Classification Of Singapore Traffic Road Sign

Al Applications

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Al Applications Student | CV Graded Assignment

Honor Pledge for Graded Assignments

"I affirm that I have not given or received any unauthorized help on this assignment, and that this work is my own."

Signature:





Summary

Data Pre-Processing

Model Creation

Prediction Results

 Difficulties Encountered & Learning Points

Data Pre-Processing

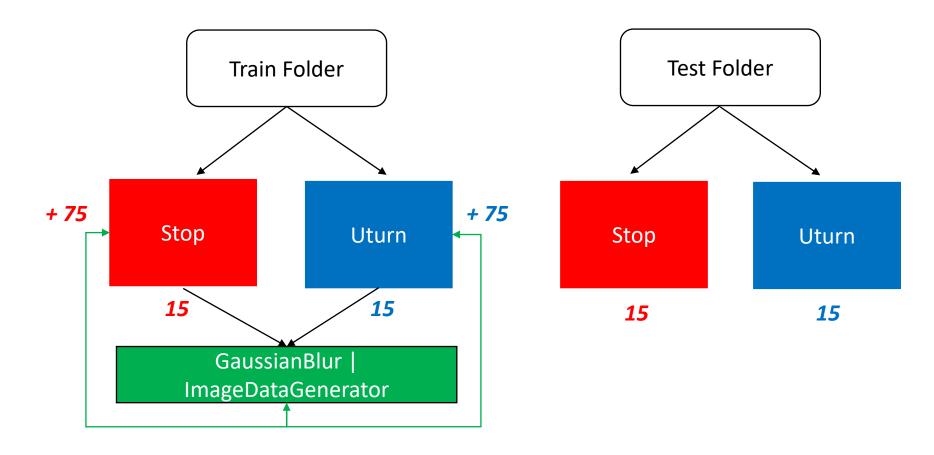
 Train Test Split Uturn Sign Folder *30* Stop Sign Folder *30* Train Test Split | 50% : 50% Test Folder Train Folder Stop Uturn Stop Uturn **15** *15* **15** *15*

Data Pre-Processing

DataAugmentation

 Add 75 more augmented images

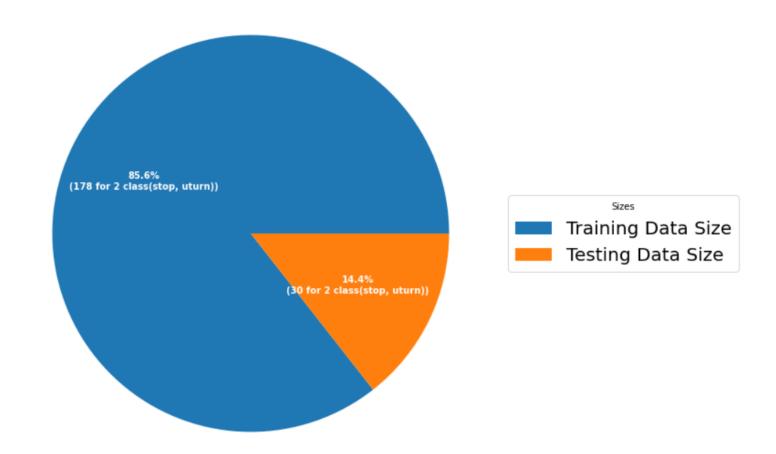
 Both classes in Train Folder



Data Pre-Processing

• Pie Chart

Training & Testing Size Comparison



- Convolutional Neural Network
- 2 Conv Layers
- 2 MaxPooling Layers
- 1 Flatten Layer
- Ouput Dense Layer

Model: "sequential_8"

Layer (type)	Output Shape	Param #
conv2d_14 (Conv2D)	(None, 150, 150, 64)	1792
<pre>max_pooling2d_14 (MaxPoolin g2D)</pre>	(None, 75, 75, 64)	Ø
conv2d_15 (Conv2D)	(None, 75, 75, 64)	36928
<pre>max_pooling2d_15 (MaxPoolin g2D)</pre>	(None, 37, 37, 64)	Ø
flatten_7 (Flatten)	(None, 87616)	Ø
dense_7 (Dense)	(None, 1)	87617
		-======

