

Image Signal Processing Optimization for Object Detection

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NEXTCHIP Overview



World-class ISP In-house Core

- Developing & optimizing vision core; Image signal processing technology for 27 years
- Tuning know-how with various MP models with global OEMs and Tiers
- Tuning capability for human vision & machine mision
- Open architecture with various image sensors, CFAs (color filter arrays)

ASIC Design Technology

- Automotive process foundry experience;
- 14nm/28nm/55nm/60nm/95nm
- Samsung/Global Foundries/USJC/TSMC

Automotive Reliability

- ISO26262; Functional safety
- Cyber security
- CMMI Lv.-3
- A-Spice process
- AEC-Q100 Gr.2 lineup

Image Signal Processing Optimization for Object Detection

Chapter 1: What is the Difference? Human Vision vs. Machine Vision

Chapter 2: The Image Tuning Challenges for Human Vision

Chapter 3: The Image Tuning Challenges for Machine Vision

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Human Vision vs. Machine Vision



• We asked this question to ChatGPT... It gave this image as an answer!



Do you feel the same way?

Image Tuning Needed for Both Types of Vision





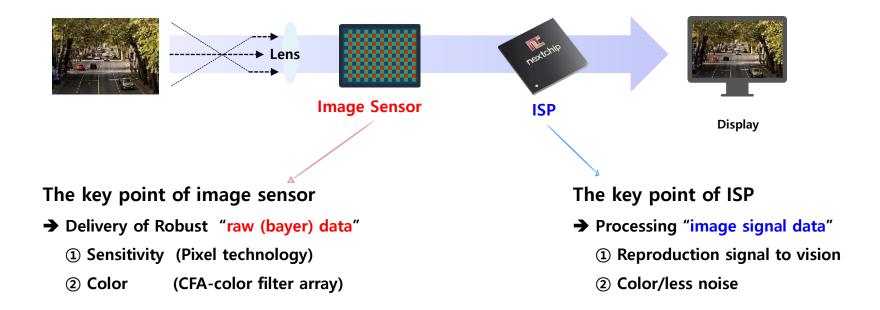


Image Signal Processing Optimization for Object Detection

Chapter 1: What is the Difference? Human Vision vs. Machine Vision

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What is the challenge?

Make the image as similar as possible to one seen through a human eye

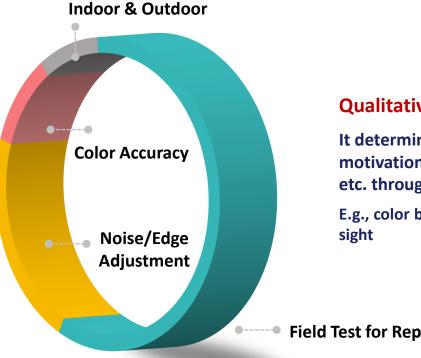
What is the key factor to tune for human vision?

- Color reproduction
- Lower noise level
- Brightness/edge/HDR (high dynamic range), etc.

Tuning under various environment, e.g., day & night







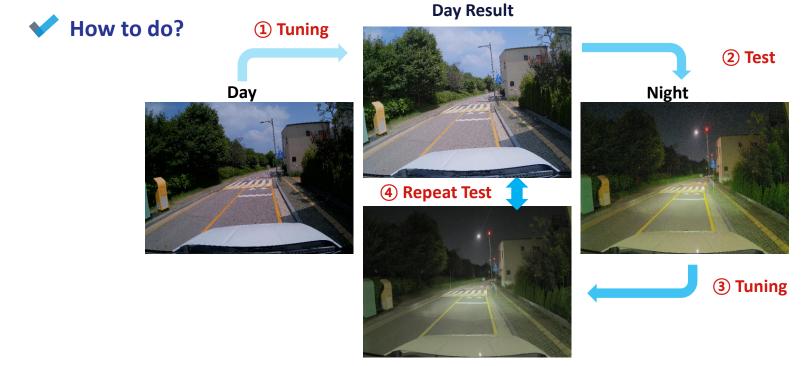
Qualitative TEST

It determines the user's motivation, comments, feeling, etc. throughout the test process.

