

IOT-TRAFFIC MANAGEMENT

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DEVELOPMENT PART 1

Start Building the IOT Enable Traffic

Management in systems:

1) Introduction:

Traffic Management Is the combination of measures that serve to preserve traffic capacity and improve the security , safety , And reliability of the overall road transport system. These measures make

Use of ITS systems , services and projects in day-to-day operations that impact on road network performance.

2)Overall design:

***Identify the problem:** The first step is to identify the problem that the traffic control system is intended to solve.

***conduct a traffic study :** A traffic study is an analysis of the traffic patterns and volume at a particular location.

3)Software Design:

Chetu designs customizable Traffic Management Systems (TMS) dashboards with modules for features such as Quick Response , GIS – powered map recording And playback , report retrieval ,

field communication , variable speed limit signage, route guidance technologies , intersection control tools including automated traffic – light and pedestrian signal sequencing. Business needs vary , and so should the software ! With our custom traffic management

software development , have full control of your solution's design , functionality , performance ,scalability , and maintenance .

Make traffic infrastructure serve drivers and businesses with our top-notch traffic management software development services ! - We collect and process traffic, Vehicle , road , and weather data to help your business with navigation, data –driven decisions ,maintenance management , and more

4)System Testing:

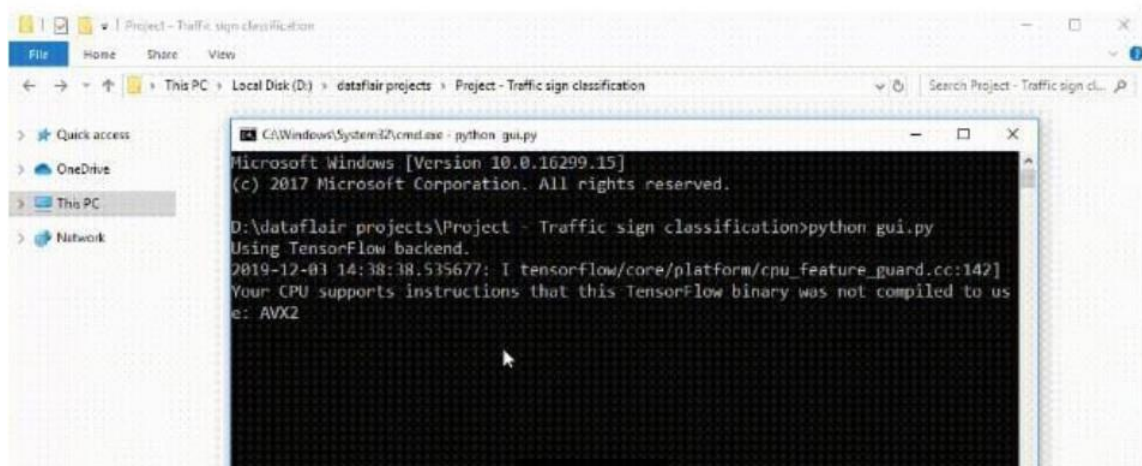
The software test program is intended to cover the software testing from design reviews through hardware/software integration testing . All software proposed for use in your TMS should be subjected to

testing before it is accepted and user to support operations. The extent and thorough ness of that testing should be based on the maturity of that software and the risk you are assuming in including it in your operations.

5)Python Script :

In this Python project example , we will build a deep neural network model that can classify traffic signs present in the image into different categories . With this model, we are able to read and understand traffic signs which are a very important task for all autonomous vehicles.

And extract the files into a folder such that you will have a train, test and a meta folder.



The screenshot shows a Windows File Explorer window with the address bar set to 'This PC > Local Disk (D:) > dataflair projects > Project - Traffic sign classification'. The table below represents the contents of this folder as shown in the image.

Name	Date modified	Type	Size
meta	04-Oct-19 12:52 PM	File folder	
test	04-Oct-19 12:52 PM	File folder	
train	04-Oct-19 12:55 PM	File folder	
Meta	25-Nov-18 6:13 PM	Microsoft Excel C...	2 KB
Test	25-Nov-18 6:13 PM	Microsoft Excel C...	418 KB
Train	25-Nov-18 6:13 PM	Microsoft Excel C...	1,896 KB

Create a python script file and name it traffic _ signs.py in the

project folder .

Our approach to building this traffic sign classification model is discussed in four

Steps:

- * Explore the dataset
- * Build a CNN model
- * Train and Validate the model

Test the model with test dataset

Step 1 : Explore the dataset

Our 'train' folder contains 43 folders each representing a different class. The range of the folder is from 0 to 42. with the help of the OS module we iterate and their respective labels in the data and labels list.

The PIL library is used to open image content into an array.

