

```
In [1]: # This Python 3 environment comes with many helpful analytics libraries
        installed
        # It is defined by the kaggle/python docker image: https://github.com/kaggle/docker-python
        # For example, here's several helpful packages to load in

import numpy as np # linear algebra
import pandas as pd # data processing, CSV file I/O (e.g. pd.read_csv)

# Input data files are available in the "../input/" directory.
# For example, running this (by clicking run or pressing Shift+Enter) will list the files in the input directory

import os
#print(os.listdir("../input"))

# Any results you write to the current directory are saved as output.

# Import modules

import numpy as np
import pandas as pd
import matplotlib.pyplot as plt

from sklearn.model_selection import train_test_split
import cv2

#keras
from keras.models import Sequential
from keras.layers import Dense, Dropout, Flatten
from keras.layers import Conv2D, MaxPooling2D
from keras.utils import np_utils
import sklearn.metrics as metrics
```

Using TensorFlow backend.

```
In [2]: train = pd.read_csv("../input/emnist-balanced-train.csv", delimiter =  
    ',')  
    test = pd.read_csv("../input/emnist-balanced-test.csv", delimiter = ',')  
    mapp = pd.read_csv("../input/emnist-balanced-mapping.txt", delimiter =  
    ', \\  
        index_col=0, header=None, squeeze=True)
```

```
In [3]: #! cat ../input/emnist-balanced-mapping.txt
```

```
In [4]: print("Train: %s, Test: %s, Map: %s" %(train.shape, test.shape, mapp.sh  
    ape))
```

Train: (112799, 785), Test: (18799, 785), Map: (47,)

```
In [5]: # Split x and y  
    train_x = train.iloc[:,1:]  
    train_y = train.iloc[:,0]  
    del train  
  
    test_x = test.iloc[:,1:]  
    test_y = test.iloc[:,0]  
    del test
```

```
In [6]: print(train_x.shape, train_y.shape, test_x.shape, test_y.shape)
```

(112799, 784) (112799,) (18799, 784) (18799,)

```
In [7]: # Constants  
    HEIGHT = 28  
    WIDTH = 28  
  
    def rotate(image):  
        image = image.reshape([HEIGHT, WIDTH])  
        image = np.fliplr(image)
```

```
image = np.rot90(image)
return image
```

```
# Flip and rotate image
train_x = np.asarray(train_x)
train_x = np.apply_along_axis(rotate, 1, train_x)
print("train_x:", train_x.shape)

train_x: (112799, 28, 28)
```

```
test_x = np.asarray(test_x)
test_x = np.apply_along_axis(rotate, 1, test_x)
print ("test_x:",test_x.shape)

test_x: (18799, 28, 28)
```

```
# Normalise
train_x = train_x.astype('float32')
train_x /= 255
test_x = test_x.astype('float32')
test_x /= 255
```

```
print(train_x[1])
plt.imshow(train_x[1], cmap=plt.get_cmap('gray'))
```

```
[ [0.          0.          0.          0.          0.          0.
   0.          0.          0.          0.          0.          0.
   0.          0.          0.          0.          0.          0.
   0.          0.          0.          0.          0.          0.
   0.          0.          0.          0.          ]
[0.          0.          0.          0.          0.          0.
 0.          0.          0.          0.          0.          0.
 0.          0.          0.          0.          0.          0.
 0.          0.          0.          0.          0.          0.
 0.          0.          0.          0.          ]
[0.          0.          0.          0.          0.          0.01568628
 0.01568628 0.          0.          0.          0.          0.
 0.          0.          0.          0.          0.          0.]
```

```

0.      0.      0.      0.      0.      0.
0.      0.      0.      0.      ]
[0.      0.      0.      0.      0.1254902  0.44313726
0.44313726 0.1254902 0.      0.      0.      0.
0.      0.      0.      0.      0.      0.
0.      0.      0.      0.      0.      0.
0.      0.      0.      0.      ]
[0.      0.      0.      0.01568628 0.4509804 0.94509804
0.94509804 0.44313726 0.01568628 0.      0.      0.
0.      0.      0.      0.      0.      0.
0.      0.      0.      0.      0.      0.
0.      0.      0.      0.      ]
[0.      0.      0.      0.08627451 0.6745098 0.98039216
0.96862745 0.49803922 0.01568628 0.      0.      0.
0.      0.      0.      0.      0.      0.
0.      0.      0.      0.      0.      0.
0.      0.      0.      0.      ]
[0.      0.      0.00784314 0.32156864 0.8980392 0.7529412
0.6666667 0.41960785 0.01176471 0.      0.      0.01176471
0.0627451 0.09803922 0.14117648 0.07843138 0.00392157 0.
0.      0.      0.      0.      0.      0.
0.      0.      0.      0.      ]
[0.      0.      0.03529412 0.54509807 0.9607843 0.50980395
0.09803922 0.07058824 0.      0.00392157 0.08627451 0.3254902
0.62352943 0.72156864 0.8392157 0.61960787 0.14117648 0.01176471
0.      0.      0.      0.      0.      0.
0.      0.      0.      0.      ]
[0.      0.      0.1254902 0.8      0.9764706 0.49803922
0.01568628 0.      0.00392157 0.18431373 0.62352943 0.9098039
0.98039216 0.9764706 0.9882353 0.9607843 0.6862745 0.30588236
0.02745098 0.      0.      0.      0.      0.
0.      0.      0.      0.      ]
[0.      0.      0.14509805 0.8509804 0.98039216 0.49803922
0.01568628 0.      0.09019608 0.80784315 0.96862745 0.96862745
0.8156863 0.5568628 0.6784314 0.9098039 0.98039216 0.80784315
0.18039216 0.      0.      0.      0.      0.
0.      0.      0.      0.      ]
[0.      0.      0.14509805 0.8509804 0.98039216 0.49803922
0.01960784 0.1254902 0.5137255 0.98039216 0.9098039 0.62352943