Total params: 126,337

Trainable params: 126,337 Non-trainable params: 0

Model Architecture | Kenneth | CV Graded Assignment

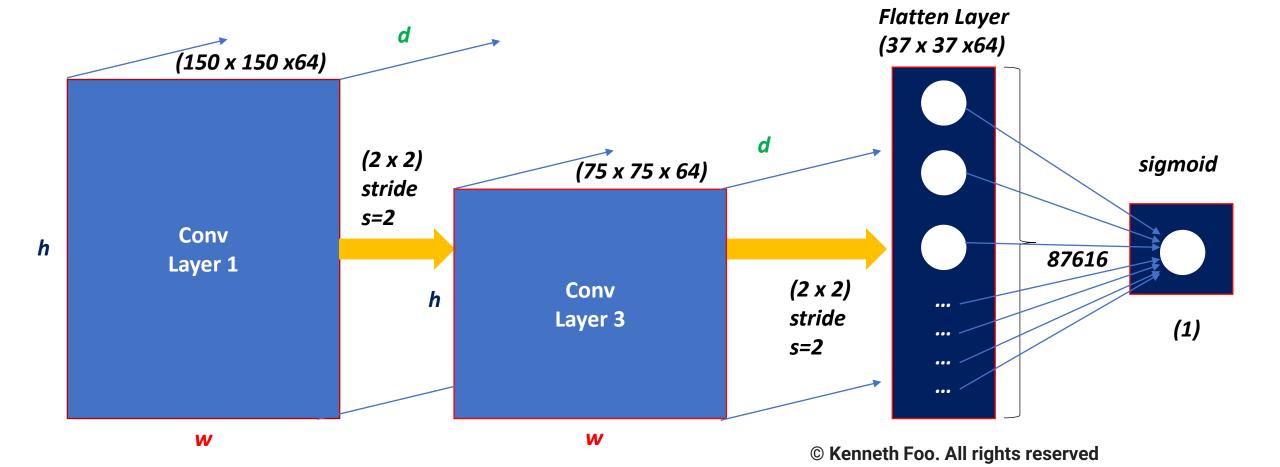
Layer Dimensions

Stride

= MaxPooling Layer

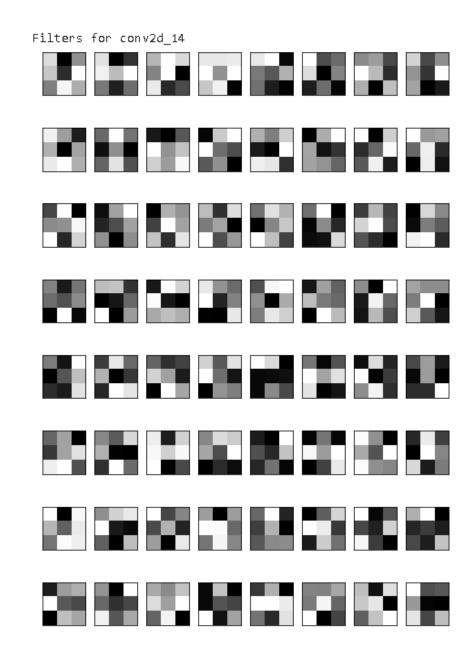
 $(h \times w \times d)$ (Filter step size)

