



Advancing Road Safety: Traffic Sign Recognition
Modified LeNet-5 CNN Classification on GTSRB Dataset

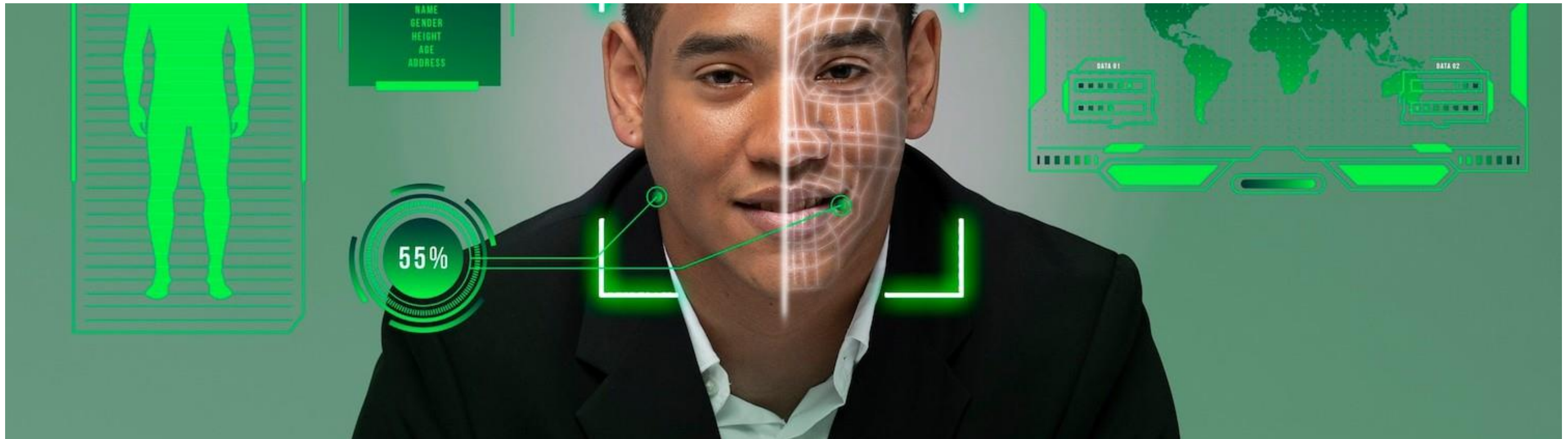
Introduction to GTSRB

- **German Traffic Sign Recognition Benchmark**
- Introduced in 2011: large, lifelike dataset capturing real-world variability.
- 43 classes, 51,839 images
- Image sizes: 15x15 to 250x250 pixels
- Train, Test, Meta (csv files for class reference)
- From video sequences on German roads
- Reflects actual sign frequency on roads
- natural class imbalance, varying illumination, and occlusions

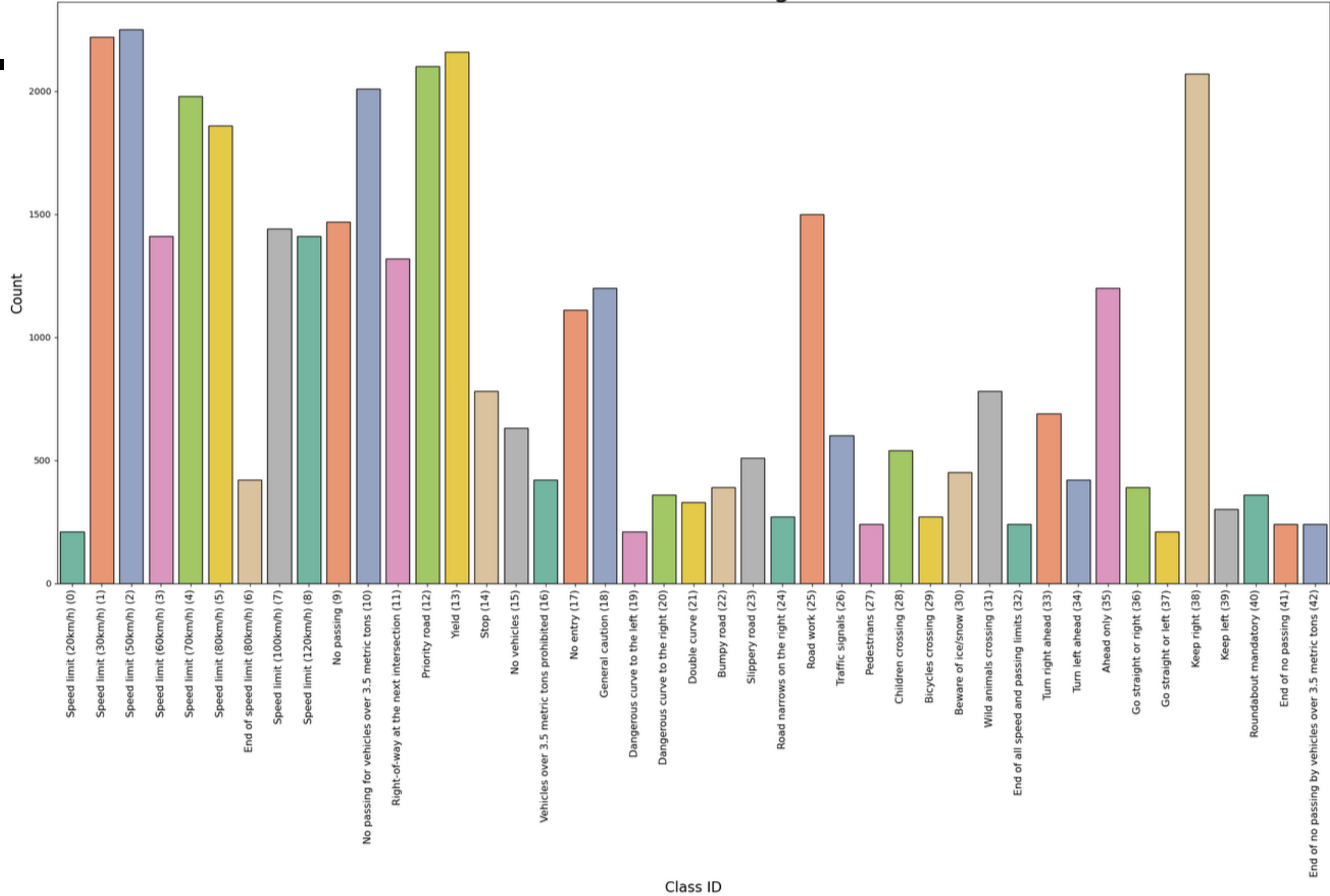


Why Traffic Sign Recognition?

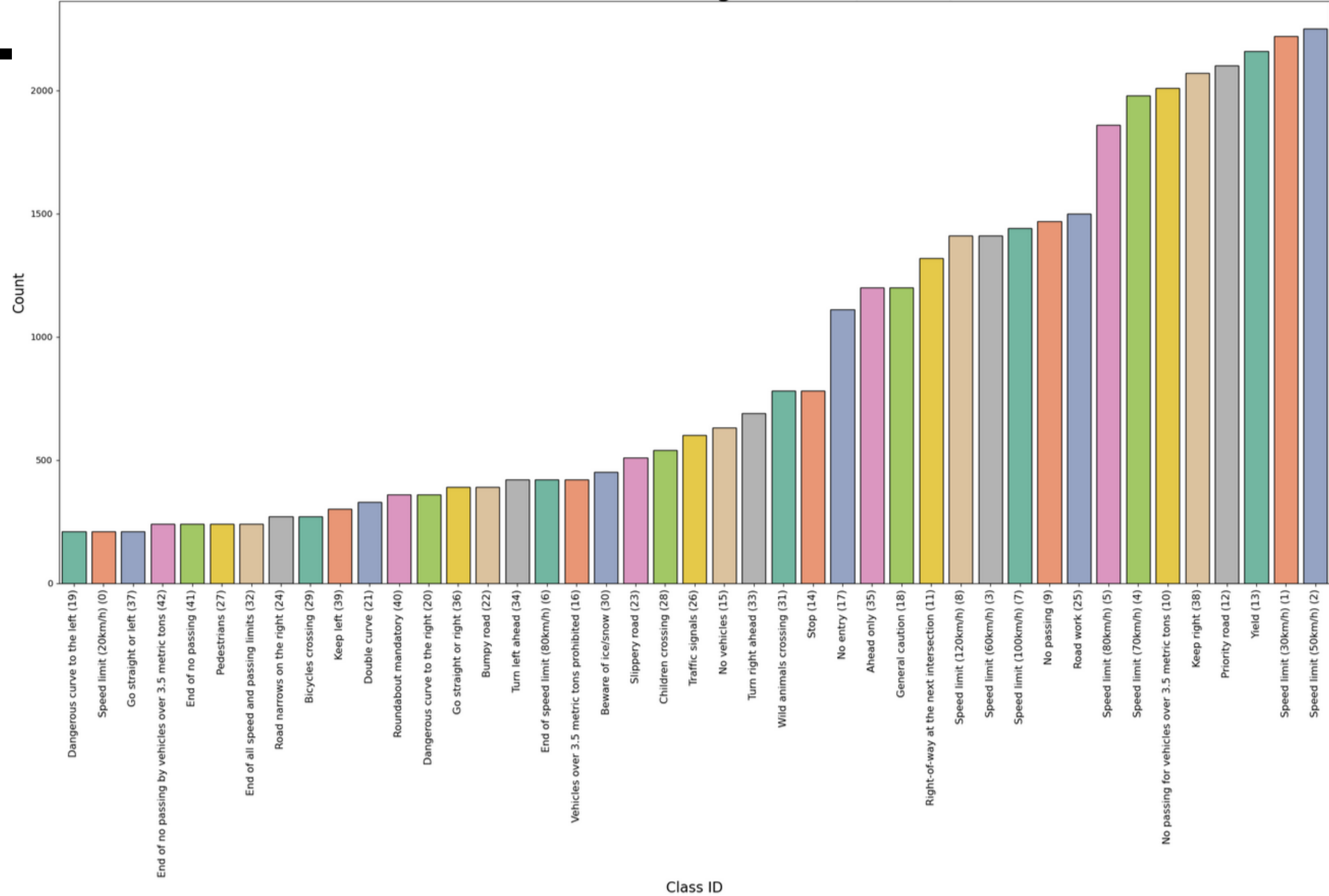
- ADAS Integration: Real-time speed limit detection and warning
- Traffic Monitoring: Automated sign inventory and maintenance
- GPS Enhancement: Context-aware navigation instructions
- Potential for reducing accidents
- Keras/Tensorflow help simulate the sign monitoring system with CNN architecture!



