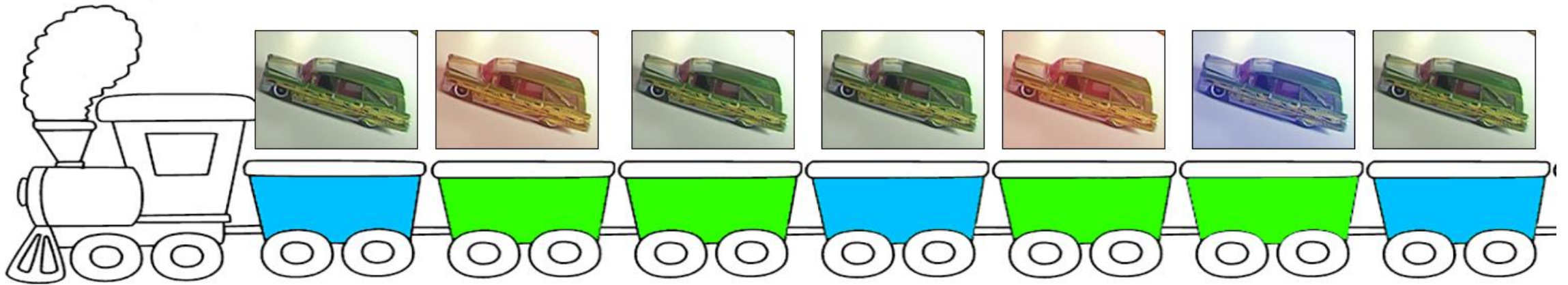
Color TEMP.
Guess

R-gain

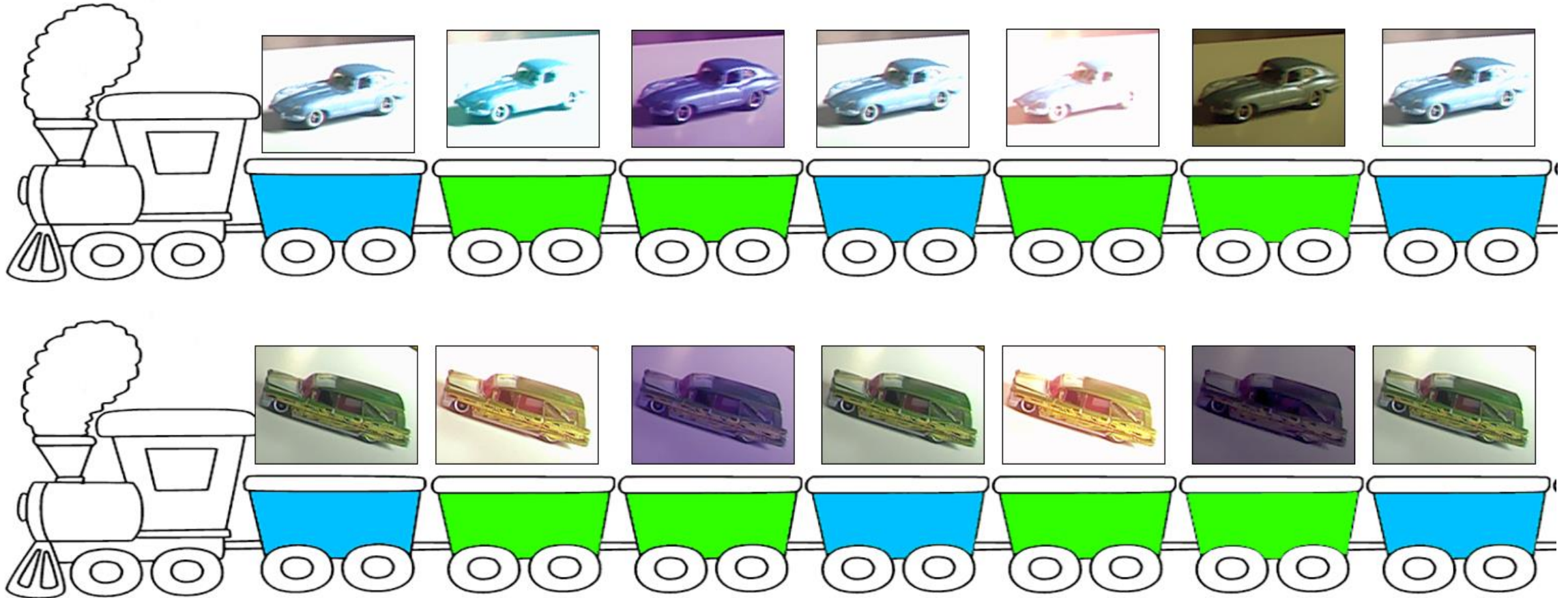
B-gain



WB augmentation (Auto-Exposure)



