

# Intro To Computer Vision Using OpenCV

Evan Stoddart

03/24/2018



**THE OHIO STATE UNIVERSITY**  
CENTER FOR AUTOMOTIVE RESEARCH

# About Me

## *Evan Stoddart*

*ADAS Subteam Leader*

*OSU EcoCAR 3*

*Graduate Research Assistant*

*OSU Center For Automotive Research*

### Hometown:

*Pittsburgh, PA*

### Major:

*Electrical and  
Computer Engineering*

### Year:

*4<sup>th</sup> Year BS, 1<sup>st</sup> Year MS*

### Internships:

*Hewlett Packard, STERIS  
Corporation, Bosch*

### Interests:

*robotics, cars, and medical devices*



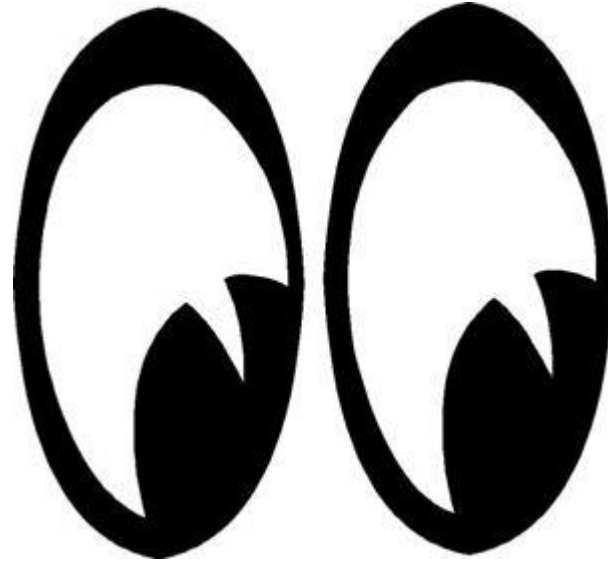


**THE OHIO STATE UNIVERSITY**

CENTER FOR AUTOMOTIVE RESEARCH

# What is Computer Vision?

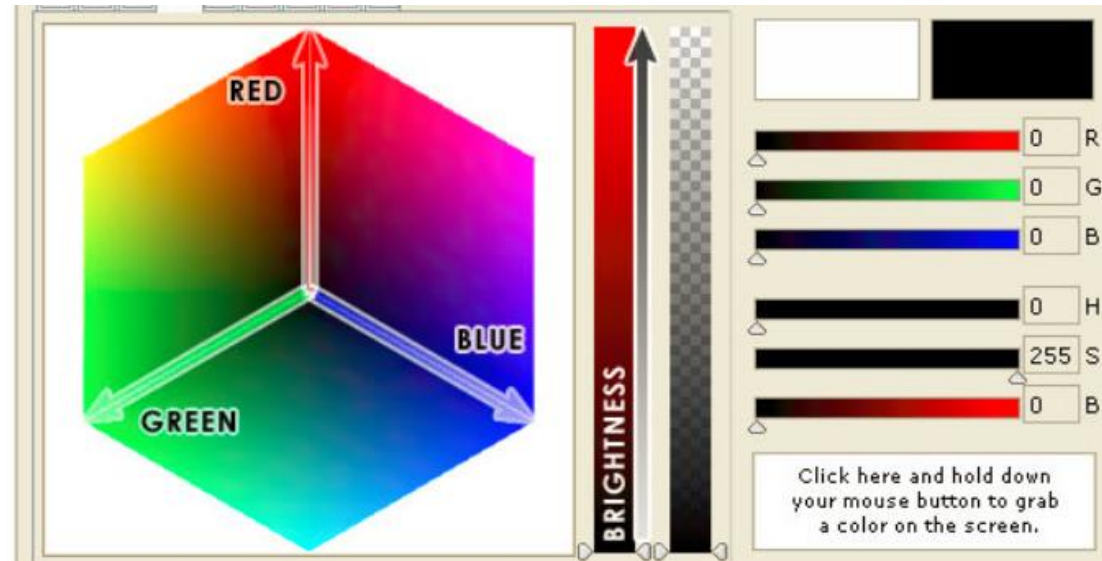
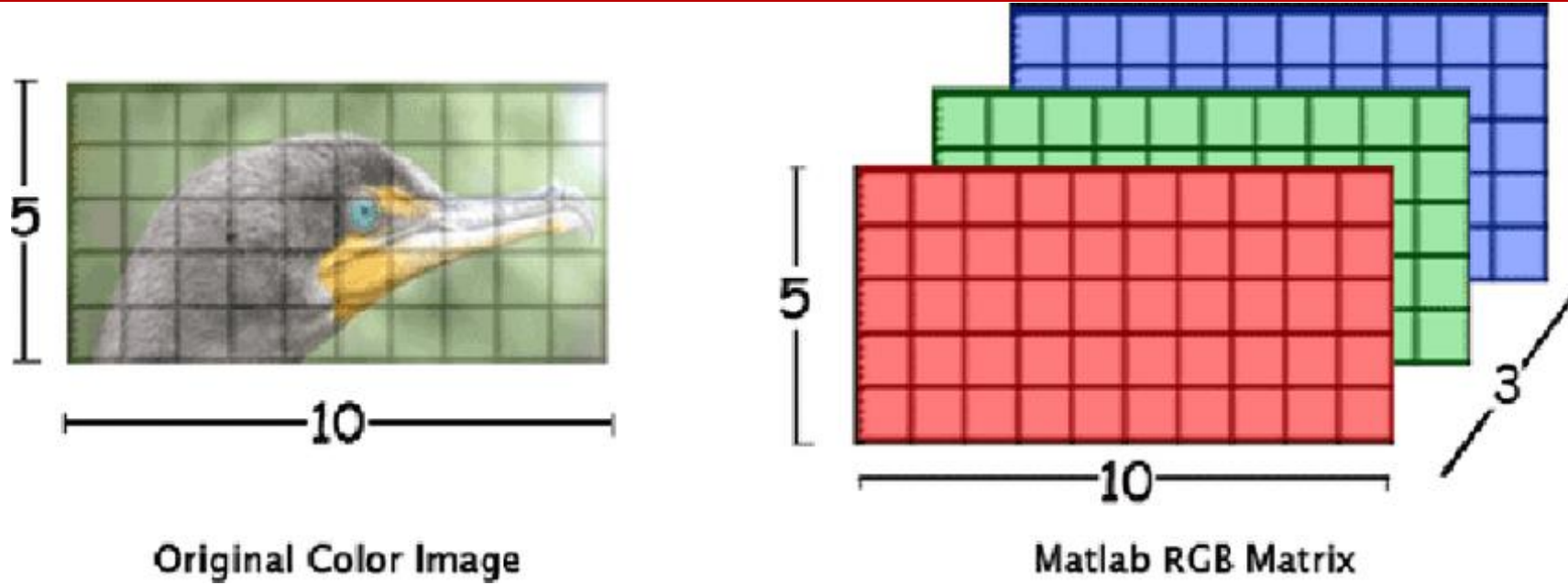




*“Computer Vision (CV) is an interdisciplinary field that deals with how computers can be made for gaining high-level understanding from digital images or videos. From the perspective of engineering, it seeks to automate tasks that the human visual system can do”*

