



# Selecting Image Sensors for Embedded Vision Applications: Three Case Studies

Monica Houston

Manager, AI / ML Team

Avnet

 AVNET®

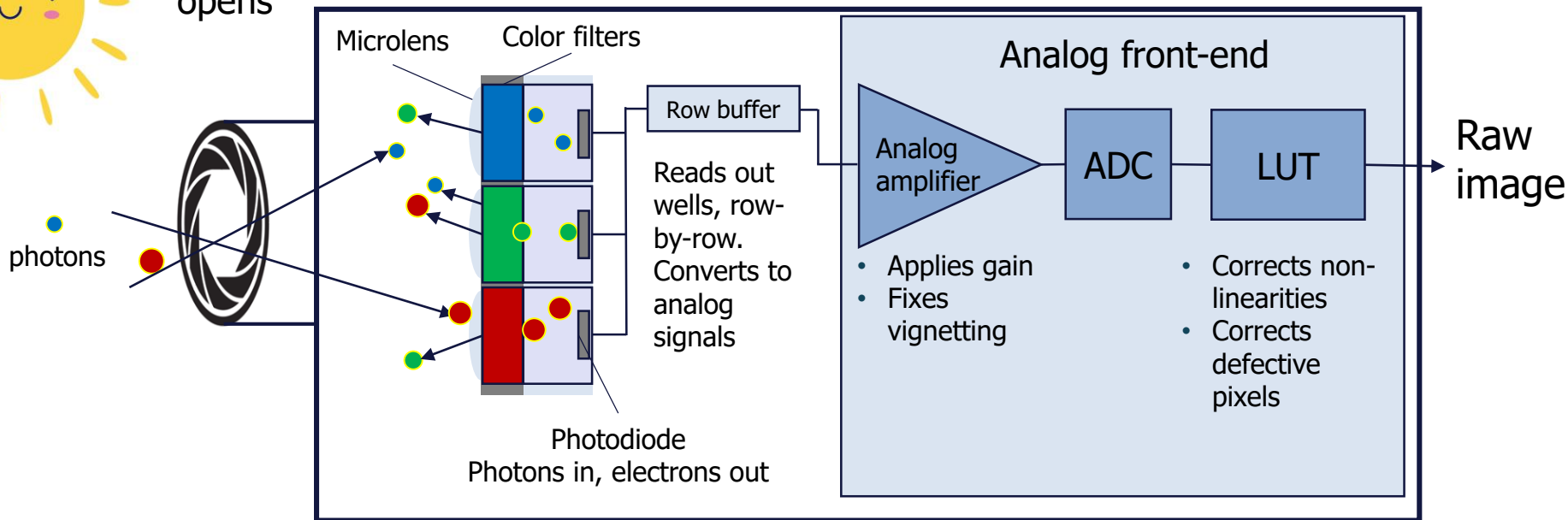
# What does an image sensor do?



1. Camera shutter opens

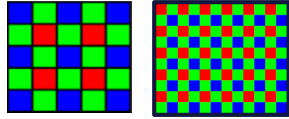
2. Photon wells ("pixels") store photons