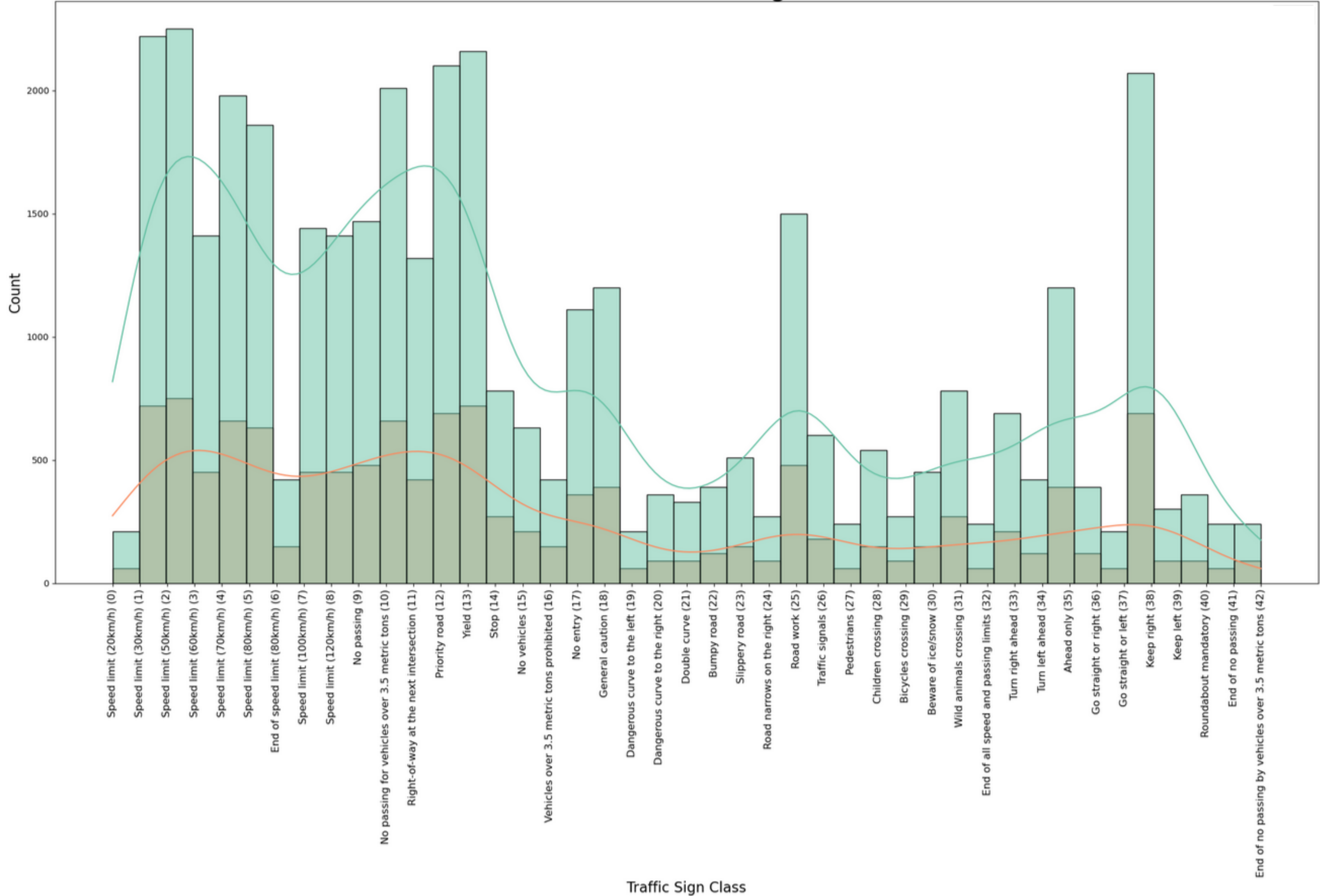
Distribution of Training Classes



Distribution of Training Classes (Sorted)



Distribution of Traffic Sign Classes



Top Ten Traffic Sign Classes

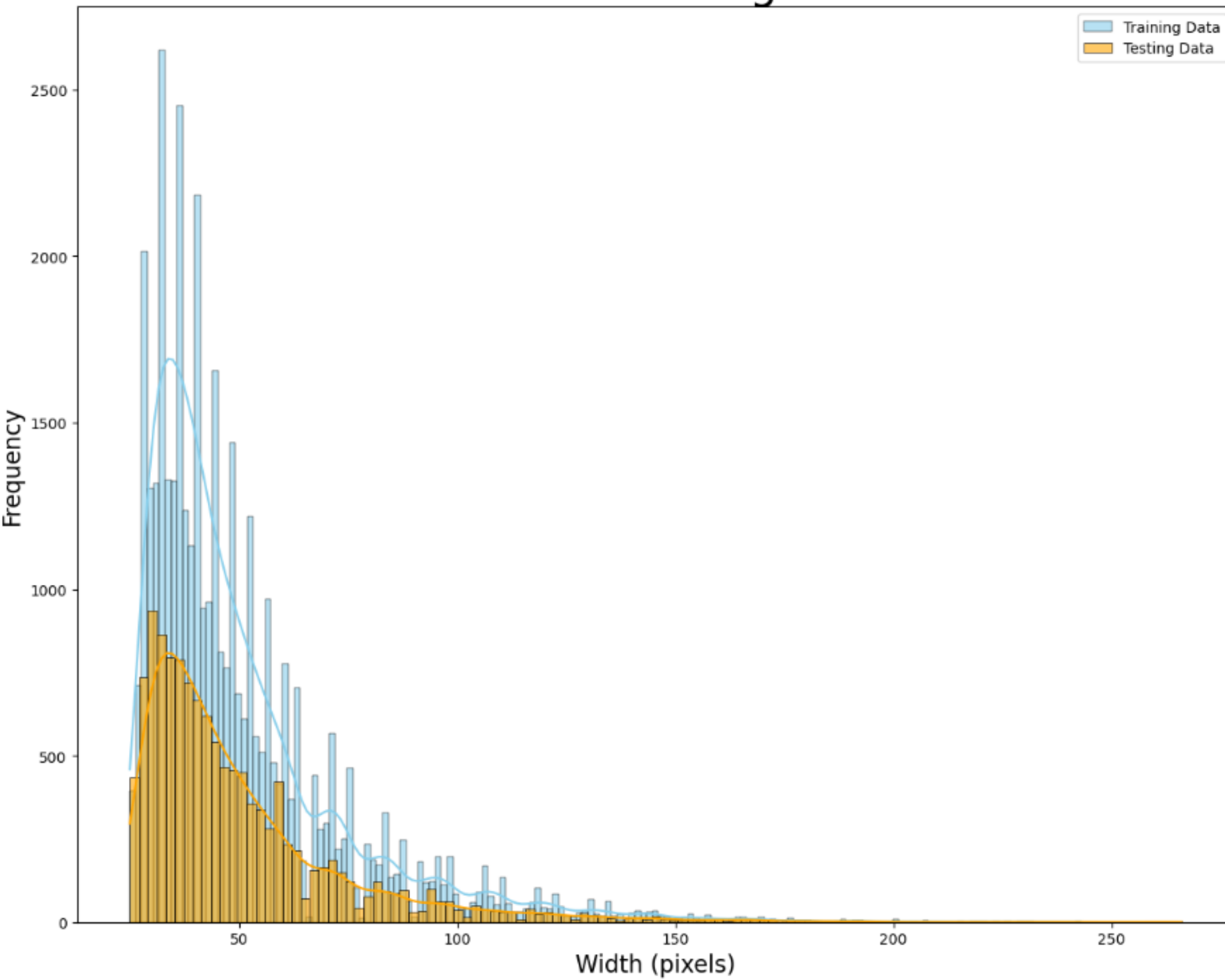
ClassId	Train Count	Train Percentage	Test Count	Test Percentage	SignName
2	2250	5.738478	750	5.938242	Speed limit (50km/h)
1	2220	5.661965	720	5.700713	Speed limit (30km/h)
13	2160	5.508939	720	5.700713	Yield
12	2100	5.355913	690	5.463183	Priority road
38	2070	5.279400	690	5.463183	Keep right
10	2010	5.126374	660	5.225653	No passing for vehicles over 3.5 metric tons
4	1980	5.049861	660	5.225653	Speed limit (70km/h)
5	1860	4.743809	630	4.988124	Speed limit (80km/h)
25	1500	3.825652	480	3.800475	Road work
9	1470	3.749139	480	3.800475	No passing

Bottom Ten Traffic Sign Classes

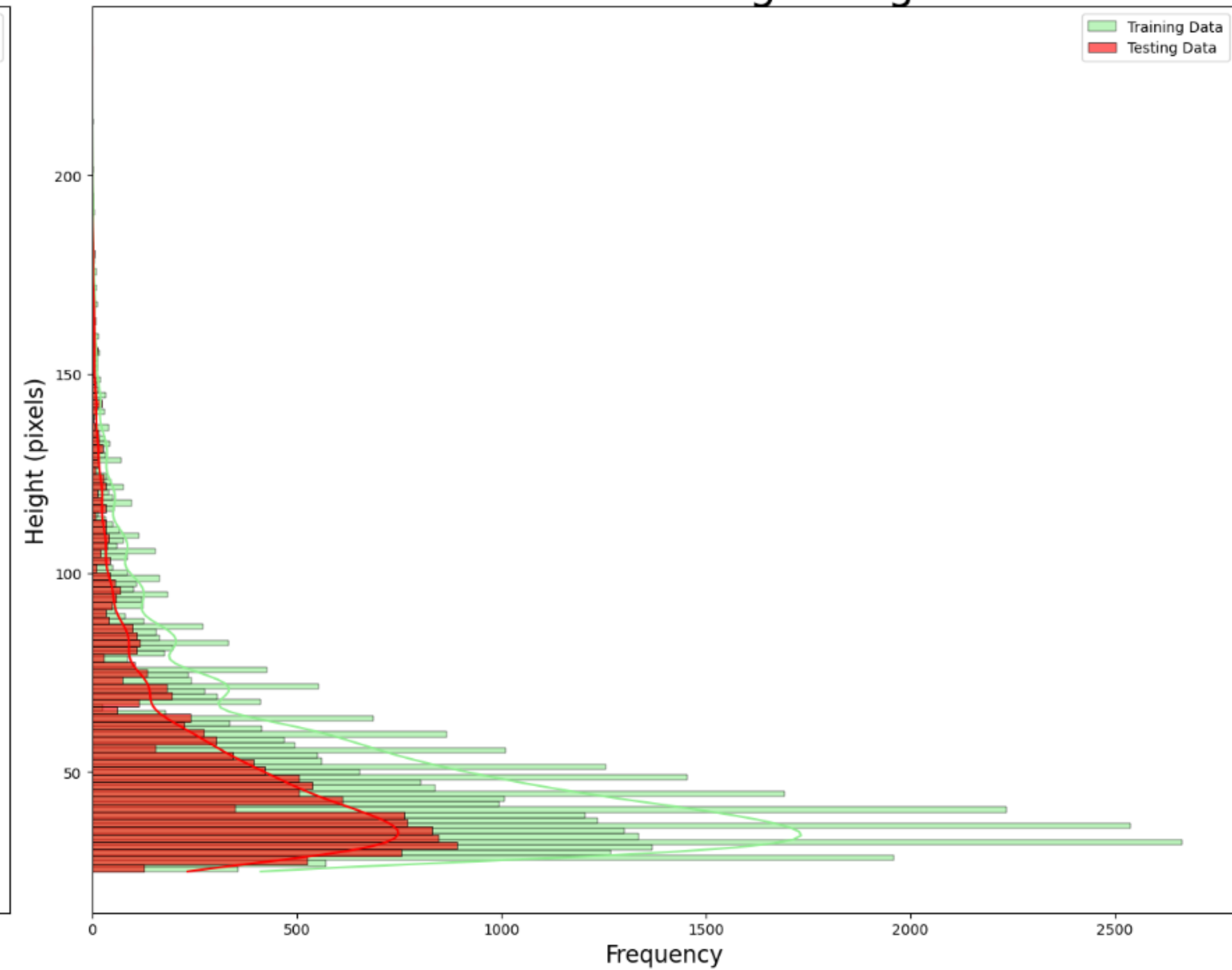
ClassId	Train Count	Train Percentage	Test Count	Test Percentage	SignName
19	210	0.535591	60	0.475059	Dangerous curve to the left
0	210	0.535591	60	0.475059	Speed limit (20km/h)
37	210	0.535591	60	0.475059	Go straight or left
42	240	0.612104	90	0.712589	End of no passing by vehicles over 3.5 metric ...
41	240	0.612104	60	0.475059	End of no passing
27	240	0.612104	60	0.475059	Pedestrians
32	240	0.612104	60	0.475059	End of all speed and passing limits
24	270	0.688617	90	0.712589	Road narrows on the right
29	270	0.688617	90	0.712589	Bicycles crossing
39	300	0.765130	90	0.712589	Keep left