*- Wikipedia*

# How do We Automate Tasks using CV?

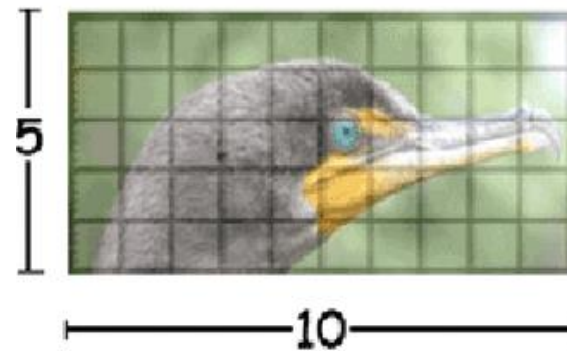




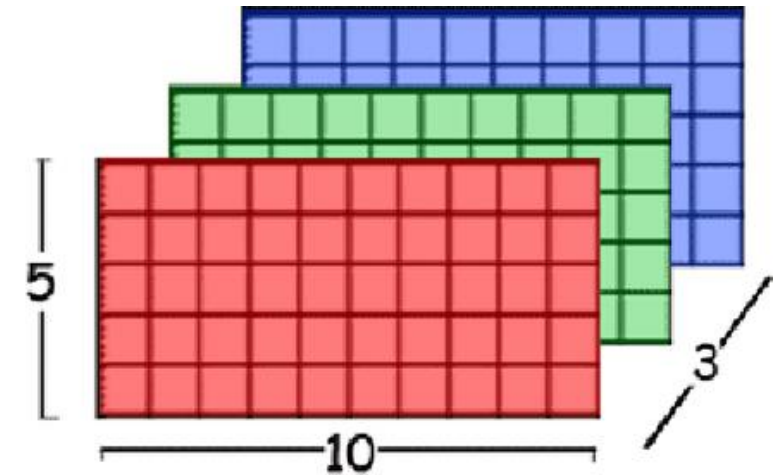
# How do We Automate Tasks using CV?

The automation strategy depends on what data we want to find in the image.

- Kernels
  - Math used behind the scenes
- Image Processing
  - Basic operations on images
  - Crop images to gather a Region of Interest (ROI)
  - Adjust image for contrast, color space
- Object Detection/Classification
  - Search for an object in the image
  - Pattern Recognition
  - Machine Learning
  - Deep Learning
- Post Processing
  - Collect information about the object
  - Filter Noise from Data
  - Area of Object
  - Position of Object
  - Color of Object



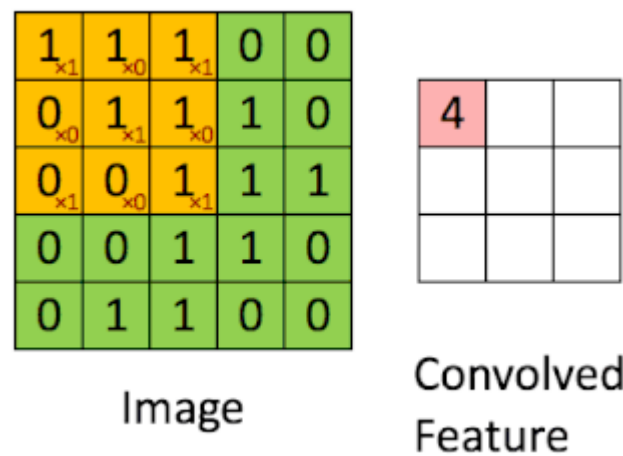
Original Color Image



Matlab RGB Matrix

# Kernels: The Basic Math used in CV

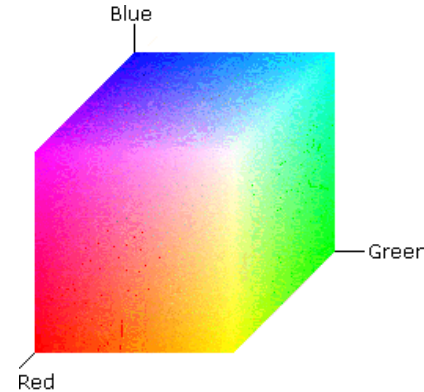
- Kernels
  - Smaller matrix that mathematically describes the desired operation
- Convolution (Sliding Window)
- When you use Photoshop, this is the math that is happening behind the scenes!
- Used in many CV algorithms for pre processing, object detection, filters and more!