3. Analog front-end converts analog to digital

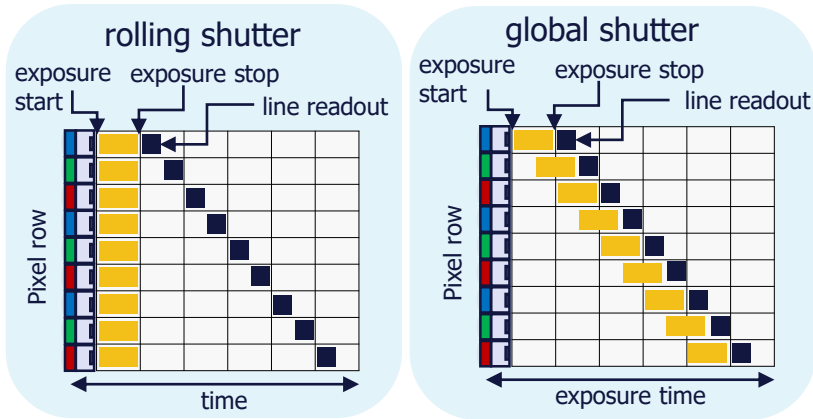


# Image sensor specifications

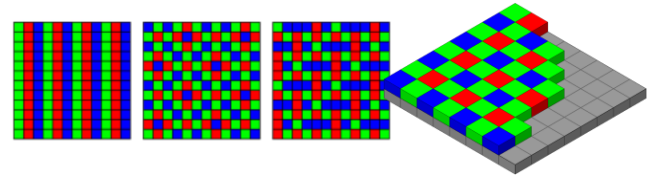
Resolution



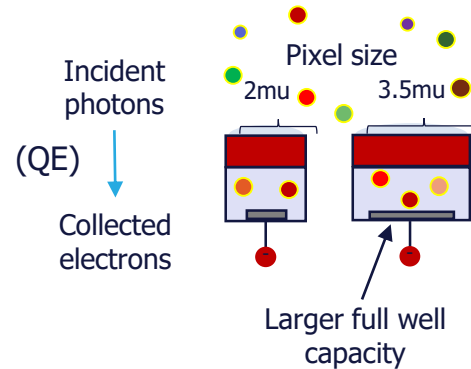
Shutter type and speed



Color Filter Arrays (CFAs)



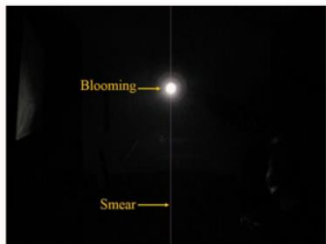
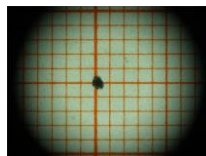
Signal to Noise Ratio (SNR)  
& Dynamic Range



## Other specs:

- Near Infrared Optimized (NIR)
- Chief Ray Angle (CRA)
- Field of View (FOV)
- Defective Pixels
- Pixel size / full well