E.g., color balance and bright in a

Field Test for Repeat





Night Result

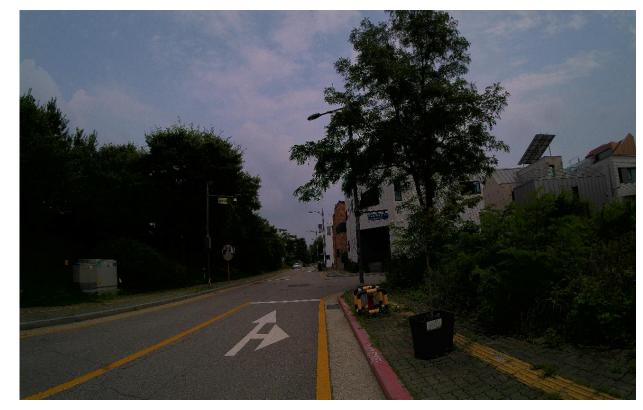


DAY

• Generally dark

Problem

- Too strong color
- Too strong edge level





DAY

- Tuning#1
- Brightness
- HDR & Contrast
- GCE
- (global contrast enhancement)





DAY

- Tuning#2
- Color (hue, saturation)
- Color suppression











Final tuned image

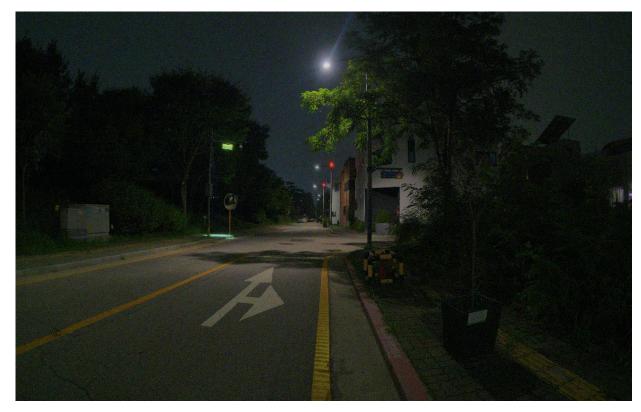






NIGHT Problem

- Generally dark
- Too strong color
- Too strong edge level





NIGHT

► Tuning#1

- Brightness
- HDR & Contrast
- GCE

(global contrast enhancement)



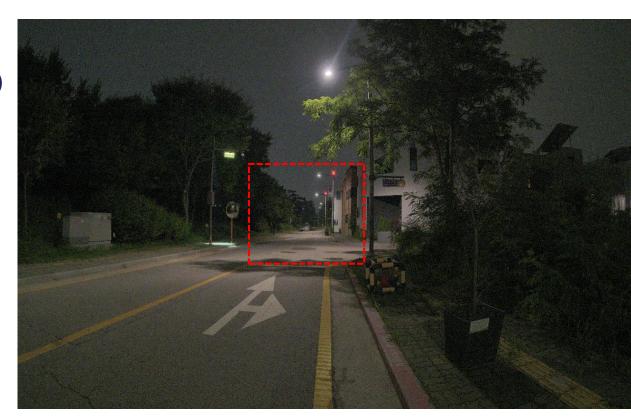


NIGHT Tuning#2

- Color (hue, saturation)
- Color suppress











Final tuned image





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Vhat is the challenge?

• Higher detection rate is needed

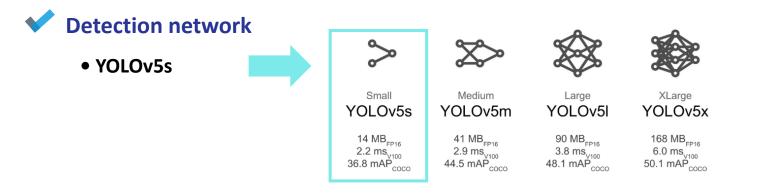
Methods to increase detection rate such as:

- Retraining
- Changing training method
- Changing field of view and resolution
- Image tuning, etc.