Finally ,we have stored all the images and their label into lists.

```
[9]: 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

data = []
labels = []
classes = 43
cur_path = os.getcwd()

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 + '\\' + a)
            image = image.resize((30,30))
            image = np.array(image)
            #im = Image.fromarray(image)
            data.append(image)
            labels.append(i)
        except:
            print("Error loading image")

data = np.array(data)
labels = np.array(labels)
```


We need to convert the list into numpy arrays for feeding to the model.

The shape of data is (39209, 30, 30, 3) which means that there are 39,209 images of size 30*30 pixels and the last 3 means the data contain colored images. We use the `train _ test _ split()` training and testing

data. From the `keras . Utils` package, we use `to_categorical` method to convert the labels present in `y _ train` and `t_test` into one –hot encoding.

```
[10]: print(data.shape, labels.shape)
      X_train, X_test, y_train, y_test = train_test_split(data, labels, test_size=0.2, random_state=42)

      print(X_train.shape, X_test.shape, y_train.shape, y_test.shape)

      y_train = to_categorical(y_train, 43)
      y_test = to_categorical(y_test, 43)

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

Step 2: Build a CNN Model To classify the images into their respective categories , we will build a CNN model .

The architecture of our model is :

.2 Conv2D Layer

(filter=32, kernel_size=(5,5), activation="relu")

. Maxpool2d layer (pool_size=(2,2))

. Dropout layer (ratio=0.25)

.2 Conv2D layer

(filter=64, kernel_size=(3,3), activation="relu")

. Maxpool2D layer (pool_size=(2,2))

. Dropout layer (rate=0.25)

. Flatten layer to squeeze the layers into 1 dimension

.Dense Fully connected layer(256 nodes,activation="relu")

.Dropout layer (rate=0.5)

.Dense layer (43 nodes,activation="softmax")

We Compile the model with adam optimizer which performs well and loss is "categorical _ crossentropy "because we have multiple classes to categorise.

```
[11]: model = Sequential()
      model.add(Conv2D(filters=32, kernel_size(5,5), activation='relu', input_shape=train_data.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'])
```

step 3: Train and Validate the model After building the model architecture, we then train the model using model .fit(). I tried with batch size 32 and 64 .our model performed better with 64

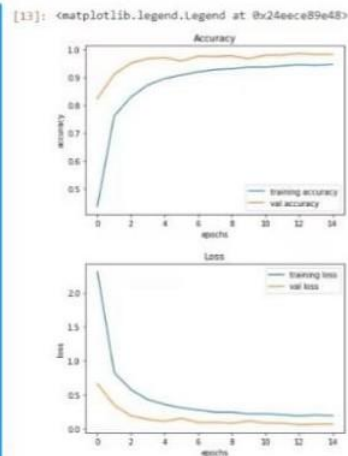
batch size . And after 15 epochs the accuracy was stable.

```
[13]: plt.figure(0)
      plt.plot(history.history['accuracy'], labels='training accuracy')
      plt.plot(history.history['val_accuracy'], labels='val accuracy')
      plt.title('Accuracy')
      plt.xlabel('epochs')
      plt.ylabel('accuracy')
      plt.legend()

      plt.figure(1)
      plt.plot(history.history['loss'], labels='training loss')
      plt.plot(history.history['val_loss'], labels='val loss')
      plt.title('Loss')
      plt.xlabel('epochs')
      plt.ylabel('loss')
      plt.legend()

[13]: <matplotlib.legend.Legend at 0x24eece89e48>
```

Plotting Accuracy



Accuracy and loss Graphs

Step 4: Test our model with test dataset Our dataset contains a test folder and in a test. CSV file ,we have the details related to the image path and their respective class labels using pandas . Then to predict the model , we have to resize our images to 30*30 pixels and make a numpy array containing all image data. From the sk learn.metrics , we imported the accuracy_score and observed how our model predicated the actual labels . We achieved a 95% accuracy in this model .

```
[14]: from sklearn.metrics import accuracy_score
import pandas as pd
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)

pred = model.predict_classes(X_test)

#Accuracy with the test data
from sklearn.metrics import accuracy_score
accuracy_score(labels, pred)
```

[14]: 0.9532066508313539

In the end, we are going to save the model that we have trained using the model that we have trained using the keras model. Save() function.

model . Save (' traffic _ classifier . h5')

Conclusion:

.With the help of ITS , the traffic congestions , rate of road accidents , wastage of fuels will be decreased to a large extent.

.This gives the people of the country a more economic mean of transportation withadvanced information of transits.

Hence with much more interest and advanced research in the field of ITS , it can be 100% efficiency in our country and can prove to be an effective solution to the traffic problems.