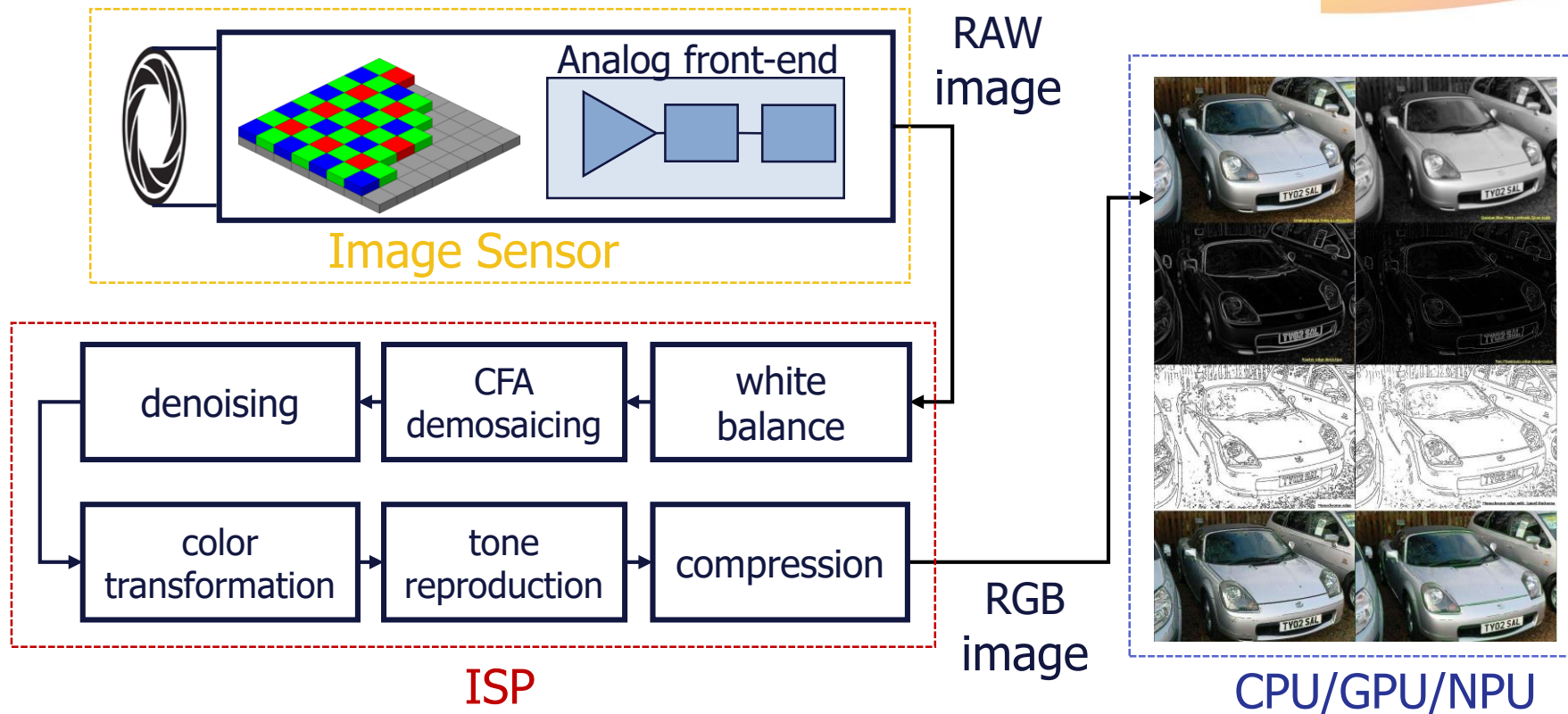
vignetting



## Other things to consider:

- Image Signal Processor (ISP)
- Interfaces
- Compression
- Sensor size
- Power consumption (heat)
- Lens
- Price
- Lead time
- Support

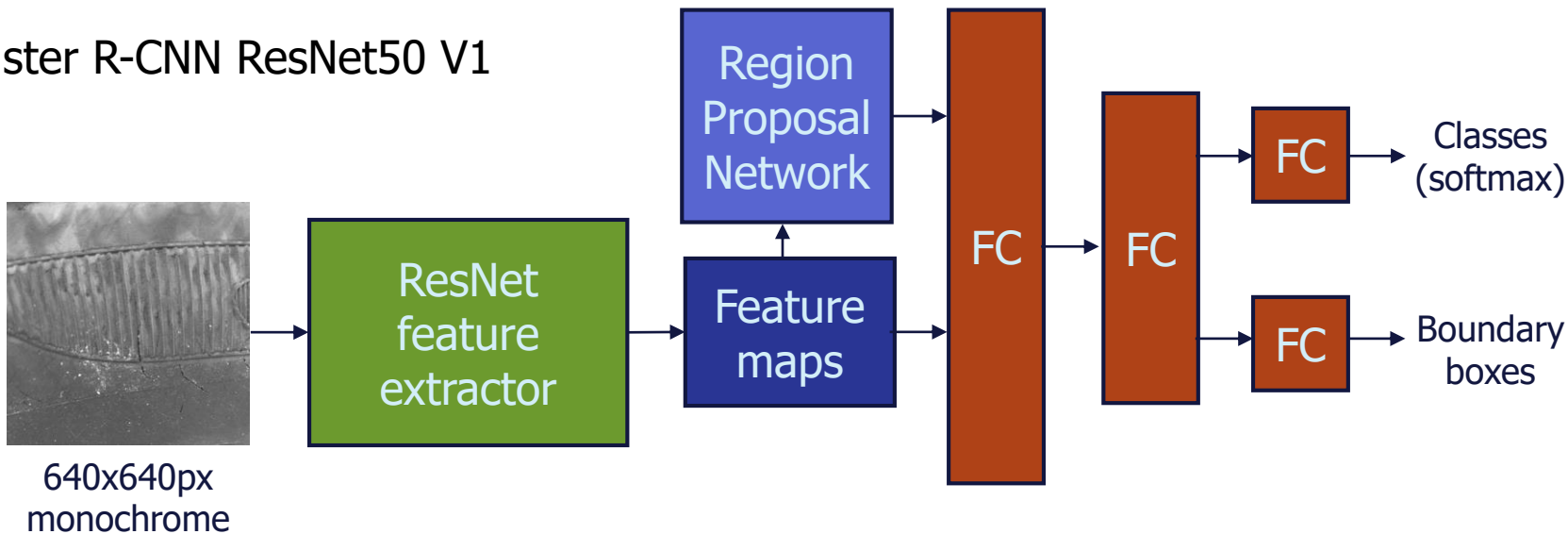
# Why choosing the right image sensor is important