Measure Factors for Test







- Location : Pangyo, Korea
- Scene : Sunny, daytime & nighttime, rearview fisheye 190°
- Training image resolution : 640x360 / training images : 12,732

Test Dataset & Tuning





- Quantitative experiments: Stationary object + Ground Truth
- Qualitative experiments: Driving scene

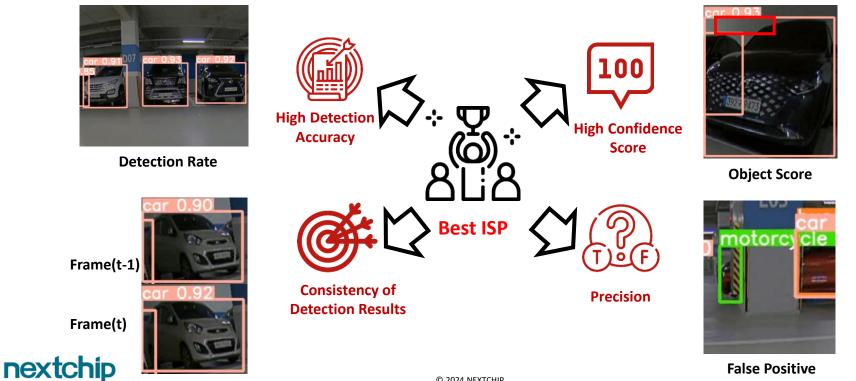


- Brightness level: Auto exposure (AE)
- Edge sharpness level: Edge enhancement (EDGE)
- Noise level: Noise reduction (NR)

Quantitative Experiments – Metric



Metrics of best ISP for object detection



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Quantitative Experiments – ISP Tuning & Test Dataset

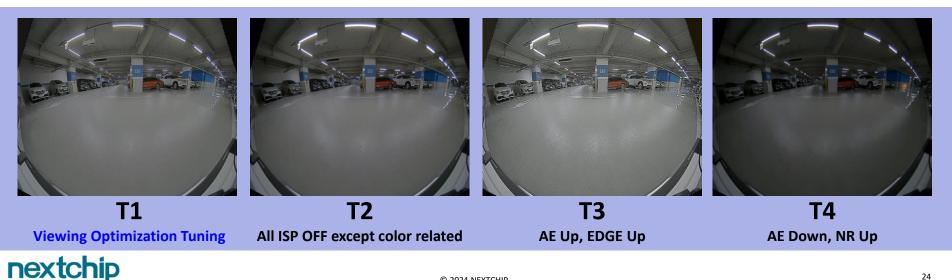




- 4 different ISP settings for the same scene
- About 3200 frames for each tuning point

ISP tuning

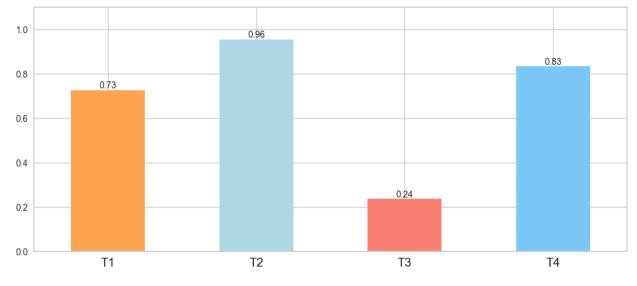
- Brightness level: Auto exposure (AE)
- Edge sharpness level: Edge enhancement (EDGE)
- Noise level: Noise reduction (NR)