Operation	Kernel	Image result
Identity	$\begin{bmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix}$	
Edge detection	$\begin{bmatrix} 1 & 0 & -1 \\ 0 & 0 & 0 \\ -1 & 0 & 1 \end{bmatrix}$	
	$\begin{bmatrix} 0 & 1 & 0 \\ 1 & -4 & 1 \\ 0 & 1 & 0 \end{bmatrix}$	
	$\begin{bmatrix} -1 & -1 & -1 \\ -1 & 8 & -1 \\ -1 & -1 & -1 \end{bmatrix}$	
Sharpen	$\begin{bmatrix} 0 & -1 & 0 \\ -1 & 5 & -1 \\ 0 & -1 & 0 \end{bmatrix}$	

# Image Processing

- Crop
- Change color space
  - Grayscale
  - HSV
  - RGB
- Filters
  - Gaussian Blur
- Morphological Operations
  - Dilation - grow image regions
  - Erosion - shrink image regions
  - Opening - structured removal of image region boundary pixels
  - Closing - structured filling in of image region boundary pixels



$1/16$	$1/8$	$1/16$
$1/8$	$1/4$	$1/8$
$1/16$	$1/8$	$1/16$





# Object Detection/Classification

**Classification**



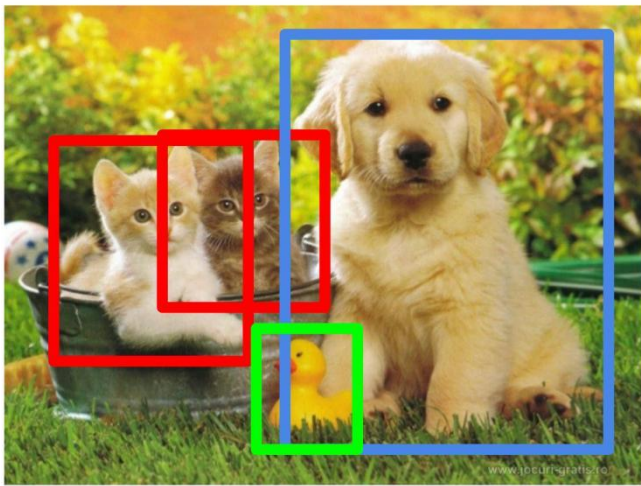
CAT

**Classification  
+ Localization**



CAT

**Object Detection**



CAT, DOG, DUCK

**Instance Segmentation**



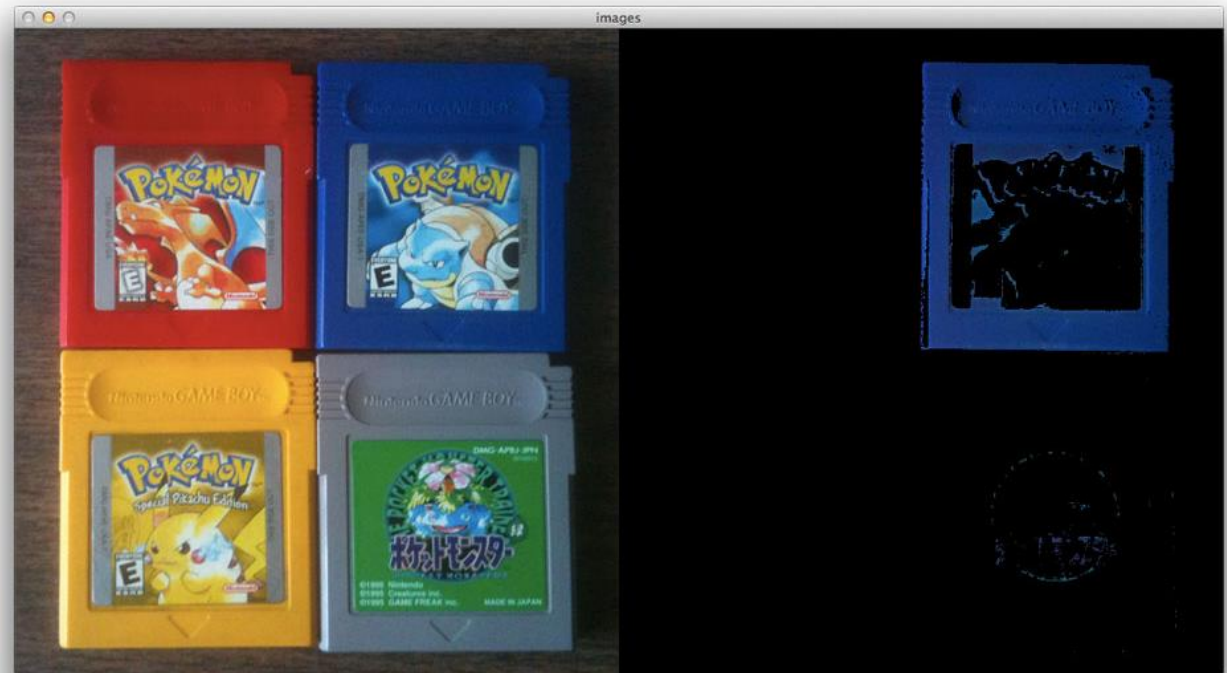