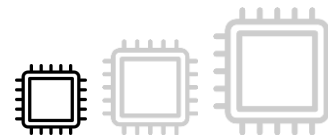
# **Case Study #1: Tire Defect Detection**

# The model

## Faster R-CNN ResNet50 V1



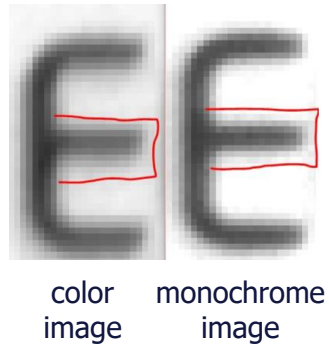
Computational  
requirement:



# Monochrome sensors vs color sensors

## Monochrome pros:

- Higher quantum efficiency
- Better features:
  - Reduced noise
  - Improved contrast
- Increased speed
- Lower cost

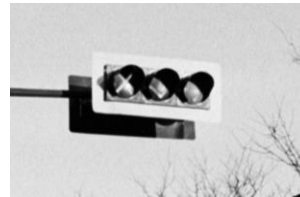


Can you just use a color image sensor and convert to greyscale?

Some models do better with color



oxidation detection



navigation



clock or stove?



tiger or snow leopard?



# Case study #1 summary

## When to consider using a monochrome image sensor:

- Model accuracy doesn't depend on color information
- Low-light conditions
- Intensity-based features such as texture and edges are important
  - Bar code scanning, OCR, defect detection
- Low cost requirements
- Limited computational ability

## Assess:

- Is your model architecture designed for monochrome input?
- Is color important at any step in the application?
  - E.G. color might not be important for your object detection but is relevant for your classification

# Case Study #2: License Plate Recognition

# The models

1920×1080px



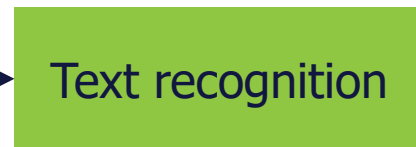
YOLOv2



YOLOv3



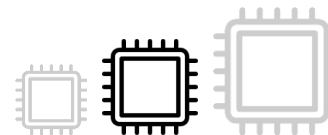
YOLO-Net & Tessaract OCR



8SLR965

6ZGE913

Computational  
requirement:



# Shutter type and speed

## Rolling shutter pros:

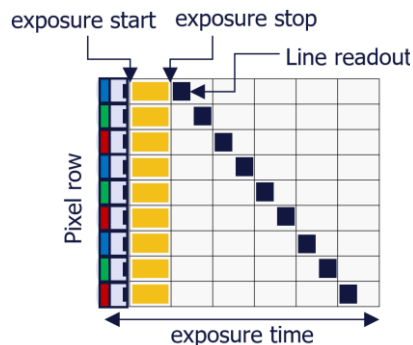
- Lower noise
- Lower cost
- Typically lower power

## Global shutter pros:

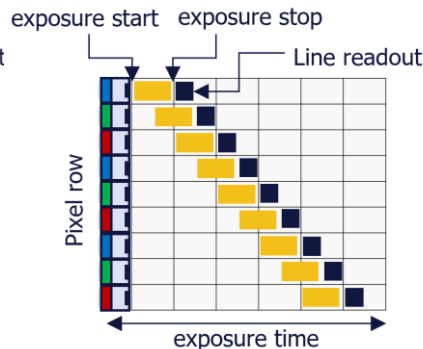
- Higher frame rates
- No motion distortion
- Better low-light quality
- Larger FOV



Global shutter

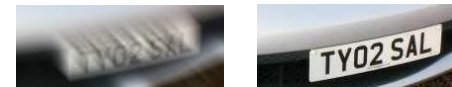


Rolling shutter



## High frame rates

- 1/60, 1/120, 1/240...
- Increased accuracy
  - Enable more detections
  - More likely to capture at least one clear image
- Reduced motion blur



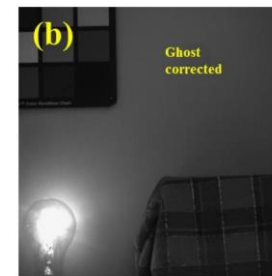
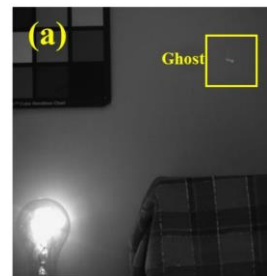
How fast must frame rate be to capture a car going 100 mph?

