

# Autonomous Vehicle AI Analysis

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## Project Background:

Using Machine Learning and Artificial Intelligence to classify a large traffic sign dataset using a convolutional neural network (CNN)

Adversarial: manipulation of images within dataset to test vulnerabilities of machine learning and our neural network model

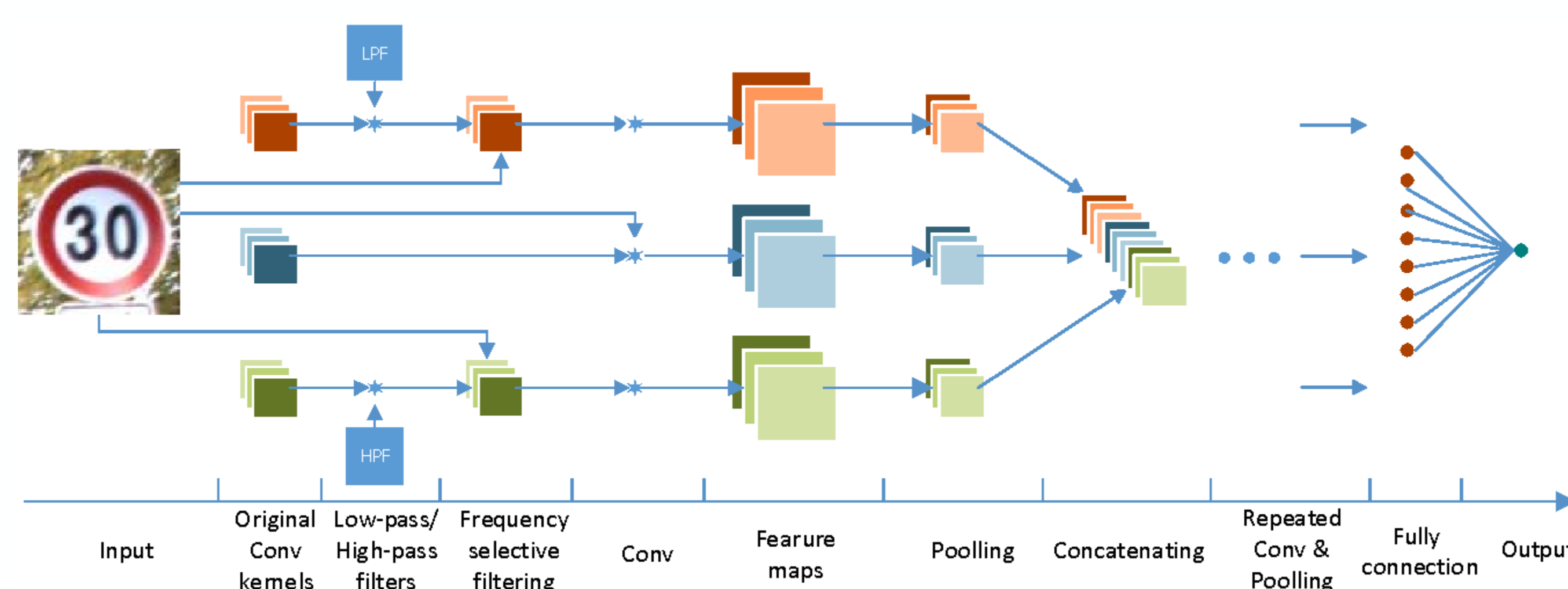


Figure 1. Simple Image showing “under the hood” structure of CNN

## Sample Images from the dataset:



(Graphs below showing the varied class set)  
Number of different classes: 47

## Data Processing:

Over 7000 raw images varying in size  
Grayscaled to reduce channels  
Images were upscaled by padding or downscaled

## Creating the Neural Network:

```
def model_fn(features, labels, mode, params):  
    5 layers (Convolutional, Flattening, Pooling, Fully  
    Connected)  
    Weights, biases, and filters to create the feature  
    maps  
  
    # First convolutional layer.  
    net = tf.layers.conv2d(inputs=x, name='layer_conv1',  
                           filters=32, kernel_size=3,  
                           padding='same', activation=tf.nn.relu)  
    net = tf.layers.max_pooling2d(inputs=net, pool_size=2, strides=2)  
  
    # Second convolutional layer.  
    net = tf.layers.conv2d(inputs=net, name='layer_conv2',  
                           filters=32, kernel_size=3,  
                           padding='same', activation=tf.nn.relu)  
    net = tf.layers.max_pooling2d(inputs=net, pool_size=2, strides=2)  
  
    # First fully-connected / dense layer.  
    # This uses the ReLU activation function.  
    net = tf.layers.dense(inputs=net, name='layer_fc1',  
                          units=128, activation=tf.nn.relu)  
  
    # Second fully-connected / dense layer.  
    # This is the last layer so it does not use an activation function.  
    net = tf.layers.dense(inputs=net, name='layer_fc_2',  
                          units=2)
```

## Key features of CNN:

Dropout: Randomly ignoring neuron nodes during the training phase to prevent overfitting in the fully connected layer

## Training and Testing Neural Network:

# of Images in Training: \_\_\_\_\_

# of Images in Testing: \_\_\_\_\_

(Place output code below showing accuracy)

## Adversarial Phase:

```
def add_noise(X_image)  
    #create copy as to not modify the actual image  
    X_image_copy = X_image.copy()  
    height,width,channel = X_image_copy.shape #gather dimensions and channel number for shape  
    mean = 0  
    stdev = 0.4  
    sigma = stdev**0.5  
    generated_noise = np.random.normal(mean, sigma, (height,width,channel))  
    #gaussian noise is a statistical method used to generate noise (distribution curve)  
    gaussian_noise = generated_noise.reshape(height,width,channel)  
    image_with_noise = X_image_copy + gaussian_noise  
  
    return image_with_noise
```

## Gaussian Noise:

Probability density function to generate the noise within the images  
(Sample picture(s) with noise added)

## Results:

(show result when fed back into CNN for testing)  
(Another histogram plotting the results)