CAT, DOG, DUCK

Single object

Multiple objects

# Pattern Recognition

- Pattern recognition was a term popular in the 70s and 80s.
- Getting a computer program to do something “smart” like recognize the character "3" in an image
  - Algorithms are manually coded by programmer
    - Tedious
    - Not robust
  - “Learning” is not performed by the computer
- Examples:
  - Color Thresholding/Masking
  - Edge Detection

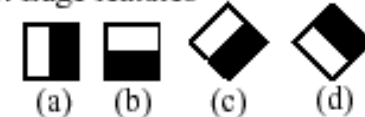


# Machine Learning

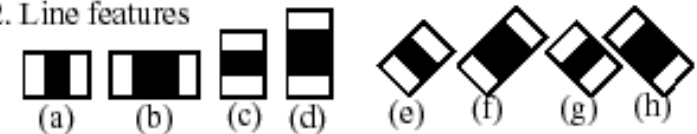
- Sometime in the early 90s people started realizing that a more powerful way to build pattern recognition algorithms is to replace an expert with data.
- Collect data (ex. a bunch of face images and non-face images), choose an algorithm, and wait for the computations “training” to finish.
- "Machine Learning" emphasizes that the computer program (or machine) must do some work to “learn” after it is given data.
- Example Algorithms:
  - Haar Features
  - HOG (Histogram of Gradients)
  - Easy to use in OpenCV,  
all you need is data! More on this later...



