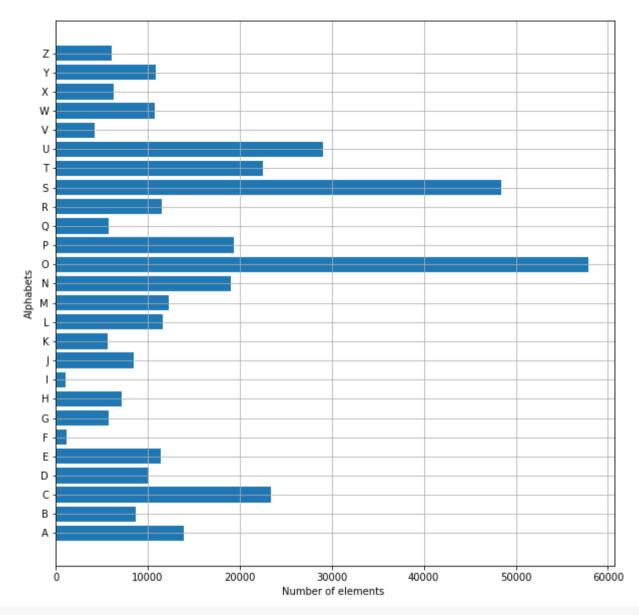
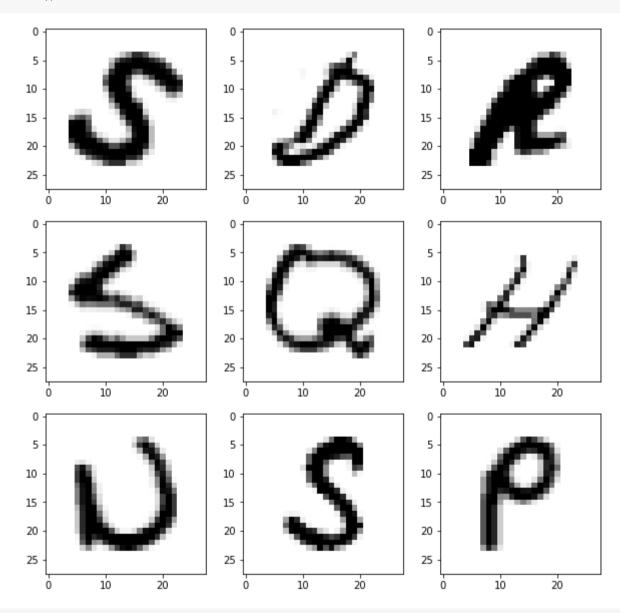
```
import matplotlib.pyplot as plt
import cv2
import numpy as np
from keras.models import Sequential
from keras.layers import Dense, Flatten, Conv2D, MaxPool2D, Dropout
from tensorflow.keras.optimizers import SGD, Adam
from keras.callbacks import ReduceLROnPlateau, EarlyStopping
from tensorflow.keras.utils import to categorical
import pandas as pd
import numpy as np
from sklearn.model selection import train_test_split
from sklearn.utils import shuffle
from tensorflow.python.keras import optimizers as opt
from google.colab import drive
drive.mount('/content/drive')
     Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount("/content/drive", force remount=True)
data = pd.read csv(r"/content/drive/MyDrive/SGP SEM 7/A Z Handwritten Data.csv.zip").astype('float32')
print(data.head(10))
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```
[10 rows x 785 columns]
X = data.drop('0',axis = 1)
y = data['0']
train x, test x, train y, test y = train test split(X, y, test size = 0.2)
train x = np.reshape(train x.values, (train x.shape[0], 28,28))
test x = np.reshape(test x.values, (test x.shape[0], 28,28))
print("Train data shape: ", train x.shape)
print("Test data shape: ", test x.shape)
     Train data shape: (297960, 28, 28)
     Test data shape: (74490, 28, 28)
word_dict = {0:'A',1:'B',2:'C',3:'D',4:'E',5:'F',6:'G',7:'H',8:'I',9:'J',10:'K',11:'L',12:'M',13:'N',14:'O',15:'P',16:'Q',17:'R',18:'
y int = np.int0(y)
count = np.zeros(26, dtype='int')
for i in y int:
    count[i] +=1
alphabets = []
for i in word dict.values():
    alphabets.append(i)
fig, ax = plt.subplots(1,1, figsize=(10,10))
ax.barh(alphabets, count)
plt.xlabel("Number of elements ")
plt.ylabel("Alphabets")
plt.grid()
plt.show()
```



```
shuff = shuffle(train_x[:100])
fig, ax = plt.subplots(3,3, figsize = (10,10))
axes = ax.flatten()
```

```
for i in range(9):
    _, shu = cv2.threshold(shuff[i], 30, 200, cv2.THRESH_BINARY)
    axes[i].imshow(np.reshape(shuff[i], (28,28)), cmap="Greys")
plt.show()
```