# Variable light conditions

Measurement	Definition	Good range
High dynamic range (HDR)	Includes temporal dark noise	120 dB – 140 dB
Signal to noise ratio (SNR)	Includes temporal dark noise, shot noise	40+ dB
ADC resolution (luminance)	Intensity of light captured	12-bit+
Near-infrared sensitivity (NIR)	Sensitivity to NIR wavelengths	650 nm to 2500 nm
Shutter efficiency ratio	Time all pixels exposed to light / shutter speed	90%+



Higher **shutter efficiency ratio** =  
less **ghosting** (light or motion artifacts)  
Rejects undesired light (outdoors)



# Case study #2 summary

## For moving objects, consider:

- Global shutter
- Frame rate

## For varied lighting conditions, including glare, overcast, strobing, and night, consider:

- HDR
- SNR
- ADC resolution
- NIR

## Assess:

- Is shutter speed fast enough to prevent motion blur but slow enough to let in enough light?
- Power requirements for HDR and higher frame rates?
- Color or monochrome sensor?

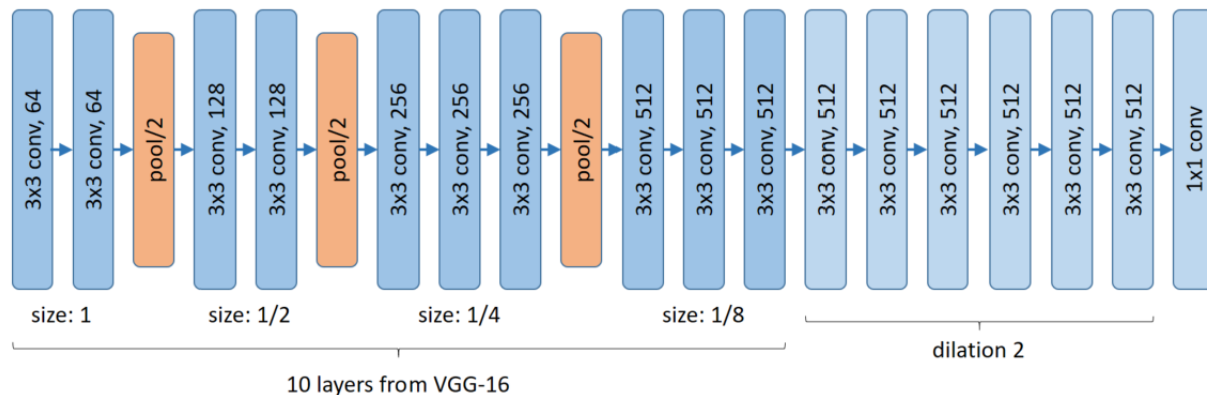
# Case Study #3: Crowd Counting

# The model

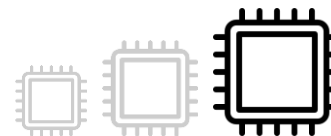
2048×1080 px



## CSRNet



Computational  
requirement:





# High resolution images

## How much resolution do you need?

Spatial resolution =  
(feature size) / (minimum resolution to find an edge)