CSV Metadata

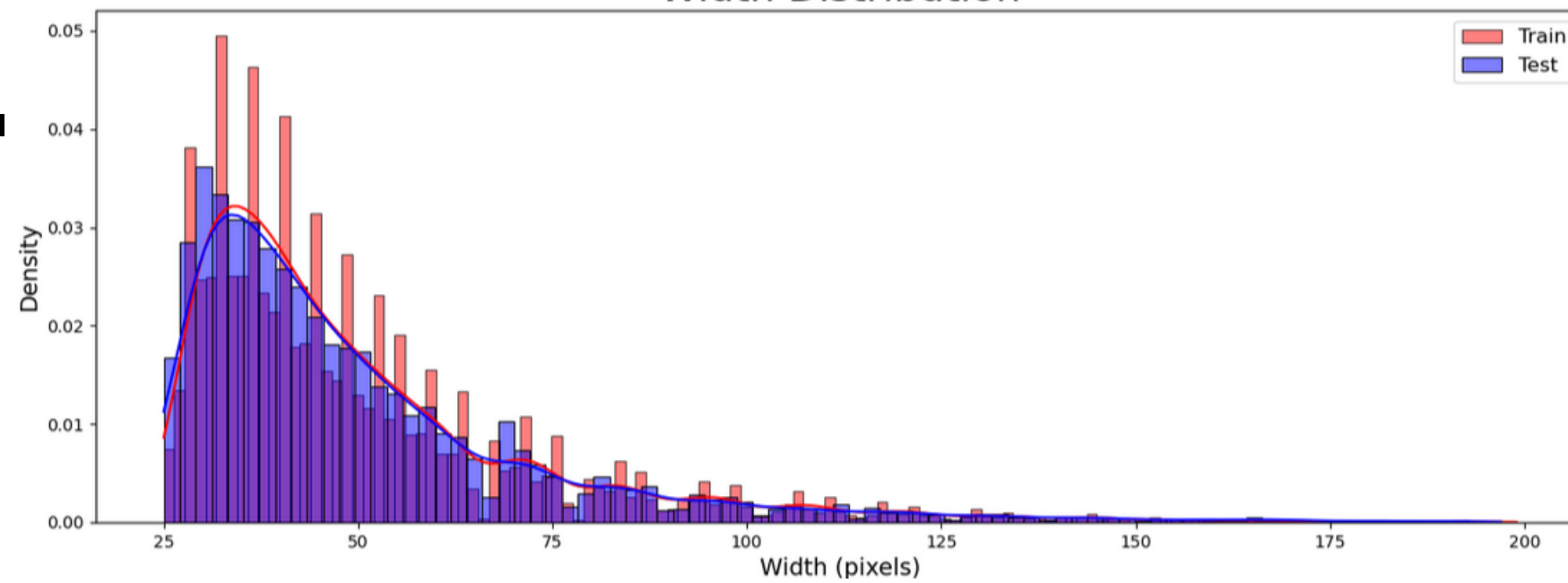
Distribution of Image Widths



Distribution of Image Heights



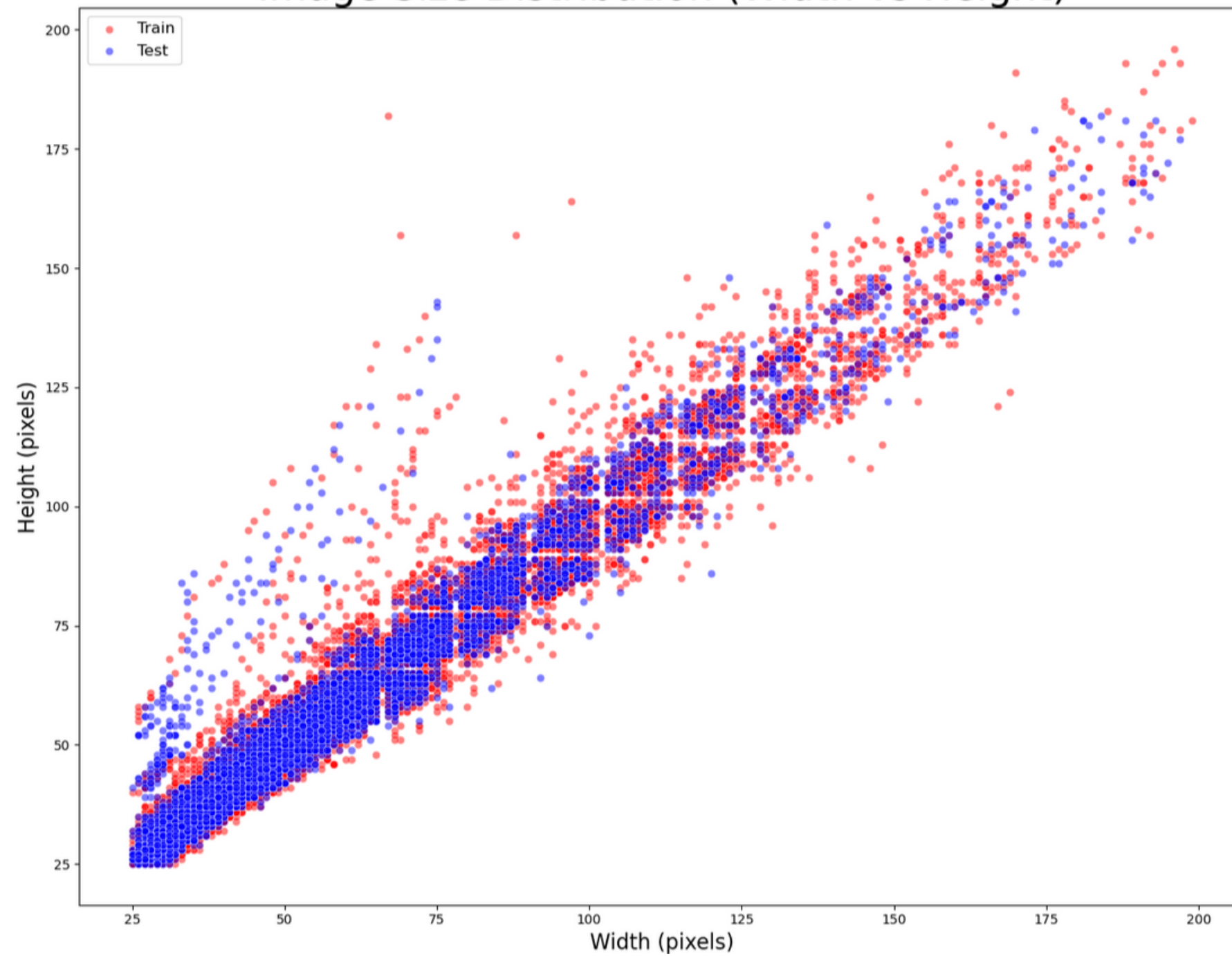
Width Distribution



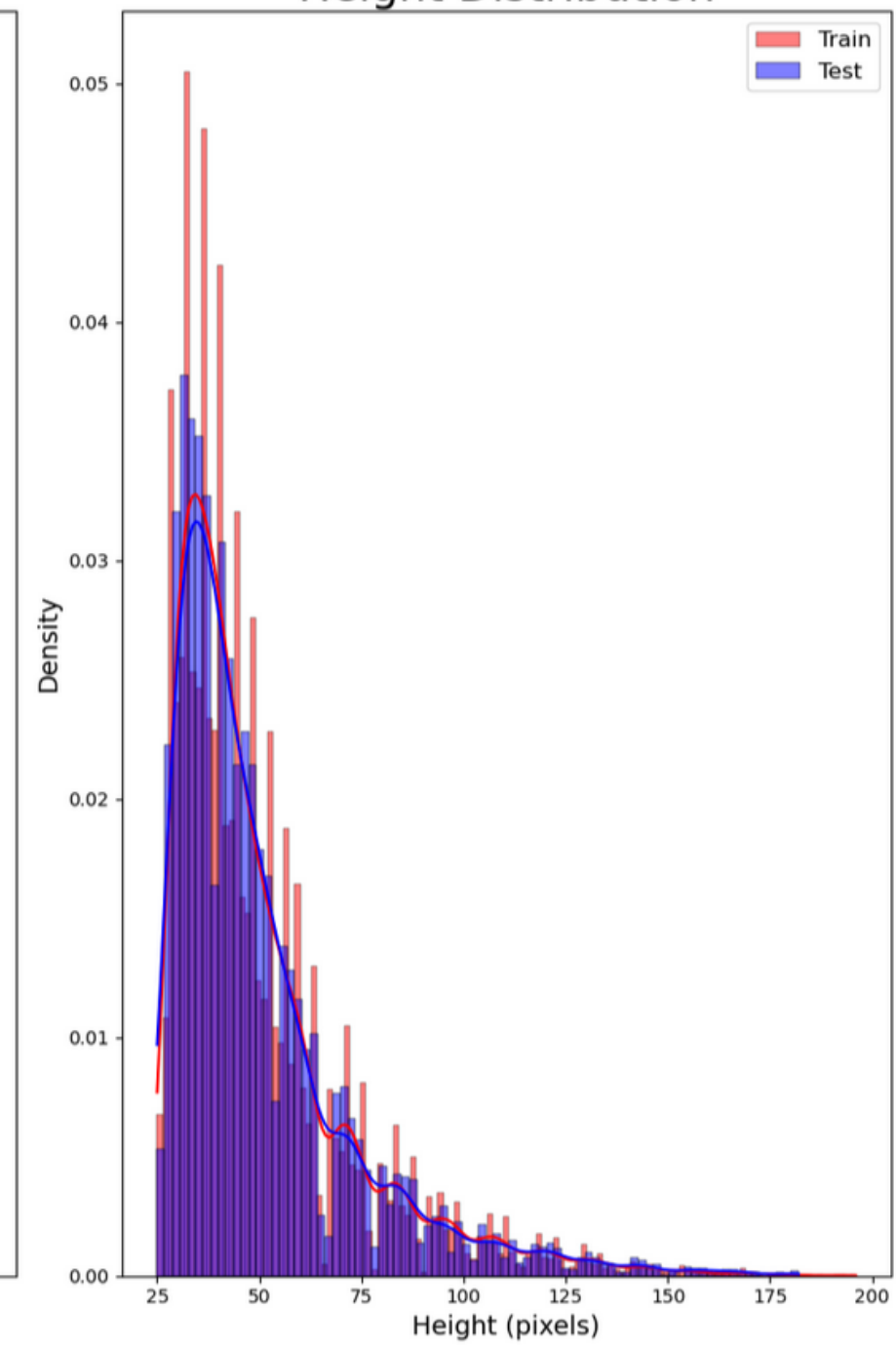
Train set:
Mean Width: 50.72
Mean Height: 50.23
Std Width: 23.93
Std Height: 22.80

