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In [ ]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import tensorflow as tf
from PIL import Image
import os
from sklearn.model_selection import train_test_split
from keras.utils import to_categorical
from keras.models import Sequential
from keras.layers import Conv2D, MaxPool2D, Dense, Flatten, Dropout
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In [ ]: data = []
labels = []
classes = 43
cur_path = os.getcwd()
#Retrieving the images and their labels
for i in range(classes):
    path = os.path.join(cur_path, 'train', str(i))
    images = os.listdir(path)
    for a in images:
        try:
            image = Image.open(path+"/"+str(a))
            image = image.resize((30,30))
            image = np.array(image)
            #sim = Image.fromarray(image)
            data.append(image)
            labels.append(i)
        except:
            print("Error loading image")
```

```
In [ ]: #Converting lists into numpy arrays
data = np.array(data)
labels = np.array(labels)
print(data.shape, labels.shape)
#Splitting training and testing dataset
X_t1, X_t2, y_t1, y_t2 = train_test_split(data, labels, test_size=0.2, random_state=42)
print(X_t1.shape, X_t2.shape, y_t1.shape, y_t2.shape)
#Converting the labels into one hot encoding
y_t1 = to_categorical(y_t1, 43)
y_t2 = to_categorical(y_t2, 43)

(39209, 30, 30, 3) (39209,)
(31367, 30, 30, 3) (7842, 30, 30, 3) (31367,) (7842,)
```