 $h = Height \mid w = Width \mid d = (depth) = number of filters$



 Understand the filters from the first layer

• 3 by 3 by 64 filter

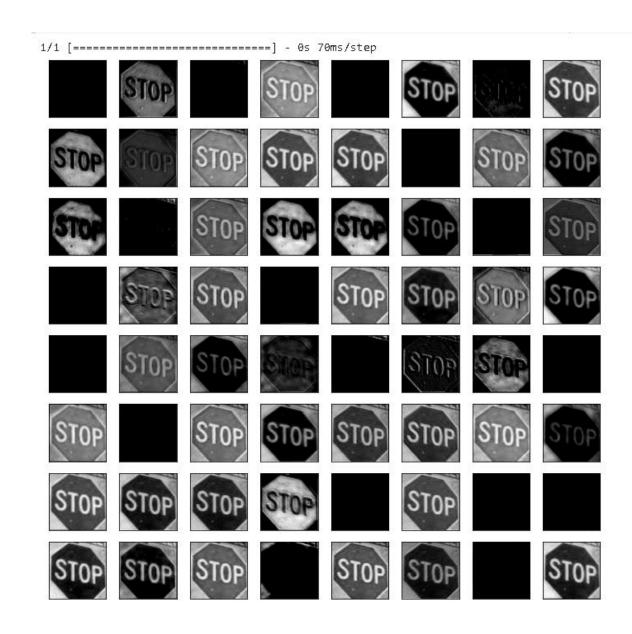


 Define a new truncated model to only include the conv layers of interest

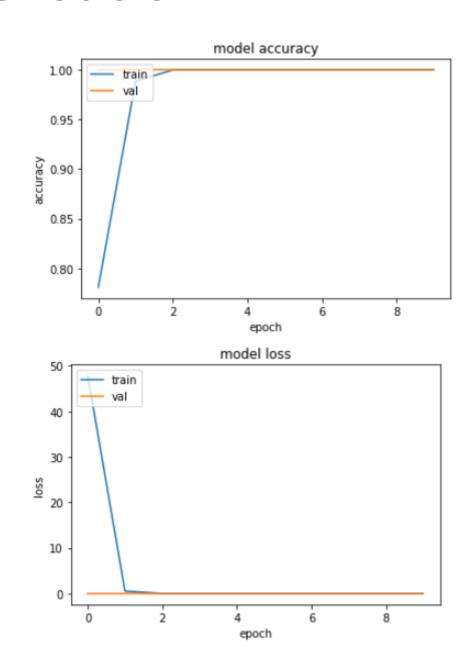
Generate feature output by predicting on input image

Representation Learning

 Low & High level features on the right side. ->



 Model Performance during training and validation



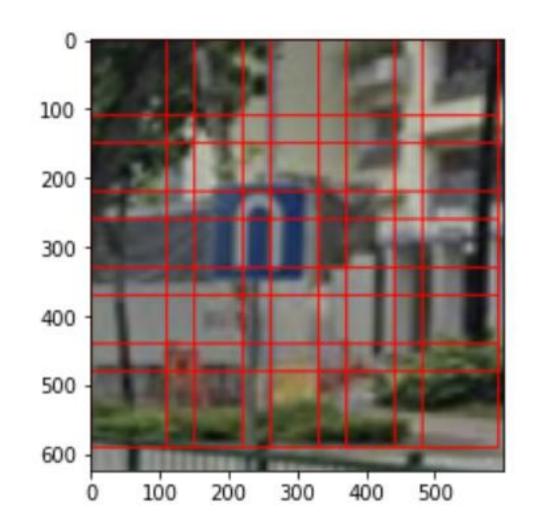
Prediction Results

- Object Detection for Image
- Constructing a sliding window

Constructing a image pyramid

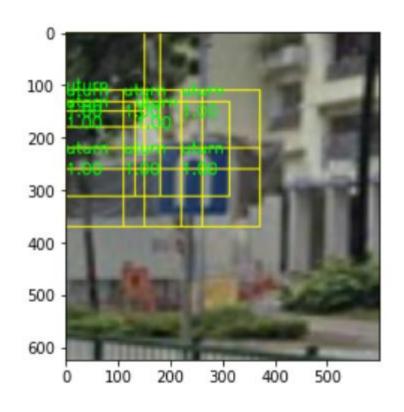
Adjust window for optimal parameters

Select image for detection



Prediction Results

- Object Detection for Image
- Looping over pyramid
- Classifying ROIs
- Do Prediction
- Generate Object Detection Results



highest prob label uturn

Difficulties Encountered & Learning Points

"The first difficulty I encountered was the overfitting of my model. I believe the reason for this was because the training and testing sets were too simple for the model to learn, so it probably became overconfident. The way to counter this issue, is to provide an equal ratio of images to both train & test sets. Also, when doing data augmentation, try to not only generate images, but to look at them at different perspectives or maybe you can even flip the image. Reason for doing that is to give a little confusion and information to the model so it learns better and performs better during predictions. The second difficulty encountered was constructing the parameters for the sliding window during image object detection, as different parameters are needed for the difference in images."

End of Presentation