Traffic_Sign

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```
Batch: G-5 (2017-21)
      Import libraries
[1]: import numpy as np
    import matplotlib.pyplot as plt
    import keras
    import cv2
    from keras.models import Sequential
    from keras.optimizers import Adam
    from keras.layers import Dense
    from keras.utils.np_utils import to_categorical
    from keras.layers import Dropout, Flatten
    from keras.layers.convolutional import Conv2D, MaxPooling2D
    import pickle
    import random
    import pandas as pd
      Import Dataset
[2]: !git clone https://bitbucket.org/jadslim/german-traffic-signs
   Cloning into 'german-traffic-signs'...
   remote: Counting objects: 6, done.
   remote: Compressing objects: 100% (6/6), done.
   remote: Total 6 (delta 0), reused 0 (delta 0)
   Unpacking objects: 100% (6/6), done.
[3]: with open('german-traffic-signs/train.p','rb') as f:
        train_data = pickle.load(f)
    with open('german-traffic-signs/valid.p','rb') as f:
        val_data = pickle.load(f)
    with open('german-traffic-signs/test.p','rb') as f:
        test_data = pickle.load(f)
[4]: X_train, y_train = train_data['features'], train_data['labels']
    X_val, y_val = val_data['features'], val_data['labels']
    X_test, y_test = test_data['features'], test_data['labels']
```

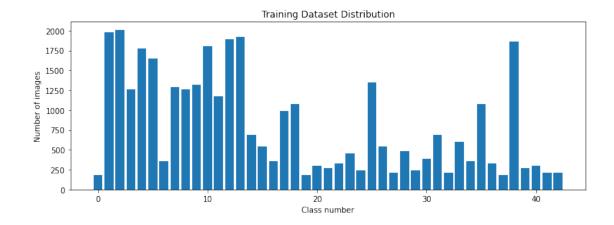
```
[5]: print(X_train.shape)
    print(X_val.shape)
    print(X_test.shape)
   (34799, 32, 32, 3)
   (4410, 32, 32, 3)
   (12630, 32, 32, 3)
      Data Visulisation
[6]: data = pd.read_csv('german-traffic-signs/signnames.csv')
    num_of_samples = []
    cols = 5
    num_classes = 43
    fig, axs = plt.subplots(nrows = num_classes, ncols = cols, figsize = (5, 50))
    fig.tight_layout()
    for i in range(cols):
        for j, row in data.iterrows():
            x_selected = X_train[y_train == j]
            axs[j][i].imshow(x_selected[random.randint(0, (len(x_selected)-1)), :, :
     →], cmap = plt.get_cmap("gray"))
            axs[j][i].axis("off")
            if i == 2:
                axs[j][i].set_title(str(j) + "_" + row["SignName"])
                num_of_samples.append(len(x_selected))
```



```
[7]: # Distribution of classes
print(num_of_samples)
plt.figure(figsize = (12, 4))
plt.bar(range(0, num_classes), num_of_samples)
plt.title("Training Dataset Distribution")
plt.xlabel("Class number")
plt.ylabel("Number of images")
```

[180, 1980, 2010, 1260, 1770, 1650, 360, 1290, 1260, 1320, 1800, 1170, 1890, 1920, 690, 540, 360, 990, 1080, 180, 300, 270, 330, 450, 240, 1350, 540, 210, 480, 240, 390, 690, 210, 599, 360, 1080, 330, 180, 1860, 270, 300, 210, 210]

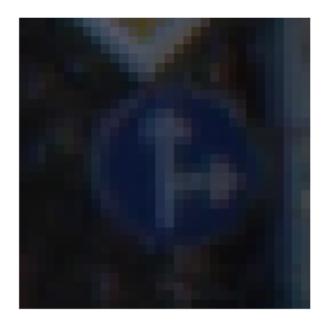
[7]: Text(0, 0.5, 'Number of images')



Data Augmentation for making images huges or add 42/43 more images

```
[8]: plt.imshow(X_train[1000])
   plt.axis('off')
   print(X_train[1000].shape)
   print(y_train[1000])
```

(32, 32, 3) 36



Pre-processing has to be done

```
[9]: # Color does not matters here shapes and sign matters
def grayscale(img):
    image = cv2.cvtColor(img, cv2.COLOR_BGR2GRAY)
    plt.axis('off')
    return image

[10]: img = grayscale(X_train[1000])
    plt.imshow(img, cmap = 'gray')
    print(img.shape)
```

(32, 32)