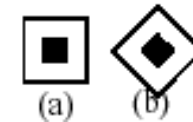
1. Edge features



2. Line features

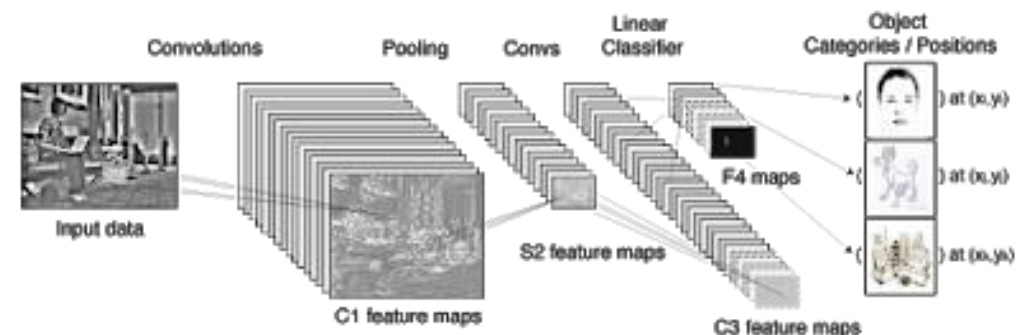


3. Center-surround features



# Deep Learning

- Similar to Machine Learning in that the computer “learns” from exposure to training data
- Except, Deep Learning programs learn not only the data, but developed its own algorithm to optimize finding the data within an input image
- Deep Learning is built on Convolutional Neural Networks
  - AlexNet
    - Used in the ImageNet Large Scale Visual Recognition Challenge in 2012.
    - The network achieved a top-5 error of 15.3%, more than 10.8 percentage points ahead of the runner up.
  - Inception
    - Modern network
    - trained on more than a million images and can classify images into 1000 object categories
- Mathematically, Deep Learning is hard, use a Deep Learning Frameworks to make it easier
  - Caffe from University of California at Berkeley
  - TensorFlow from Google
  - PyTorch/Torch
- Deep Learning is computationally expensive
  - Use GPU to run the algorithm in parallel





# Post Processing

- Filters
  - Object detection isn't always perfect → noise
  - Filters are used to remove unwanted data
  - Examples: Temporal (Heat Map)
- Tracking
  - Persistently track an object as it moved about the screen, image by image, in a video
- Physical Traits
  - Obtain information from image objects
  - Size of object
  - Position of object in space
    - Depth estimation



		Prediction outcome		total
		p	n	
actual value	p'	True Positive	False Negative	P'
	n'	False Positive	True Negative	N'
total		P	N	



# How do We Automate Tasks using CV?

- In theory, implementing CV algorithms into a program can be mathematical and convoluted (no pun intended)
- In practice, libraries exist which abstract the math to provide the desired functionality.
  - More often than not, the function you need is built into a library
  - All you need is a basic understanding of the function, enough to enter parameters needed for your application



**THE OHIO STATE UNIVERSITY**

CENTER FOR AUTOMOTIVE RESEARCH

# What is OpenCV?



# What is OpenCV?

- OpenCV is a widely used open-source computer vision framework
  - Over 14 million downloads
  - Library available in C++ and Python
  - Compatible with Tensorflow, Torch/PyTorch, and Caffe Deep Learning Frameworks
  - Open source – anyone can use, edit, contribute to the library
  - First released in 1999 by Intel Corporation
  - Website: [Opencv.org](http://opencv.org)



# Example Functions

- Using Python
- Image Preprocessing
  - Crop
  - Change colorspace (grayscale)
- Pattern Recognition
  - Color thresholding
- Machine Learning
  - Haar Classifier
- Post Processing
  - Temporal Filter
  - Overlaying Image

# What to do if you get stuck

- Forums & Online Resources
  - Stackoverflow
  - pyimagesearch.com
  - OpenCV.org
  - Other application specific forums and websites
- Useful tips for Google searching
  - OpenCV syntax varies by version(popular versions are 2.4, 3.0, 3.3)
  - When searching, specify the version you are using as well as the language (Python/C++)
    - Include the function you are interested in
    - Include an error code if one was printed by your program
    - Example:
      - OpenCV 2.14 Python 2.7 ImportError: DLL load failed: %1 is not a valid Win32 application