```
In [ ]: #Building the model
model = Sequential()
model.add(Conv2D(filters=32, kernel_size=(5,5), activation='relu', input_shape=X_t1.shape[1:]))
model.add(Conv2D(filters=32, kernel_size=(5,5), activation='relu'))
model.add(MaxPool2D(pool_size=(2, 2)))
model.add(Dropout(rate=0.25))
model.add(Conv2D(filters=64, kernel_size=(3, 3), activation='relu'))
model.add(Conv2D(filters=64, kernel_size=(3, 3), activation='relu'))
model.add(MaxPool2D(pool_size=(2, 2)))
model.add(Dropout(rate=0.25))
model.add(Flatten())
model.add(Dense(256, activation='relu'))
model.add(Dropout(rate=0.5))
model.add(Dense(43, activation='softmax'))
#Compilation of the model
model.compile(loss='categorical_crossentropy', optimizer='adam', metrics=['accuracy'])
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In [ ]: model.summary
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Out[ ]: <bound method Model.summary of <keras.engine.sequential.Sequential object at 0x000002641E3F7510>>
```

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In [ ]: eps = 15
traffic = model.fit(X_t1, y_t1, batch_size=64, epochs=eps, validation_data=(X_t2, y_t2))
```

```

Epoch 1/15
491/491 [=====] - 159s 315ms/step - loss: 2.0490 - accuracy: 0.5066 - val_loss: 0.3927
- val_accuracy: 0.9068
Epoch 2/15
491/491 [=====] - 157s 320ms/step - loss: 0.4860 - accuracy: 0.8609 - val_loss: 0.1178
- val_accuracy: 0.9719
Epoch 3/15
491/491 [=====] - 354s 722ms/step - loss: 0.2919 - accuracy: 0.9179 - val_loss: 0.0921
- val_accuracy: 0.9791
Epoch 4/15
491/491 [=====] - 160s 326ms/step - loss: 0.2161 - accuracy: 0.9405 - val_loss: 0.0617
- val_accuracy: 0.9860
Epoch 5/15
491/491 [=====] - 160s 326ms/step - loss: 0.1818 - accuracy: 0.9505 - val_loss: 0.0416
- val_accuracy: 0.9902
Epoch 6/15
491/491 [=====] - 158s 322ms/step - loss: 0.1476 - accuracy: 0.9591 - val_loss: 0.0393
- val_accuracy: 0.9904
Epoch 7/15
491/491 [=====] - 156s 317ms/step - loss: 0.1454 - accuracy: 0.9598 - val_loss: 0.0378
- val_accuracy: 0.9913
Epoch 8/15
491/491 [=====] - 150s 305ms/step - loss: 0.1393 - accuracy: 0.9624 - val_loss: 0.0508
- val_accuracy: 0.9874
Epoch 9/15
491/491 [=====] - 149s 304ms/step - loss: 0.1282 - accuracy: 0.9658 - val_loss: 0.0473
- val_accuracy: 0.9871
Epoch 10/15
491/491 [=====] - 162s 330ms/step - loss: 0.1215 - accuracy: 0.9673 - val_loss: 0.0329
- val_accuracy: 0.9922
Epoch 11/15
491/491 [=====] - 162s 331ms/step - loss: 0.1141 - accuracy: 0.9700 - val_loss: 0.0229
- val_accuracy: 0.9949
Epoch 12/15
491/491 [=====] - 151s 307ms/step - loss: 0.1177 - accuracy: 0.9691 - val_loss: 0.0314
- val_accuracy: 0.9922
Epoch 13/15
491/491 [=====] - 154s 313ms/step - loss: 0.1115 - accuracy: 0.9712 - val_loss: 0.0172
- val_accuracy: 0.9963
Epoch 14/15
491/491 [=====] - 168s 341ms/step - loss: 0.1062 - accuracy: 0.9718 - val_loss: 0.0303
- val_accuracy: 0.9921
Epoch 15/15
491/491 [=====] - 190s 386ms/step - loss: 0.1166 - accuracy: 0.9698 - val_loss: 0.0427
- val_accuracy: 0.9898

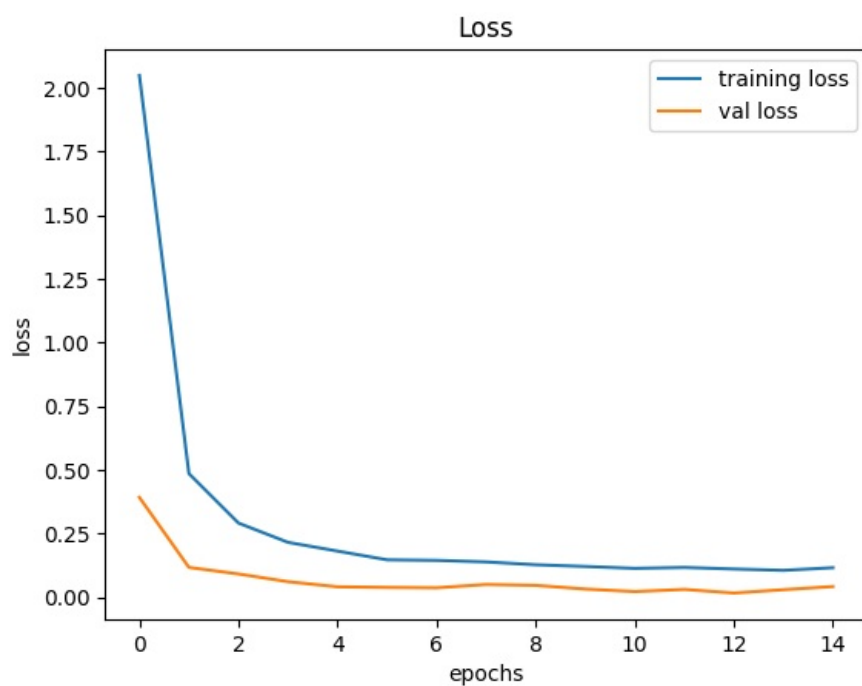
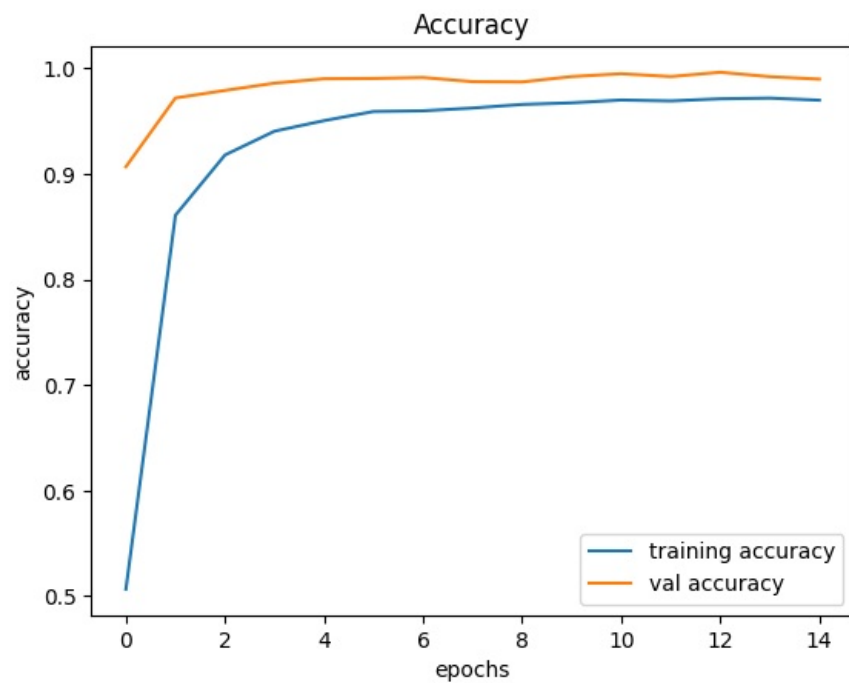
```