train_X = train_x.reshape(train_x.shape[0],train_x.shape[1],train_x.shape[2],1)

```
print("train data shape: ", train X.shape)
test_X = test_x.reshape(test_x.shape[0], test_x.shape[1], test_x.shape[2],1)
print("test data shape: ", test X.shape)
train x.shape: (297960, 28, 28, 1)
test x.shape: (74490, 28, 28, 1)
     train data shape: (297960, 28, 28, 1)
     test data shape: (74490, 28, 28, 1)
train yOHE = to categorical(train y, num classes = 26, dtype='int')
print("New shape of train labels: ", train yOHE.shape)
test yOHE = to categorical(test y, num classes = 26, dtype='int')
print("New shape of test labels: ", test vOHE.shape)
     New shape of train labels: (297960, 26)
     New shape of test labels: (74490, 26)
model = Sequential()
model.add(Conv2D(filters=32, kernel size=(3, 3), activation='relu', input shape=(28,28,1)))
model.add(MaxPool2D(pool_size=(2, 2), strides=2))
model.add(Conv2D(filters=64, kernel size=(3, 3), activation='relu', padding = 'same'))
model.add(MaxPool2D(pool size=(2, 2), strides=2))
model.add(Conv2D(filters=128, kernel size=(3, 3), activation='relu', padding = 'valid'))
model.add(MaxPool2D(pool size=(2, 2), strides=2))
model.add(Flatten())
model.add(Dense(64,activation ="relu"))
model.add(Dense(128,activation ="relu"))
```

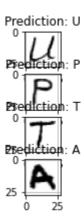
Model: "sequential"

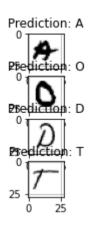
model.save(r'model hand.h5')

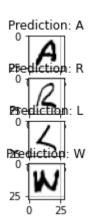
Layer (type)	Output Shape	Param #
conv2d (Conv2D)	 (None, 26, 26, 32)	320
<pre>max_pooling2d (MaxPooling2D)</pre>	(None, 13, 13, 32)	0
conv2d_1 (Conv2D)	(None, 13, 13, 64)	18496
<pre>max_pooling2d_1 (MaxPooling 2D)</pre>	(None, 6, 6, 64)	0
conv2d_2 (Conv2D)	(None, 4, 4, 128)	73856
<pre>max_pooling2d_2 (MaxPooling 2D)</pre>	(None, 2, 2, 128)	0
flatten (Flatten)	(None, 512)	0
dense (Dense)	(None, 64)	32832
dense_1 (Dense)	(None, 128)	8320
dense_2 (Dense)	(None, 26)	3354

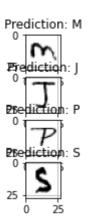
Total params: 137,178
Trainable params: 137,178
Non-trainable params: 0

```
print("The validation accuracy is :", history.history['val accuracy'])
print("The training accuracy is :", history.history['accuracy'])
print("The validation loss is :", history.history['val loss'])
print("The training loss is :", history.history['loss'])
     The validation accuracy is : [0.9770841598510742]
     The training accuracy is : [0.9569170475006104]
     The validation loss is : [0.08246023952960968]
     The training loss is : [0.15866714715957642]
fig, axes = plt.subplots(4,4, figsize=(14,3))
axes = axes.flatten()
for i,ax in enumerate(axes):
    img = np.reshape(test X[i], (28,28))
    ax.imshow(img, cmap="Greys")
    pred = word dict[np.argmax(test yOHE[i])]
    ax.set title("Prediction: "+pred)
    ax.grid()
```









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