```
[11]: # Histogram equalisation
    def equalize(img):
        img = cv2.equalizeHist(img)
        return img

[12]: img = equalize(img)
    plt.imshow(img, cmap = 'gray')
    plt.axis('off')
    print(img.shape)
```

(32, 32)



```
[13]: # Function to preprocess the images in bulk and to normalise it's values
def preprocessing(img):
    img = grayscale(img)
    img = equalize(img)
    img = img/255
    return img

[14]: X_train = np.array(list(map(preprocessing, X_train)))
    X_val = np.array(list(map(preprocessing, X_val)))
    X_test = np.array(list(map(preprocessing, X_test)))
```

```
X_{val} = X_{val.reshape}(4410, 32, 32, 1)
    X_test = X_test.reshape(12630, 32, 32, 1)
[16]: from keras.preprocessing.image import ImageDataGenerator
     # Increasing the zoom range height range and width range
    datagen = ImageDataGenerator(width_shift_range = 0.1,
                      height_shift_range = 0.1,
                       zoom_range = 0.2,
                      shear_range = 0.1,
                      rotation_range = 10)
    datagen.fit(X_train)
[17]: # If you directly pass 22 the memory gets occupied completely
    batches = datagen.flow(X_train, y_train, batch_size = 20)
    X_batch, y_batch = next(batches)
    fig, axs = plt.subplots(1, 15, figsize = (20, 5))
    fig.tight_layout()
    for i in range(15):
      axs[i].imshow(X_batch[i].reshape(32, 32))
      axs[i].axis('off')
                       [18]: y_train = to_categorical(y_train, 43)
    y_val = to_categorical(y_val, 43)
    y_test = to_categorical(y_test, 43)
[19]: def neural_model():
        model = Sequential()
        model.add(Conv2D(60, (5, 5), input_shape = (32, 32, 1), activation = (32, 32, 1)

¬'relu'))
        model.add(Conv2D(60, (5, 5), input\_shape = (32, 32, 1), activation = __

¬'relu'))
        model.add(MaxPooling2D(pool_size = (2,2)))
        model.add(Conv2D(30, (3, 3), activation = 'relu'))
        model.add(Conv2D(30, (3, 3), activation = 'relu'))
```

[15]: X_train = X_train.reshape(34799, 32, 32, 1) # 1- Section height weight

```
model.add(MaxPooling2D(pool_size = (2, 2)))
         \#model.add(Dropout(0.5))
         model.add(Flatten())
         model.add(Dense(500, activation = 'relu'))
         model.add(Dropout(0.5))
         model.add(Dense(num_classes, activation = 'softmax'))
         # Learning rate is gradient descent
        model.compile(Adam(lr = 0.001), loss = 'categorical_crossentropy', metrics_
      →= ['accuracy'])
         return model
[20]: model = neural_model()
     print(model.summary())
```

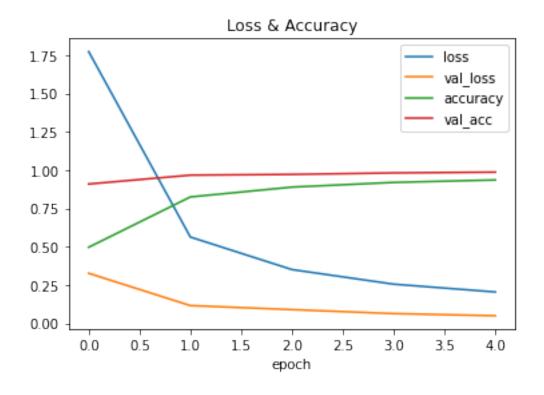
Model: "sequential"

Layer (type)	-		Param #
conv2d (Conv2D)		28, 28, 60)	1560
conv2d_1 (Conv2D)	(None,	24, 24, 60)	90060
max_pooling2d (MaxPooling2D)	(None,	12, 12, 60)	0
conv2d_2 (Conv2D)	(None,	10, 10, 30)	16230
conv2d_3 (Conv2D)	(None,	8, 8, 30)	8130
max_pooling2d_1 (MaxPooling2	(None,	4, 4, 30)	0
flatten (Flatten)	(None,	480)	0
dense (Dense)	(None,	500)	240500
dropout (Dropout)	(None,	500)	0
dense_1 (Dense)	(None,	43)	21543
Total params: 378,023 Trainable params: 378,023 Non-trainable params: 0			
None			

Training the model