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In [ ]: model.save("my_model.h5")
```

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In [ ]: #plotting graphs for accuracy
plt.figure(0)
plt.plot(traffic.history['accuracy'], label='training accuracy')
plt.plot(traffic.history['val_accuracy'], label='val accuracy')
plt.title('Accuracy')
plt.xlabel('epochs')
plt.ylabel('accuracy')
plt.legend()
plt.show()
plt.figure(1)
plt.plot(traffic.history['loss'], label='training loss')
plt.plot(traffic.history['val_loss'], label='val loss')
plt.title('Loss')
plt.xlabel('epochs')
plt.ylabel('loss')
plt.legend()
plt.show()

```



```
In [ ]: #testing accuracy on test dataset
from sklearn.metrics import accuracy_score
y_test = pd.read_csv('Test.csv')
labels = y_test["ClassId"].values
imgs = y_test["Path"].values
data=[]
for img in imgs:
    image = Image.open(img)
    image = image.resize((30,30))
    data.append(np.array(image))
X_test=np.array(data)
```

```
In [ ]: labels
```

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Out[ ]: array([16,  1, 38, ...,  6,  7, 10], dtype=int64)
```

```
In [ ]: ypred = model.predict(X_test)

395/395 [=====] - 16s 41ms/step
```

```
In [ ]: ypred
```

```
Out[ ]: array([[2.7266858e-21, 6.5895818e-16, 1.4038198e-19, ..., 3.2540478e-14,
1.9330711e-17, 9.6278761e-17],
[8.4233666e-27, 1.0000000e+00, 3.4080128e-19, ..., 1.0838145e-29,
0.0000000e+00, 0.0000000e+00],
[0.0000000e+00, 0.0000000e+00, 0.0000000e+00, ..., 0.0000000e+00,
0.0000000e+00, 0.0000000e+00],
...,
[2.1785378e-05, 1.0880911e-02, 8.1380625e-04, ..., 2.3891672e-03,
1.4649702e-03, 3.7375483e-01],
[4.4127537e-16, 3.1284180e-09, 1.6430777e-11, ..., 7.6630968e-10,
1.4179572e-18, 1.1752815e-10],
[5.1035364e-20, 2.5668330e-11, 4.1388633e-12, ..., 2.0338837e-12,
1.3927731e-19, 5.6584370e-13]], dtype=float32)
```

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In [ ]: ypreda=np.argmax(ypred,axis=1)
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In [ ]: ypreda
```

```
Out[ ]: array([16,  1, 38, ...,  6,  7, 10], dtype=int64)
```

```
In [ ]: from sklearn.metrics import accuracy_score
#Accuracy with the test data
print(accuracy_score(labels, ypreda))
model.save('traffic_classifier.h5')

0.9557403008709422
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In [ ]:
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In [ ]:
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