**THE OHIO STATE UNIVERSITY**

CENTER FOR AUTOMOTIVE RESEARCH

# Computer Vision Applications



# CV Applications

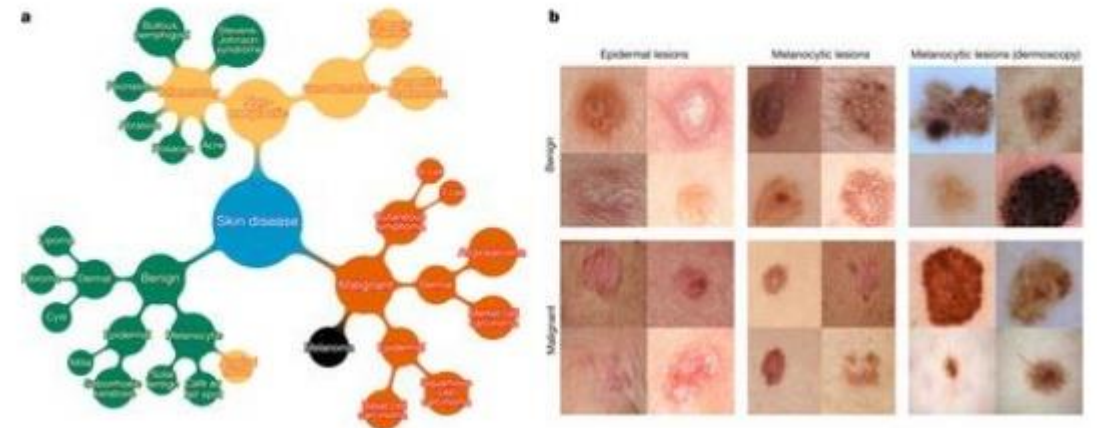
- Robotics
- Medical Devices
- Automotive
- ...the list goes on!



## Classification: skin cancer detection

Images organized in a **tree taxonomy** of 2032 diseases (by medical experts)

CNN trained **757** disease classes: a disease partitioning algorithm to generate classes clinically and visually similar





**THE OHIO STATE UNIVERSITY**

CENTER FOR AUTOMOTIVE RESEARCH

# EcoCAR ADAS Applications



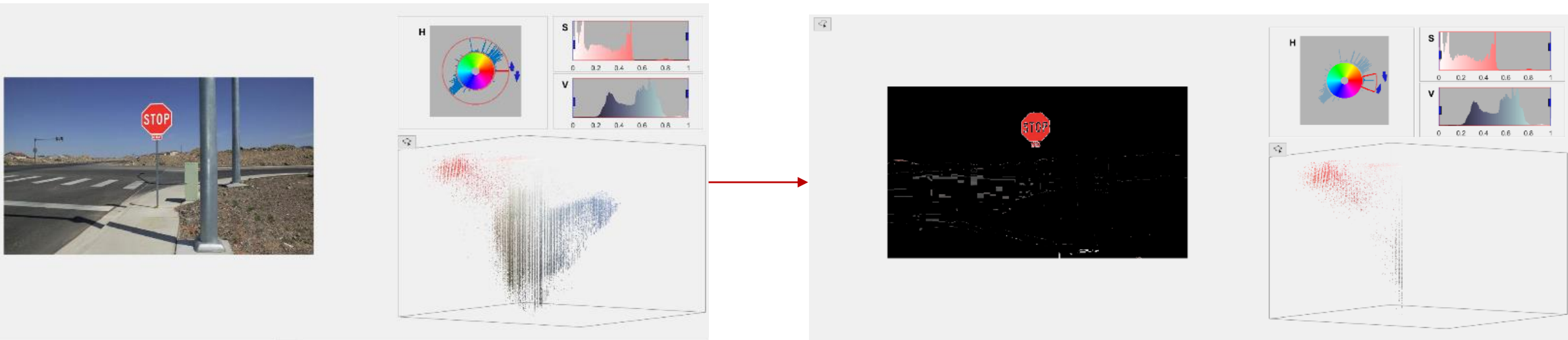
# EcoCAR ADAS Applications

- Advanced Driver Assistance Systems (ADAS)
  - Forward Collision Warning
  - Lane Departure Warning
  - Park assist
  - Adaptive cruise control
- We can use cameras with Computer Vision as well as other sensors like Radar and LiDAR to develop this functionality



# Pattern Recognition

- Color thresholding:
  - An image is just a matrix of values where each value represents a color.
  - We can use computer algorithms to show the colors we want to see and “mask” those that we don’t by selecting corresponding ranges of these values.
  - ADAS application: detect the color red to produce a stop sign detector.