```
[21]: history = model.fit_generator(datagen.flow(X_train, y_train, batch_size = 50),__
     ⇒epochs = 5, validation data =(X val, y val), shuffle = 1)
    /usr/local/lib/python3.7/dist-
    packages/tensorflow/python/keras/engine/training.py:1844: UserWarning:
    `Model.fit generator` is deprecated and will be removed in a future version.
    Please use `Model.fit`, which supports generators.
      warnings.warn('`Model.fit_generator` is deprecated and '
    Epoch 1/5
    696/696 [============= ] - 46s 19ms/step - loss: 2.6127 -
    accuracy: 0.2875 - val loss: 0.3267 - val accuracy: 0.9095
    696/696 [============ ] - 12s 18ms/step - loss: 0.6485 -
    accuracy: 0.7976 - val_loss: 0.1158 - val_accuracy: 0.9673
    696/696 [============ ] - 13s 18ms/step - loss: 0.3807 -
    accuracy: 0.8780 - val_loss: 0.0894 - val_accuracy: 0.9723
    696/696 [============ ] - 13s 18ms/step - loss: 0.2737 -
    accuracy: 0.9130 - val_loss: 0.0634 - val_accuracy: 0.9819
    Epoch 5/5
    696/696 [============= ] - 13s 18ms/step - loss: 0.2106 -
    accuracy: 0.9326 - val_loss: 0.0495 - val_accuracy: 0.9871
      Evaluating the model
[22]: plt.plot(history.history['loss'])
    plt.plot(history.history['val loss'])
    plt.plot(history.history['accuracy'])
    plt.plot(history.history['val_accuracy'])
    plt.legend(['loss', 'val_loss', 'accuracy', 'val_acc'])
    plt.title('Loss & Accuracy')
    plt.xlabel('epoch')
```

[22]: Text(0.5, 0, 'epoch')



[23]: score = model.evaluate(X_test, y_test, verbose = 1)



Preprocessing images

```
[25]: img = np.asarray(image)
  img = cv2.resize(img, (32, 32))
  img = preprocessing(img)
  plt.imshow(img, cmap = plt.get_cmap('gray'))
  print(img.shape)
  img = img.reshape(1, 32, 32, 1) #1 image 32 height and width and 1 channel
```

(32, 32)



```
[26]: prediction = str(model.predict_classes(img))
    prediction = prediction[1:-1]
##print("predicted sign: "+ prediction )
```

/usr/local/lib/python3.7/distpackages/tensorflow/python/keras/engine/sequential.py:450: UserWarning:
`model.predict_classes()` is deprecated and will be removed after 2021-01-01.
Please use instead:* `np.argmax(model.predict(x), axis=-1)`, if your model does multi-class classification (e.g. if it uses a `softmax` last-layer activation).* `(model.predict(x) > 0.5).astype("int32")`, if your model does binary classification (e.g. if it uses a `sigmoid` last-layer activation).

warnings.warn('`model.predict_classes()` is deprecated and '

```
[27]: pred = int(prediction)
plt.imshow(image)
plt.axis('off')

for num, name in data.iteritems():
   name = name.values
   print("predicted sign: "+ str(name[pred]))
```

predicted sign: 34

predicted sign: Turn left ahead



```
[]: !wget -nc https://raw.githubusercontent.com/brpy/colab-pdf/master/colab_pdf.py
   from colab_pdf import colab_pdf
   colab_pdf('Traffic_Sign.ipynb')
  --2021-04-17 06:48:44-- https://raw.githubusercontent.com/brpy/colab-
  pdf/master/colab_pdf.py
  Resolving raw.githubusercontent.com (raw.githubusercontent.com)...
  185.199.108.133, 185.199.109.133, 185.199.110.133, ...
  Connecting to raw.githubusercontent.com
  (raw.githubusercontent.com) | 185.199.108.133 | :443... connected.
  HTTP request sent, awaiting response... 200 OK
  Length: 1864 (1.8K) [text/plain]
  Saving to: colab_pdf.py
  colab_pdf.py
                     in Os
  2021-04-17 06:48:44 (44.8 MB/s) - colab_pdf.py saved [1864/1864]
  Mounted at /content/drive/
  WARNING: apt does not have a stable CLI interface. Use with caution in scripts.
  WARNING: apt does not have a stable CLI interface. Use with caution in scripts.
```

Extracting templates from packages: 100%