Test set:
Mean Width: 50.33
Mean Height: 50.22
Std Width: 24.44
Std Height: 23.21

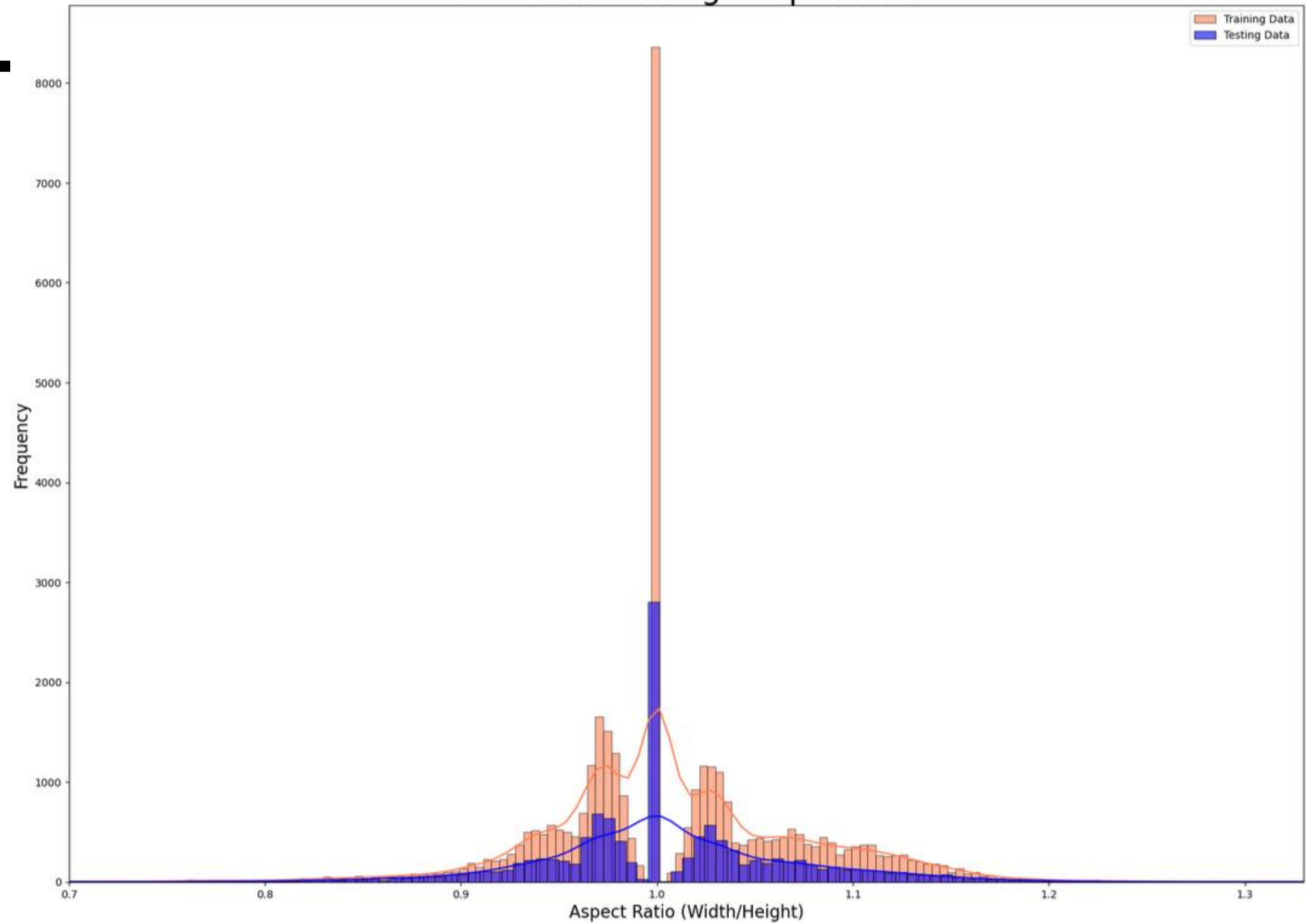
Image Size Distribution (Width vs Height)

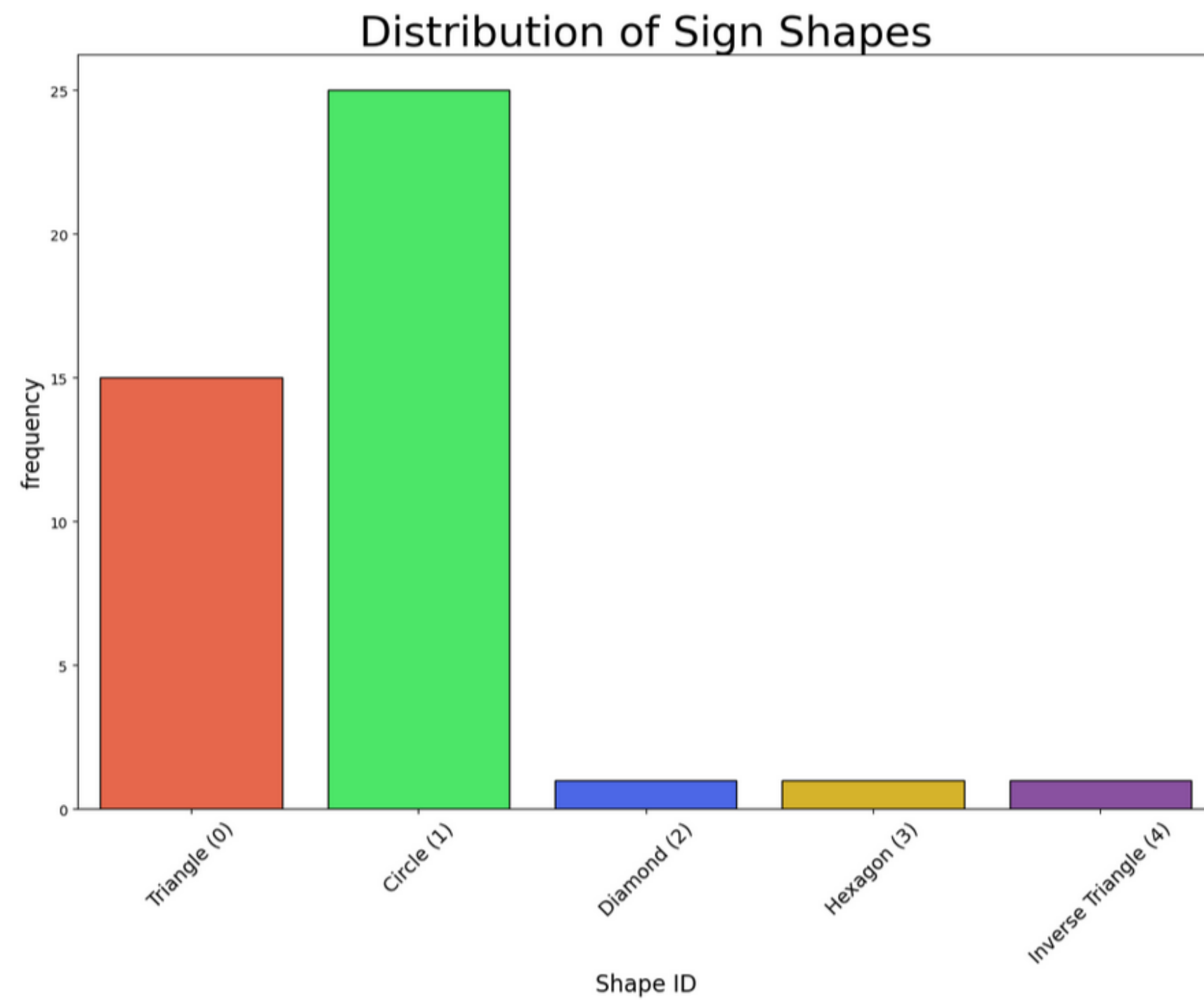
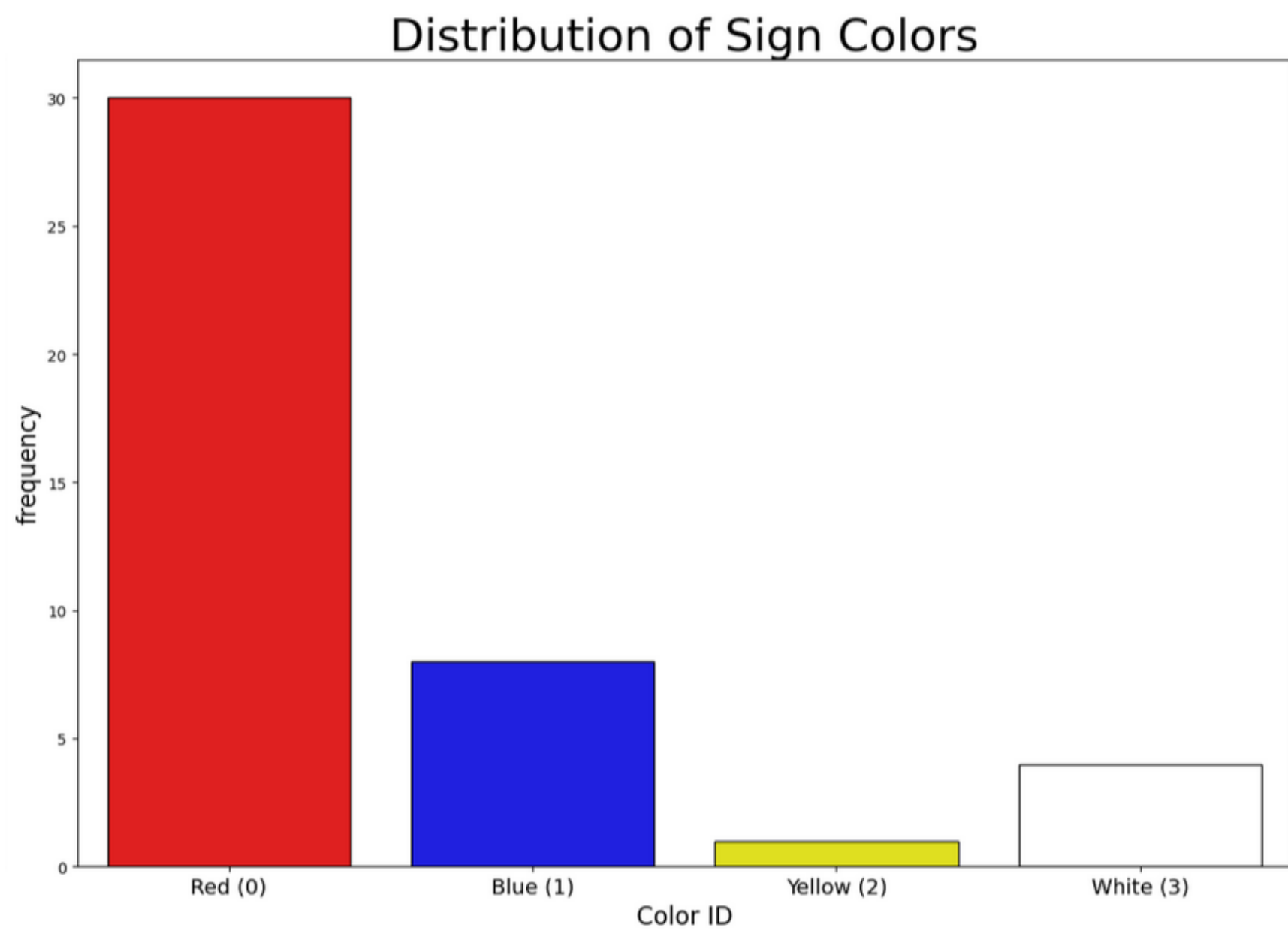