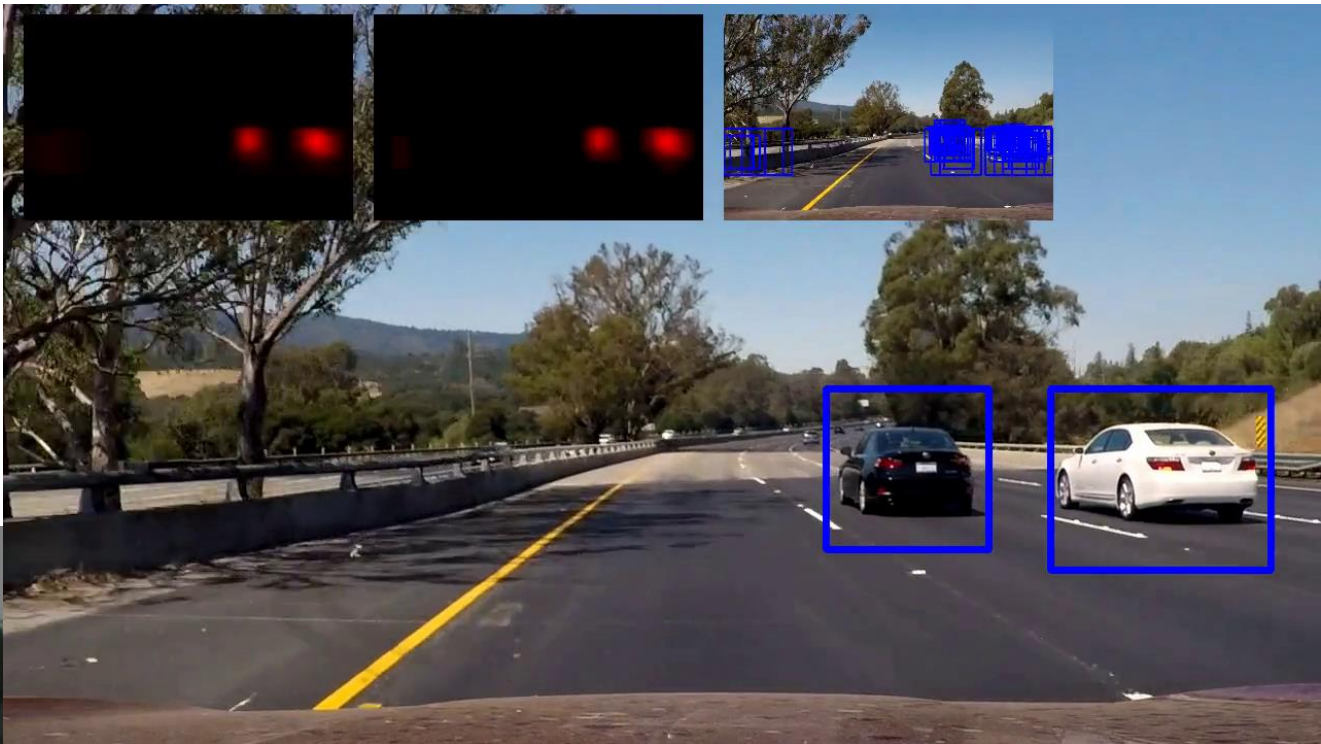


# Pattern Recognition

- Canny edge detector + Hough transforms:
  - We can use computer algorithms to highlight edges (i.e. areas in the image with large changes in color/pixel value) – Canny edge detector.
  - After detecting edges, we can determine which edges belong to lines using Hough transforms.
  - ADAS application: detect lines on the road to produce a lane detector/lane departure warning system.



# Object Detection





**THE OHIO STATE UNIVERSITY**

CENTER FOR AUTOMOTIVE RESEARCH

# QUESTIONS?