$$15\text{cm}^* / 4\text{px}^* = 3.75\text{cm px}$$

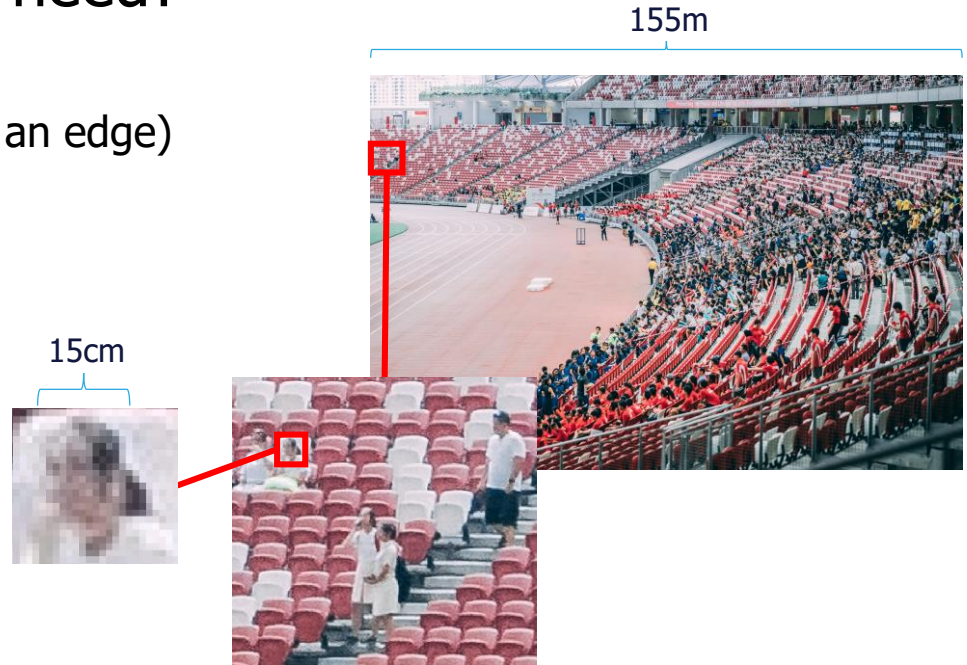
\*average diameter of a human head is 15cm

\*minimum of 3-4px in order to find an edge

Image sensor resolution =  
(FOV) / (spatial resolution)

$$155\text{m} / 3.75\text{cm px} = 4,133\text{px}^* \text{ wide}$$

\*This image is  
5758px wide



# Case study #3 summary

## When to consider high resolution image sensors:

- Large field of view is needed
  - Multiple cameras are not feasible
- Image pipeline requires cropping, resizing, or zooming
- Computation power, memory, and data interfaces are adequate

## Assess:

- How much resolution can camera interfaces handle at required speed?
- Are you able to collect or find a dataset with the desired resolution?

## Tips and Tricks for high resolution input:

- Downsampling
- Patch-based training
- Try monochrome input
- Trade-off with frame rate
- Trade-off with pixel size

1. Different embedded vision applications require different image sensors
2. Choice of image sensor has a significant impact on the accuracy, power, and computational cost of your vision system
3. Choosing the right image sensor will require trade-offs

## Consider:

- What features of the input image are most important?
  - Edges, textures, hue, etc.
- Lighting conditions
- Motion and speed (of subject or camera)
- Size of required field of view
- Size of ROI or detail

## Specifications

- [Understanding the In-Camera Image Processing Pipeline for Computer Vision](#)
- [Operating principle and features of CMOS sensors](#)
- [Image Sensor Terminology](#)

## Monochrome vs Color

- [How much resolution do I lose using a color industrial camera in a mono mode?](#)
- [Does Colour Really Matter? Evaluation via Object Classification](#)

## Shutter Type and Speed

- [Real-Time Camera Tracking: When is High Frame-Rate Best?](#)
- [Global Shutter Efficiency Improvement to >100dB in Advanced Global Shutter Imager](#)

## Resolution

- [Imaging Electronics 101: Camera Resolution for Improved Imaging System Performance](#)
- [Efficient High-Resolution Deep Learning: A Survey](#)

# More information on models and techniques

## Defect Detection

- [Faster R-CNN](#)
- [Tire Defect Detection Using Fully Convolutional Network](#)

## License Plate Recognition

- [License Plate Recognition in Urban Road Based on Vehicle Tracking and Result Integration](#)
- [License Plate Detection and Recognition in Unconstrained Scenarios](#)

## Crowd Counting

- [CSRNet](#)
- [NWPU-Crowd: A Large-Scale Benchmark for Crowd Counting and Localization](#)
- [To choose or to fuse? scale selection for crowd counting](#)
- [Efficient High-Resolution Deep Learning: A Survey](#)