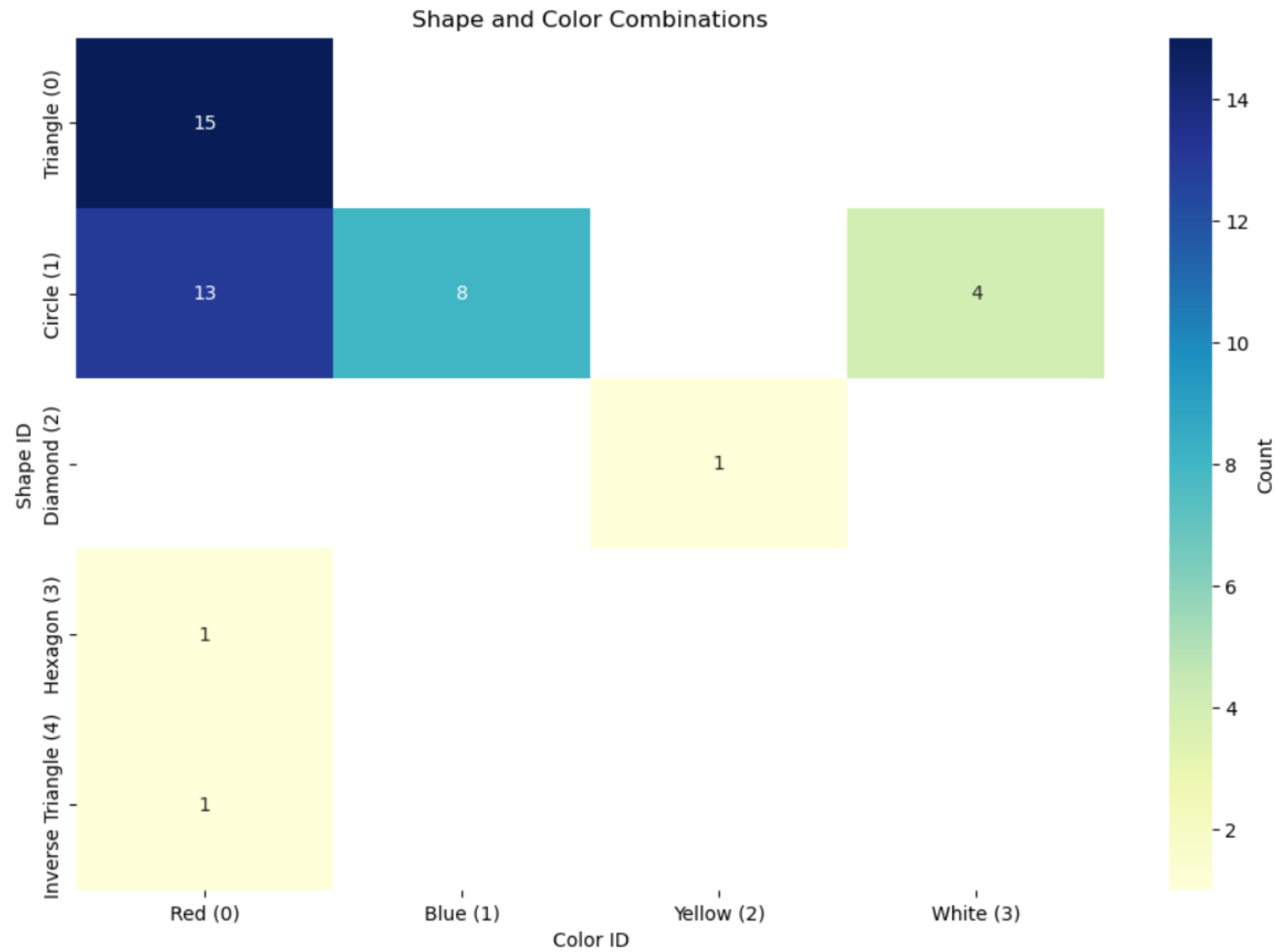
Height Distribution

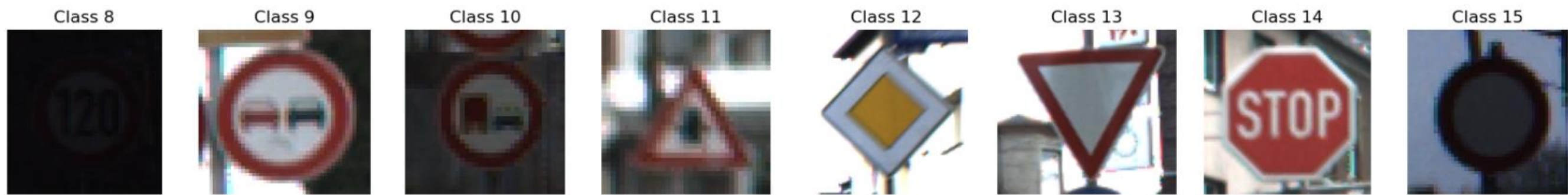


Distribution of Image Aspect Ratios









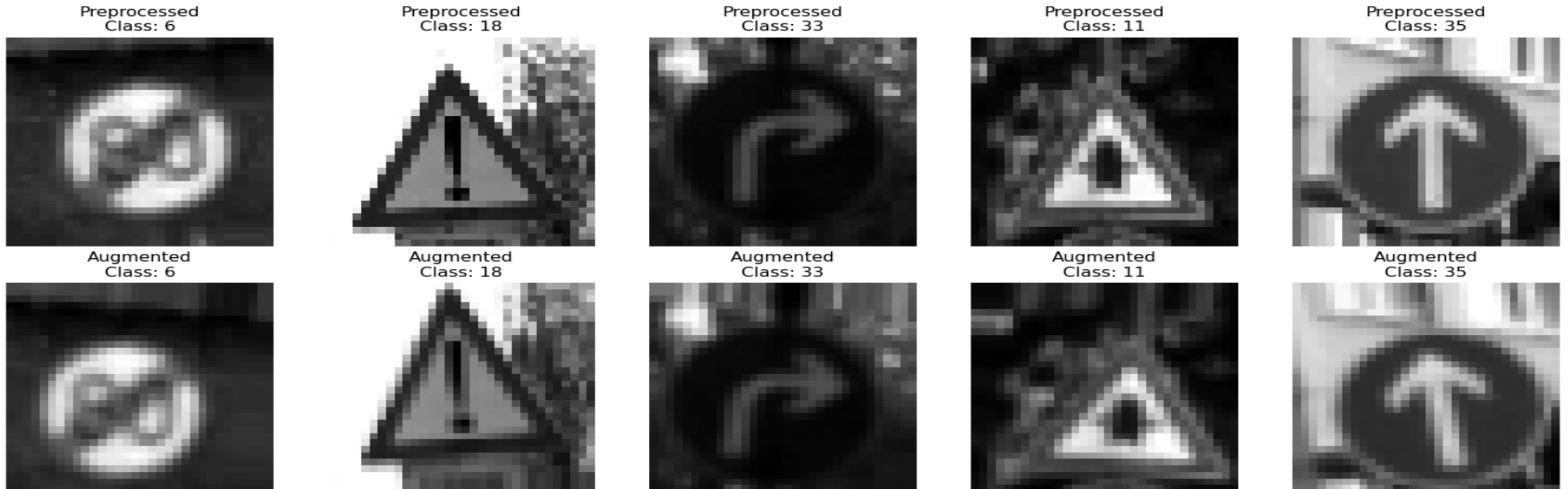


Data Augmentation Techniques

- Data augmentation is performed during training using Keras' ImageDataGenerator.
- Monochrome colorspace increases detection accuracy and improves model generalization.

Keras ImageDataGenerator:

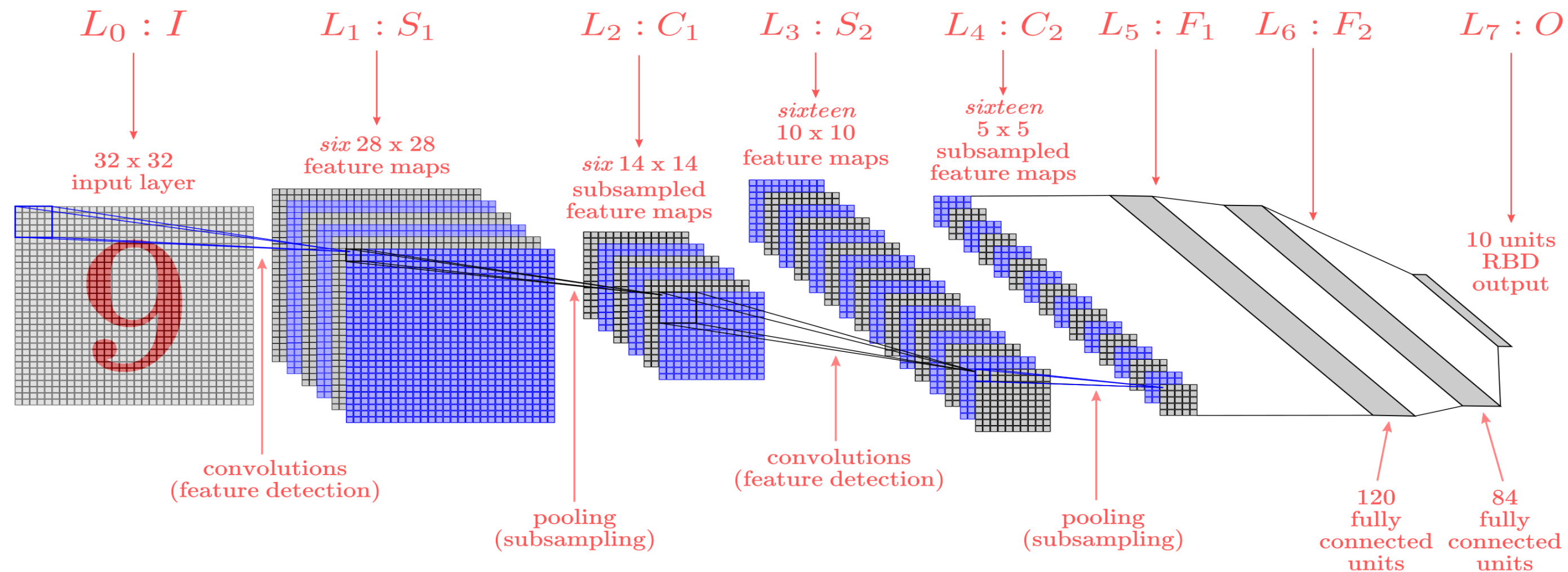
- Rotation: ± 10 degrees
- Width/Height Shift: $\pm 10\%$
 - Zoom: $\pm 10\%$
 - Shear: 0.1 radians



Preprocessing and data augmentation complete.
Training set shape: (31367, 32, 32, 1)
Validation set shape: (7842, 32, 32, 1)

LeNet-5 CNN Architecture

- Developed by Yann LeCun, Léon Bottou, Yoshua Bengio, and Patrick Haffner in 1998
- 2 convolutional layers, 2 subsampling layers, and 2 fully connected layers
- Designed for 32x32 pixel grayscale images
- LeNet-5 was revolutionary in using convolution for feature extraction. It demonstrated superior performance on the MNIST dataset, achieving a 0.95% error rate, a significant improvement over traditional methods.