Quantitative Experiments – Detection Accuracy



• An indicator of recognition accuracy for each tuning point



accuracy for each Tuning Point

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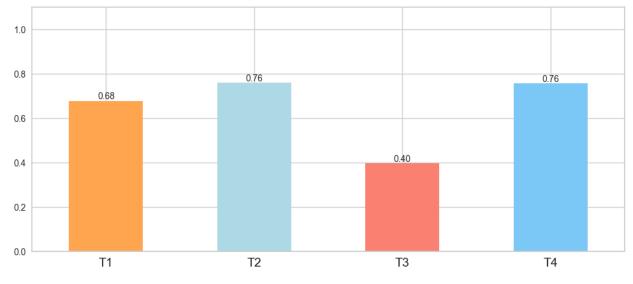
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Quantitative Experiments – Confidence Score



✓ High confidence score

• A score which represents likelihood that the bounding box contains an object



confidence for each Tuning Point

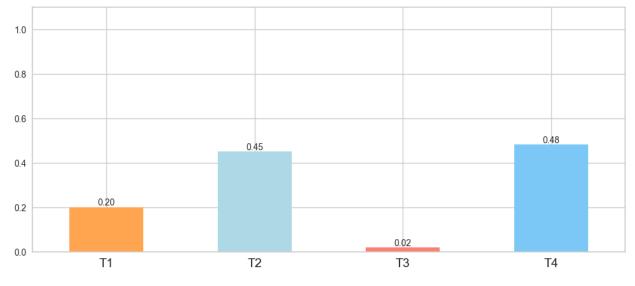
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Quantitative Experiments – Detection Consistency



• An indicator of whether the same object is consistently recognized



consistency for each Tuning Point

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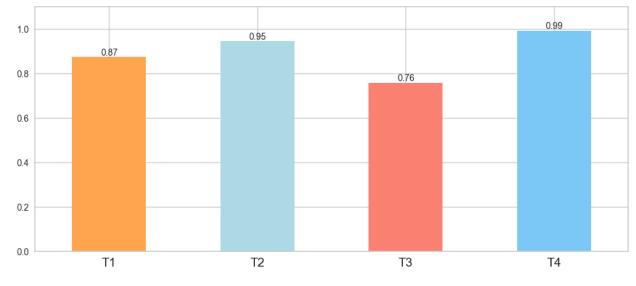
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Quantitative Experiments – Precision



Precision

• An indicator of recognition precision



precision for each Tuning Point

Quantitative Experiments – Result





• For all metrics, the higher the better



total_score for each Tuning Point

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Quantitative Experiments - Conclusion



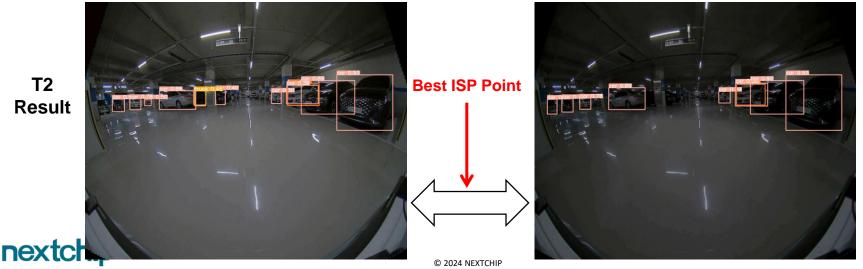
T4

Result

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- EDGE has the greatest impact on detection performance
 - 1. Too many EDGE worse detection performance
 - 2. More EDGE

- More false detections
- Darker image Reduced false detection rate and accuracy
- Need to fine the best ISP setting value between T2 and T4



Qualitative Experiments – Evaluation Methods



- Estimate the false detection rate
- Counting false positives (FP) for period in which false detection occurs in all tuning points

T1 : Original Setting



T5 : Edge Sharpness Off + Bright Up



T6 : Edge Sharpness Off + Bright Up + NR Up

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T7 : Edge Sharpness Off + Bright Down