WB + Exposure augmentation



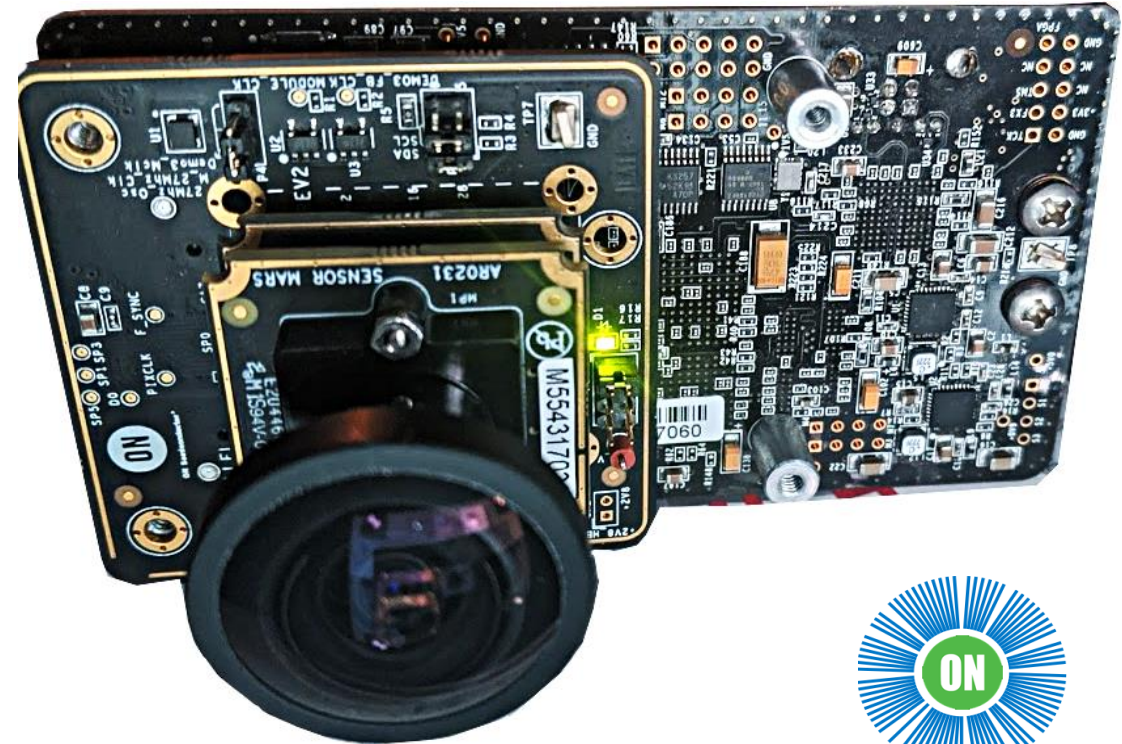
Augmented demo

Prototype:

- Single exposure mode CIS
- ISP: OnSemi AP0200

Target Platform:

OnSemi AP0300
CEVA XM-4 dsp



ISP augmentation vs. Post processing

	ISP augmentation	Post ISP Pre-processing
Baseline	Augments capturing of the object on physical level (different physical level of noise, readout, motion blur)	Only changes image data after ISP processing
Simulation ISP parameters	Physically accurated simulation of oject representation at different lighting condition	Trying to mimic contrast, gamma, etc on existing processed image.
Data	RAW 12-16bit (pre-ISP)	8 bit RGB/YUB (post ISP)
ISP parameters	Augmented based on exposure change in realistic	ISP parameters are not change/augmented
Compute	Real-time, hw-accelerated	Uses software, GPU



ISP augmentation for imaging, post processing for geometry

1. Exposure in photography (Wikipedia)

[https://en.wikipedia.org/wiki/Exposure_\(photography\)](https://en.wikipedia.org/wiki/Exposure_(photography))

2.The Exposure Triangle: Making Sense of Aperture, Shutter Speed, and ISO (Petapixel)

<https://petapixel.com/2017/03/25/exposure-triangle-making-sense-aperture-shutter-speed-iso>

3. Boosting sensitivity (Vision research, Phantom):

<https://phantomhighspeed-service.force.com/servlet/servlet.FileDownload?file=00P1N00000dSuGDUA0>

4. Contrast Resolution chart and analysis (Imatest)

<https://www.imatest.com/docs/contrast-resolution/>

5. Color Temperature (Wikipedia)

https://en.wikipedia.org/wiki/Color_temperature

6. Understanding White Balance (Cambridgecolor)

<https://www.cambridgeincolour.com/tutorials/white-balance.htm>

7. AutoML for Data Augmentation (Baris Ozmen)

<https://blog.insightdatascience.com/automl-for-data-augmentation-e87cf692c366>

8. Learning Data Augmentation Strategies for Object Detection (Google Research):

<https://arxiv.org/abs/1906.11172>

Thank you!