Modified LeNet-5 CNN



Yann LeCun

@ylecun · 16h

I'm a scientist, not a business or product person.

108

43

1.3K

195K



Elon Musk

@elonmusk · 13h

What “science” have you done in the past 5 years?

342

280

2.5K

337K



Yann LeCun

@ylecun

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243

156

4.3K

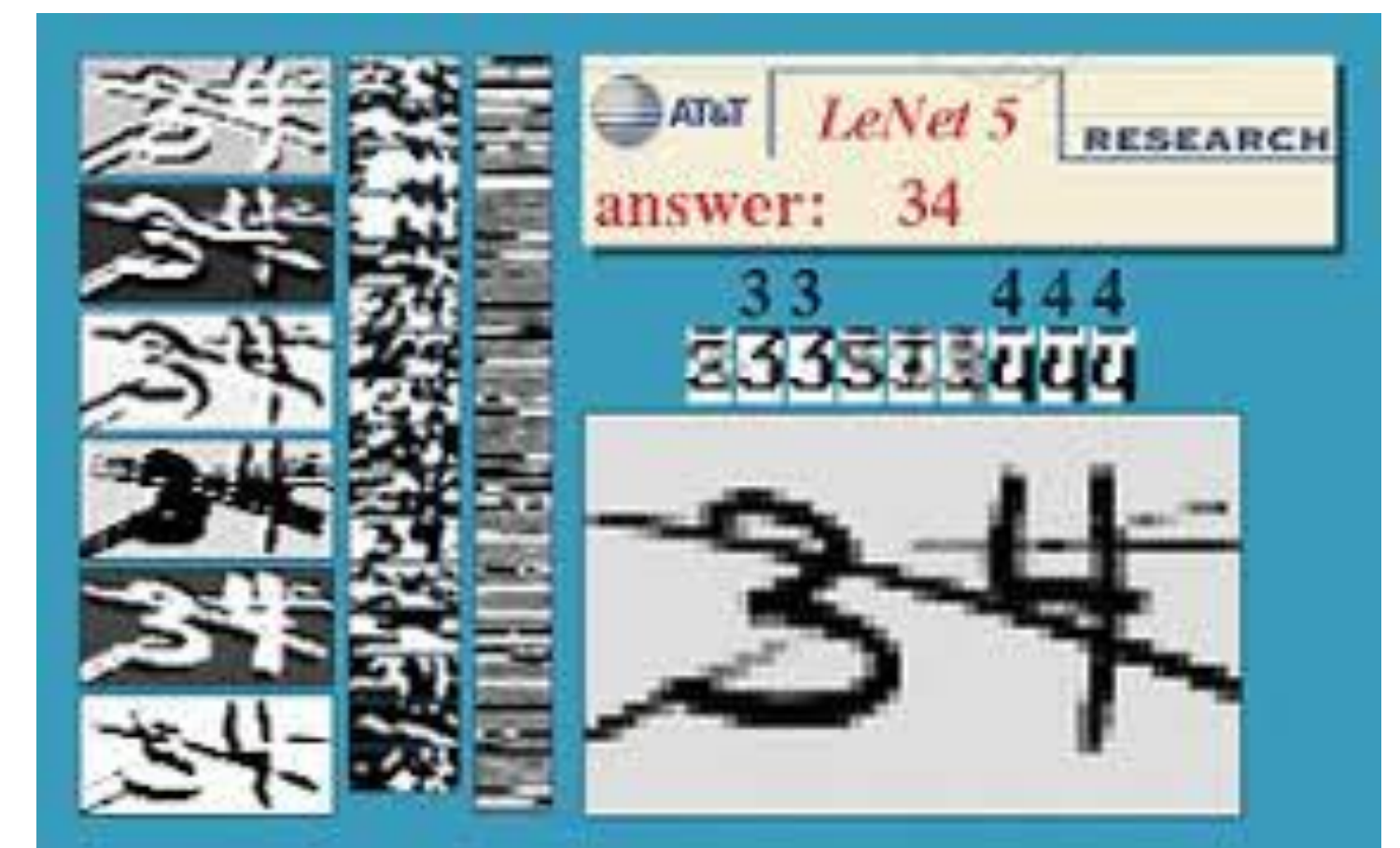
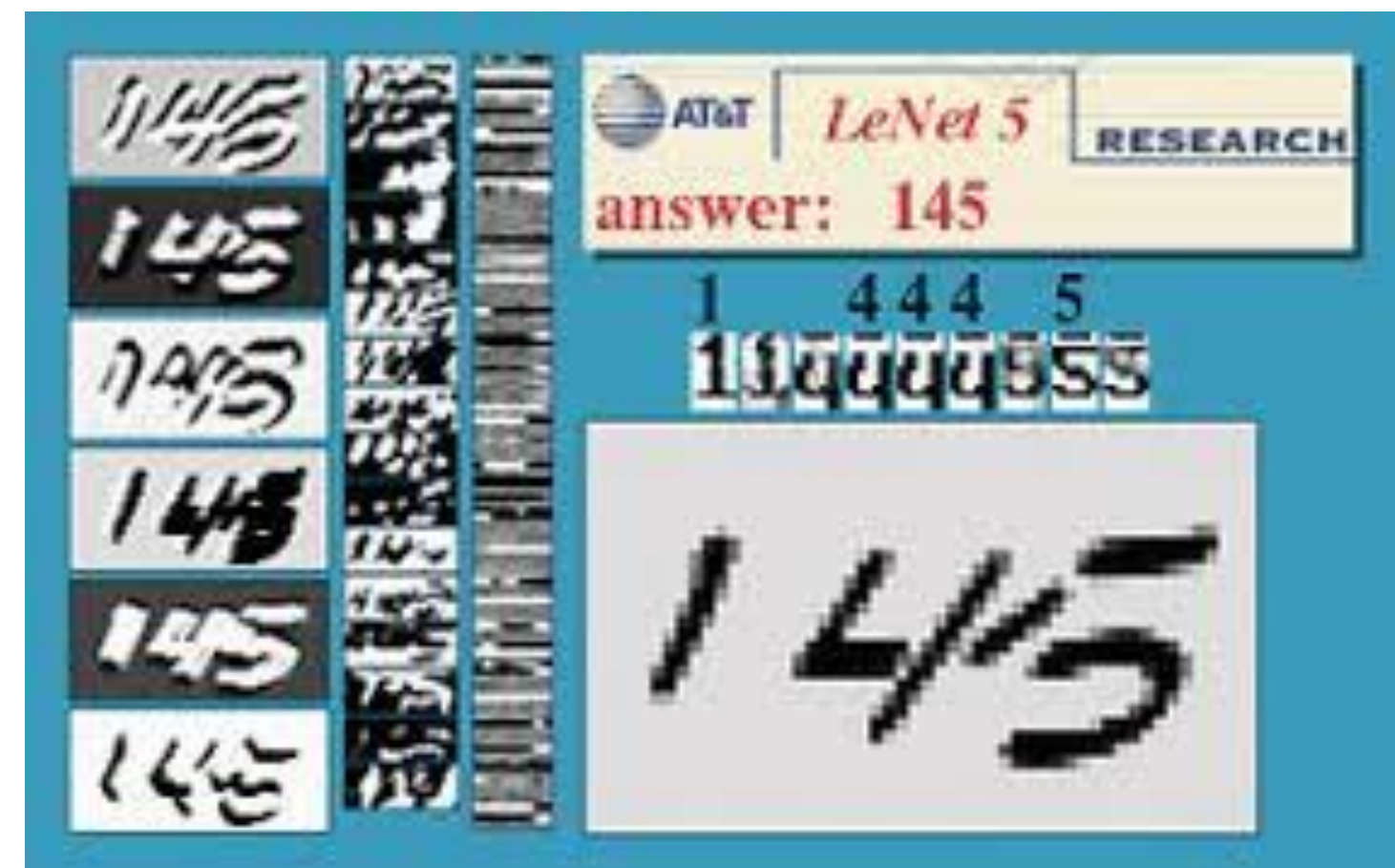
253

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- The original LeNet-5 had two convolutional layers; our modified version has three. This increased depth allows the network to learn more complex features specific to traffic signs.
- Increased the number of filters in the convolutional layers (32 and 64) compared to the original LeNet-5, allowing for more feature maps.
- A dropout layer with a rate of 0.5 was added before the final dense layer to prevent overfitting, a crucial consideration given our larger and more complex dataset.
- The final dense layer has 43 units, corresponding to the 43 classes in our traffic sign dataset, with a softmax activation for multi-class classification.
- Despite these modifications, our model remains relatively lightweight with only 124,139 parameters, making it suitable for real-time applications.

Forbidden Slide

Model: "sequential"

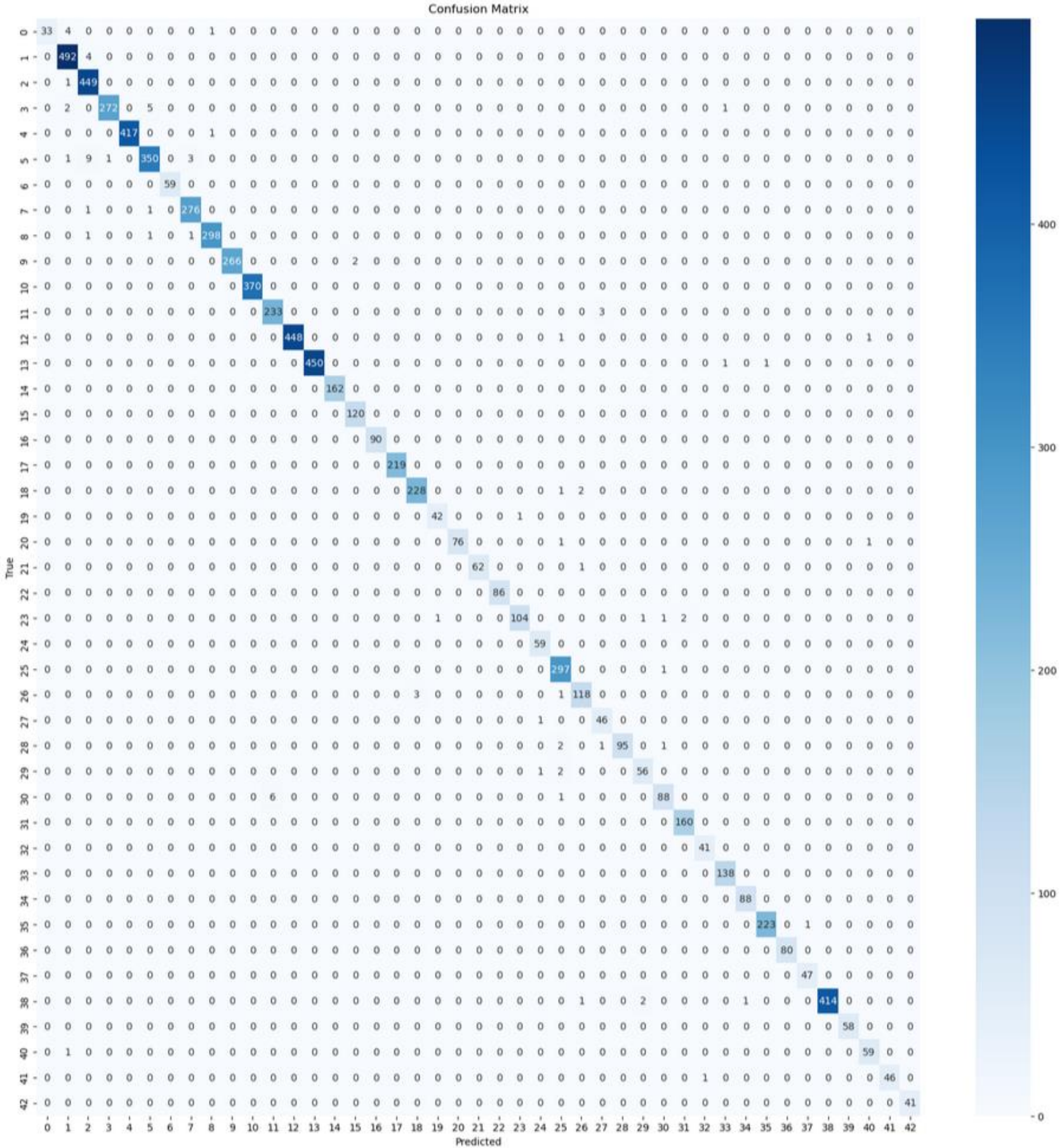
Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 30, 30, 32)	320
max_pooling2d (MaxPooling2D)	(None, 15, 15, 32)	0
conv2d_1 (Conv2D)	(None, 13, 13, 64)	18,496
max_pooling2d_1 (MaxPooling2D)	(None, 6, 6, 64)	0
conv2d_2 (Conv2D)	(None, 4, 4, 64)	36,928
flatten (Flatten)	(None, 1024)	0
dense (Dense)	(None, 64)	65,600
dropout (Dropout)	(None, 64)	0
dense_1 (Dense)	(None, 43)	2,795