T8 : Edge Sharpness Off + Bright Down + NR Up



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Qualitative Experiments – Best ISP for Object Detection

Daytime test

• Additional 5 ISP settings for the same driving path

T1 : Original Setting

T5 : Edge Sharpness Off + Bright Up

T8 : Edge Sharpness Off + Bright Down + NR Up







T7 : Edge Sharpness Off + Bright Down





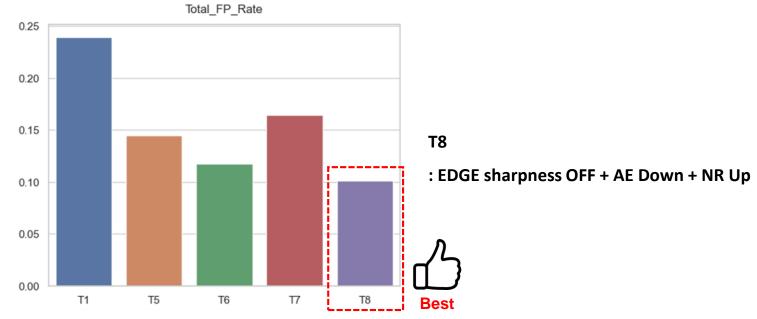


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Qualitative Experiments – Best ISP for Object Detection





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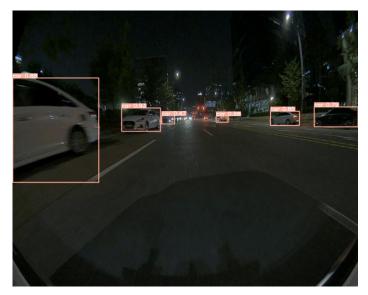
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Qualitative Experiments – Best ISP for Object Detection

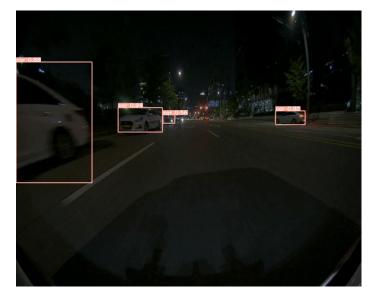


Nighttime test
2 ISP settings are applied for same driving path



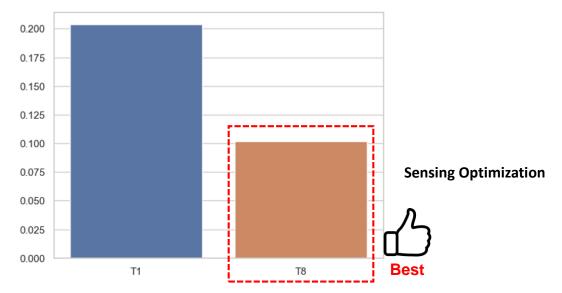
T1 : Viewing Optimization Tuning

T8: Edge Sharpness Off + AE Down + NR Up



Qualitative Experiments – Best ISP for Object Detection





False Detection per Image

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Quantitative Experiments – Conclusion

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- Qualitatively, the detection rates are similar at all tuning points
- Datasets1 (Day time)
 - **1.** When noise level is high, reduces false detection rate
 - 2. In daytime, brightness does not seem to have a significant effect on false detection
- Datasets2 (Night time)
 - 1. T8 (Sensing) false detection rate is 0.1 better than T1 (viewing tuning)
 - 2. At nighttime, when brightness level is low, reduced false detection rate

Future Works



The problem with current experiments

• Since the performance is evaluated only for specific points,

there are some limitations to estimate the tendency value for each tuning factor.

Further experiments

• We keep working to analyze the trends while changing the AE (brightness), EDGE, and the noise level in optimal ISP tuning.

Resources

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ChatGPT <u>https://chatgpt.com/n</u>

• Test by Nextchip Internal Standard of Image Quantitative & Qualitative Test

2024 Embedded Vision Summit

- Booth#109
- Mr. Young-Jun Yoo
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