Total params: 124,139 (484.92 KB)

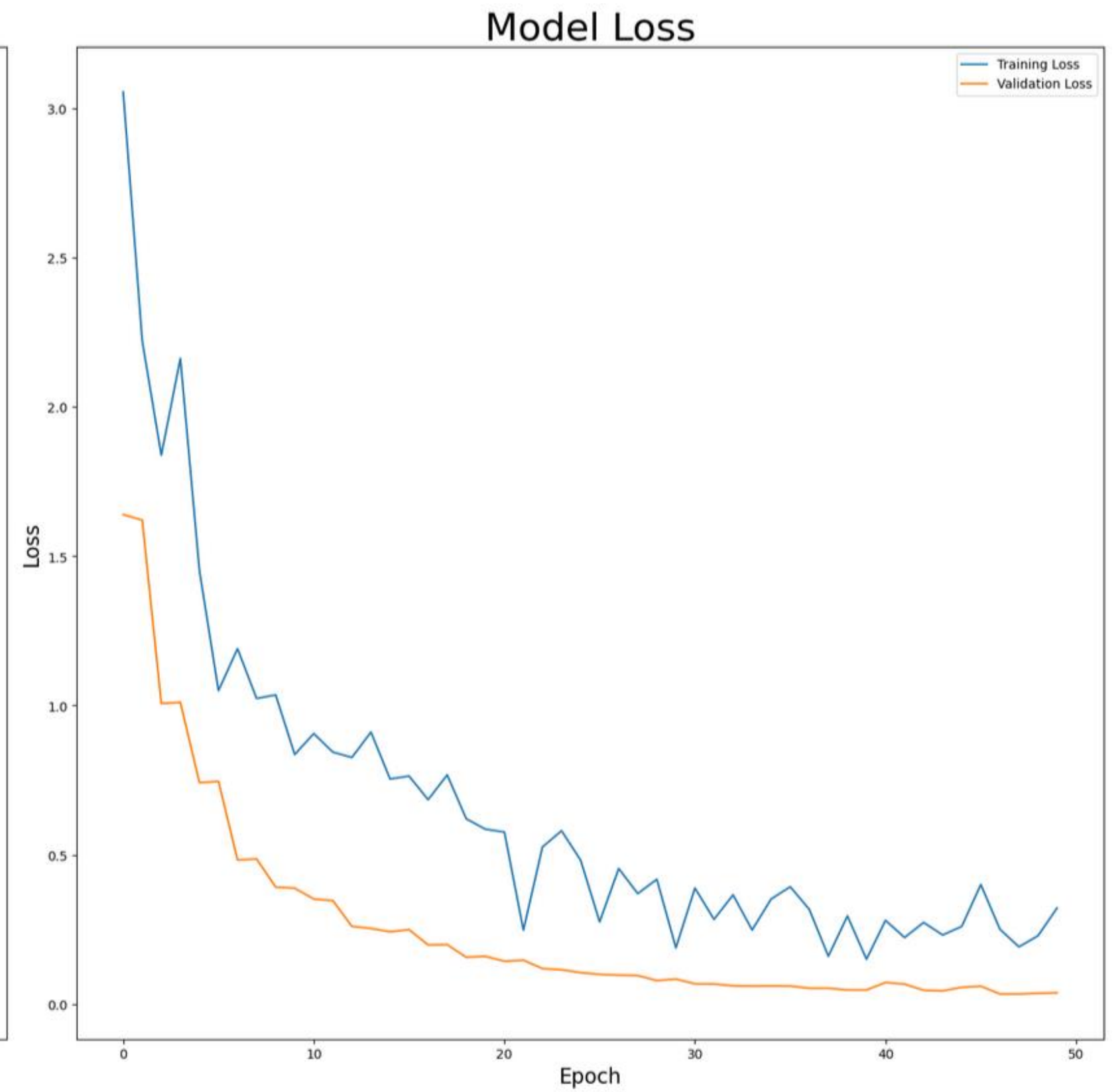
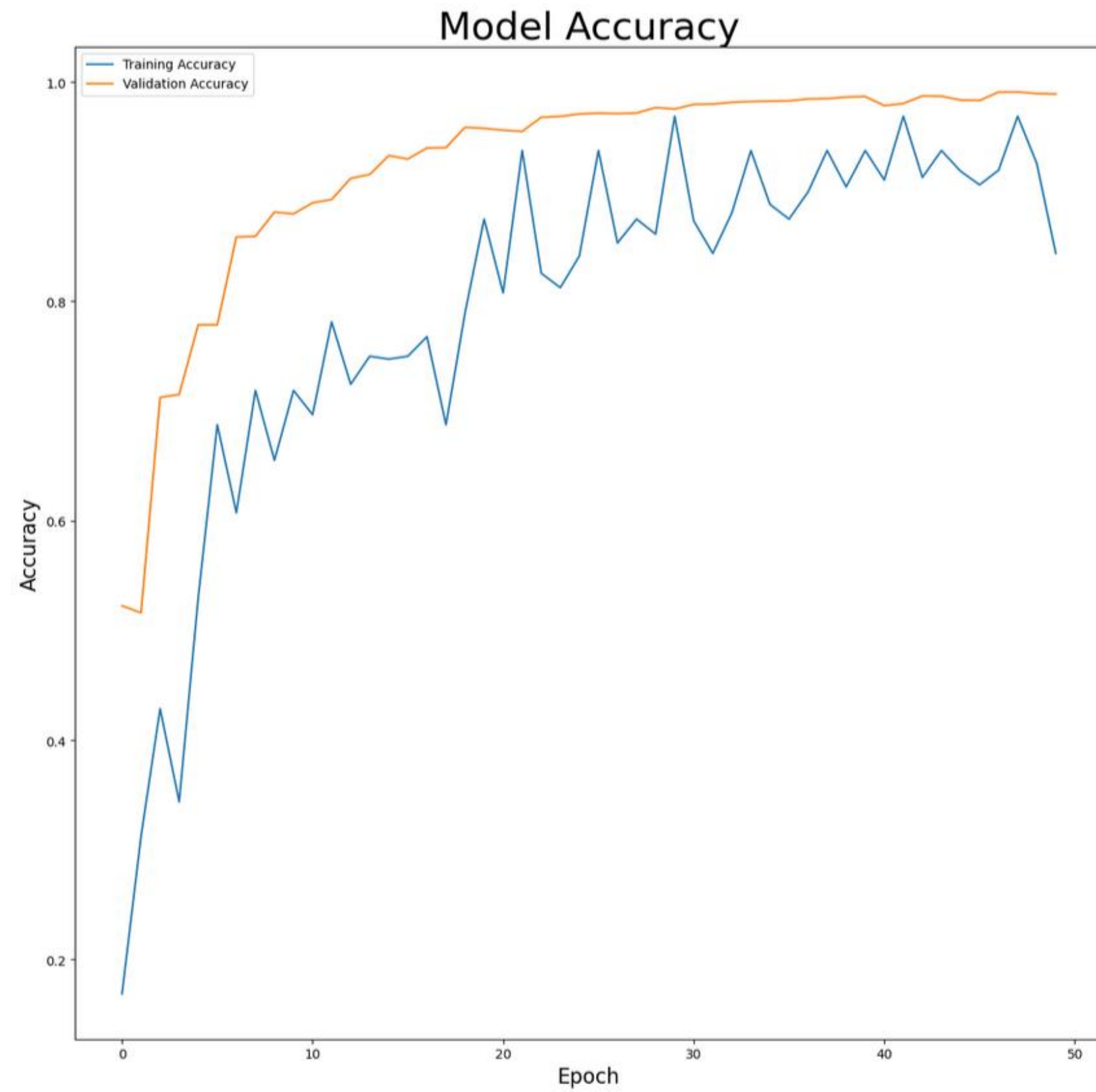
Trainable params: 124,139 (484.92 KB)

Non-trainable params: 0 (0.00 B)

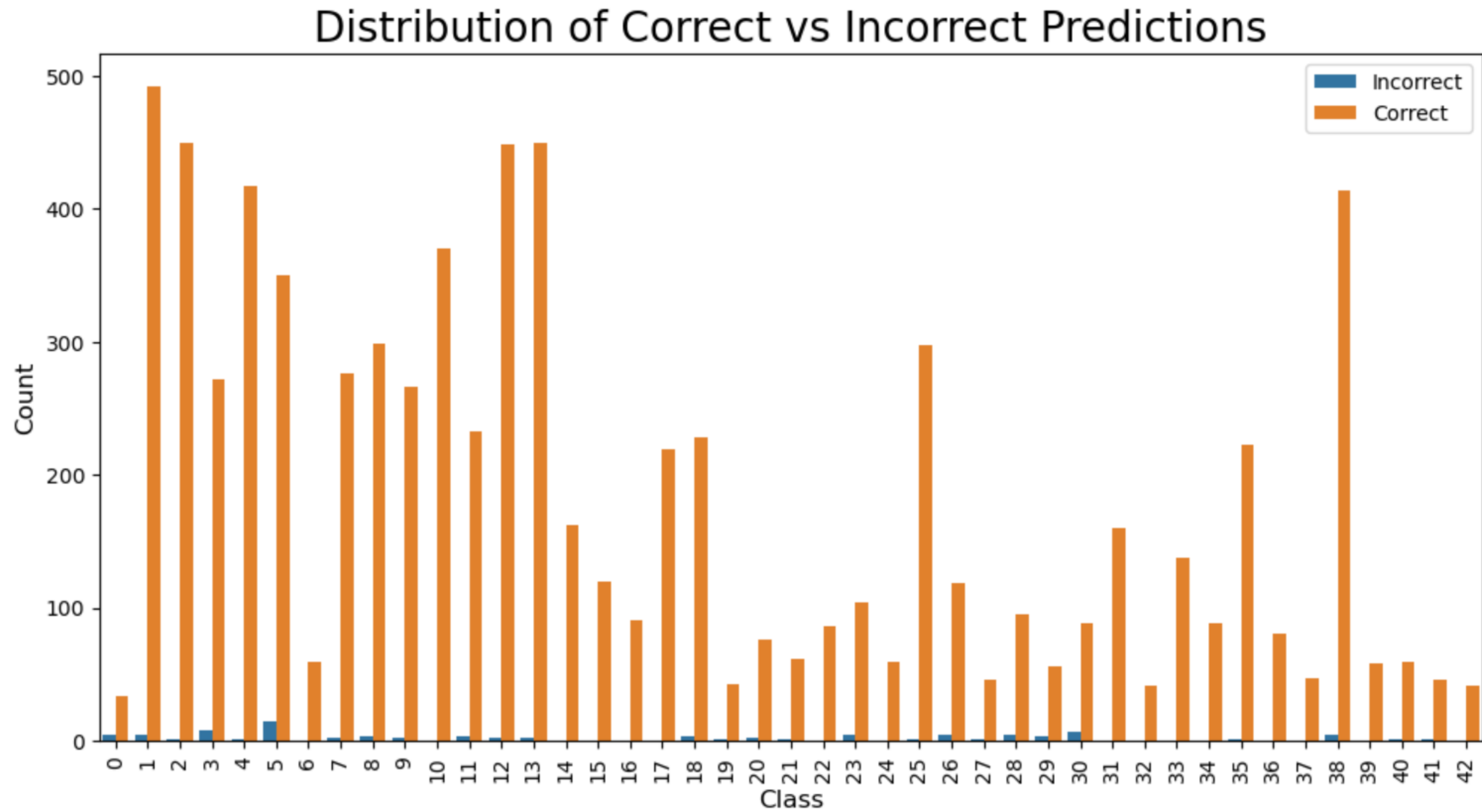
Epoch 45/50	980/980	26s	27ms/step	- accuracy: 0.9202	- loss: 0.2591	- val_accuracy: 0.9836	- val_loss: 0.0563
Epoch 46/50	980/980	1s	1ms/step	- accuracy: 0.9062	- loss: 0.4002	- val_accuracy: 0.9832	- val_loss: 0.0599
Epoch 47/50	980/980	26s	26ms/step	- accuracy: 0.9181	- loss: 0.2547	- val_accuracy: 0.9908	- val_loss: 0.0342
Epoch 48/50	980/980	1s	1ms/step	- accuracy: 0.9688	- loss: 0.1918	- val_accuracy: 0.9909	- val_loss: 0.0341
Epoch 49/50	980/980	26s	27ms/step	- accuracy: 0.9256	- loss: 0.2280	- val_accuracy: 0.9894	- val_loss: 0.0364
Epoch 50/50	980/980	1s	1ms/step	- accuracy: 0.8438	- loss: 0.3218	- val_accuracy: 0.9890	- val_loss: 0.0377



Plotting Model Metrics



Visualize Confusion Matrix



Similar Signs?



(1)

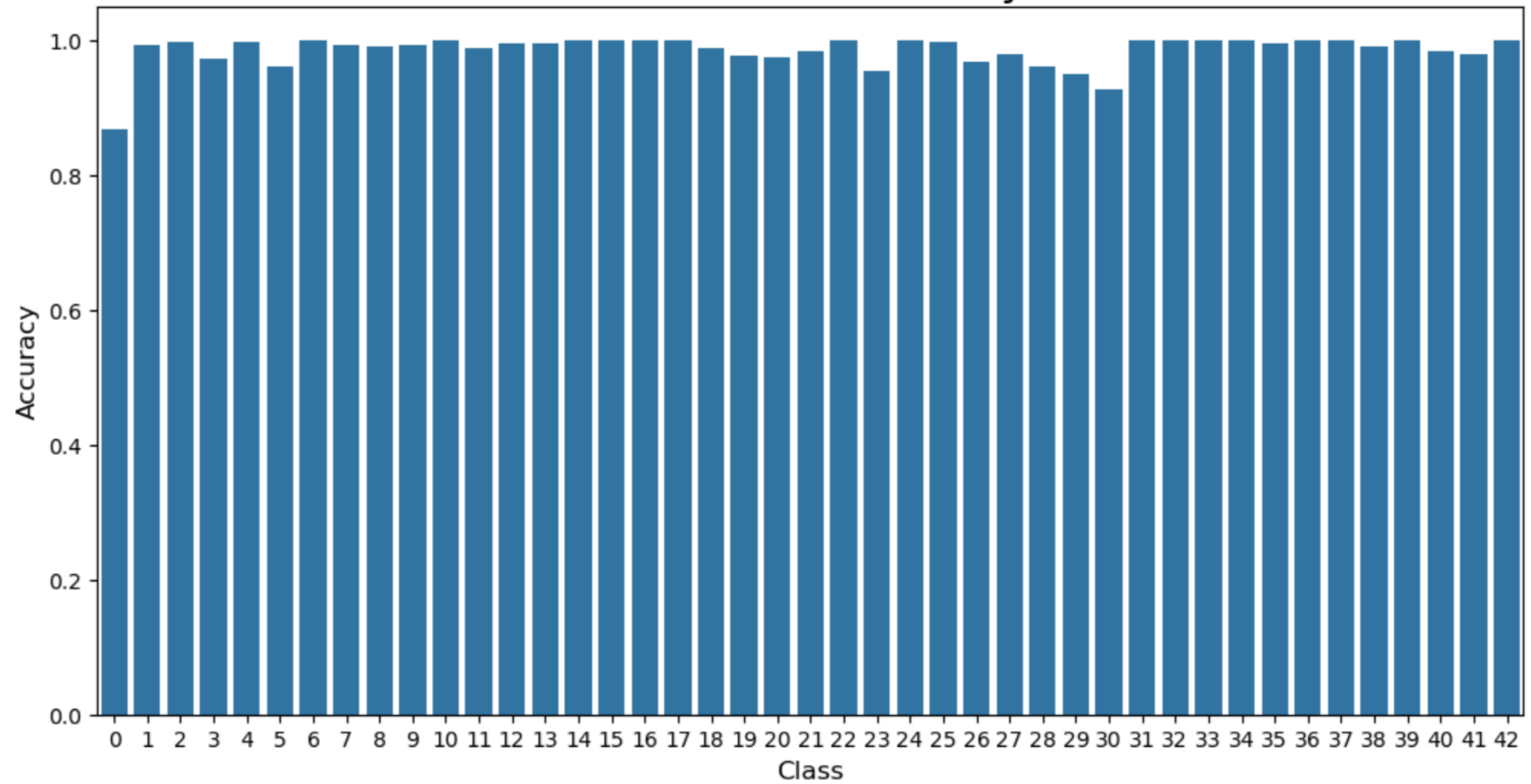
(0)



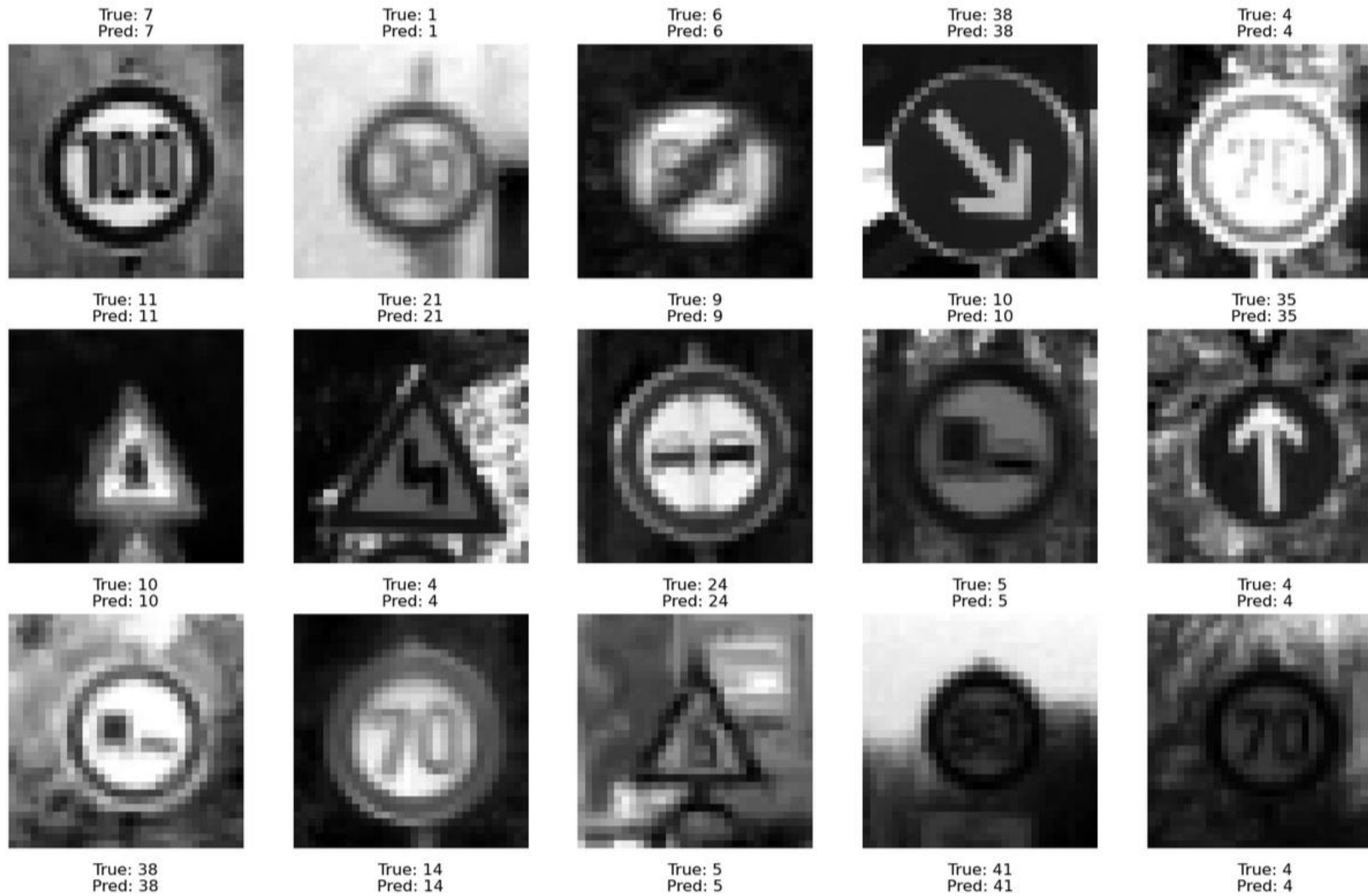
(18)

(26)

Class-wise Accuracy



Actual vs Predicted Class ID



Future Use Cases

- Transfer Learning: Utilize EfficientNetB0 or MobileNetV2, pre-trained on ImageNet
- GPU Acceleration: Implement mixed precision training for 2-3x speedup
- YOLO v8 Integration: For real-time sign detection (expected 45 FPS on GPU)
- Significantly enhances both the accuracy and real-world applicability of the model.



Demonstration Overview

Streamlit App Features:

- Image upload and real-time classification
- Top-5 predictions with confidence scores

Webcam Capture:

- Continuous prediction on video stream
- Threshold-based classification display

- Streamlit app provides an intuitive interface for model interaction.
- The webcam feature, while sometimes inconsistent due to lighting and angle variations, demonstrates the potential for real-time applications.



Thanks!

Do you have any questions?