```

```

0.18431373 0.03921569 0.09019608 0.39215687 0.91764706 0.9607843
0.49803922 0. 0. 0. 0. 0.
0. 0. 0. 0. ]
[0. 0. 0.14509805 0.8509804 0.98039216 0.49803922
0.05882353 0.49803922 0.9490196 0.85490197 0.37254903 0.08627451
0.00392157 0. 0. 0.13333334 0.8 0.99215686
0.79607844 0.01568628 0. 0. 0. 0.
0. 0. 0. 0. ]
[0. 0. 0.14509805 0.8509804 0.98039216 0.5254902
0.37254903 0.8627451 0.9137255 0.30980393 0.03137255 0.
0. 0. 0. 0.03137255 0.49803922 0.9607843
0.87058824 0.03529412 0. 0. 0. 0.
0. 0. 0. 0. ]
[0. 0. 0.14509805 0.8509804 0.98039216 0.6784314
0.827451 0.98039216 0.67058825 0.01568628 0. 0.
0. 0. 0. 0. 0.2 0.87058824
0.9607843 0.13333334 0. 0. 0. 0.
0. 0. 0. 0. ]
[0. 0. 0.14509805 0.8509804 0.99607843 0.99215686
0.9843137 0.6862745 0.13333334 0. 0. 0.
0. 0. 0. 0. 0.08235294 0.6745098
0.9843137 0.4509804 0.01568628 0. 0. 0.
0. 0. 0. 0. ]
[0. 0. 0.2 0.87058824 0.99607843 0.99607843
0.9098039 0.3254902 0.01176471 0. 0. 0.
0. 0. 0. 0. 0.01960784 0.49803922
0.9764706 0.6745098 0.08235294 0. 0. 0.
0. 0. 0. 0. ]
[0. 0.01568628 0.44705883 0.9607843 1. 0.9843137
0.67058825 0.08627451 0. 0. 0. 0.
0. 0. 0. 0. 0.00784314 0.32156864
0.9098039 0.8627451 0.19607843 0. 0. 0.
0.00392157 0. 0. 0. 0. ]
[0. 0.01568628 0.49019608 0.9764706 0.99607843 0.91764706
0.33333334 0.01176471 0. 0. 0. 0.
0. 0. 0. 0. 0. 0.08627451
0.6666667 0.9607843 0.5019608 0.03921569 0. 0.01568628
0.18431373 0.07058824 0. 0. 0. ]
[0. 0.01176471 0.37254903 0.93333334 0.99607843 0.9137255

```

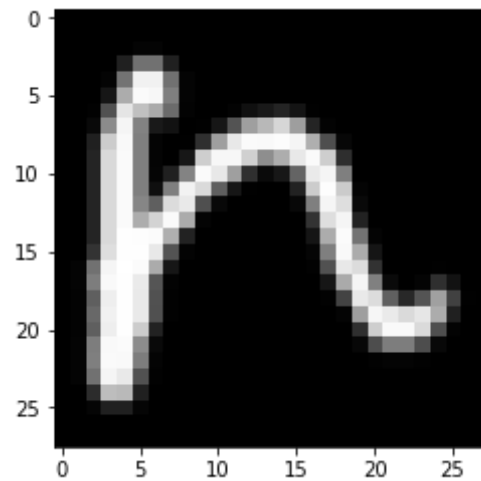
```

0.32156864 0.00784314 0.      0.      0.      0.
0.      0.      0.      0.      0.      0.00784314
0.2627451 0.9137255 0.8627451 0.3764706 0.16078432 0.32941177
0.627451 0.25490198 0.00392157 0.      ]
[0.      0.01176471 0.40784314 0.94509804 0.99607843 0.9137255
0.3137255 0.00784314 0.      0.      0.      0.
0.      0.      0.      0.      0.      0.
0.01568628 0.62352943 0.9647059 0.91764706 0.85490197 0.9019608
0.6745098 0.1254902 0.      0.      ]
[0.      0.01176471 0.37254903 0.93333334 0.99607843 0.8
0.13333334 0.      0.      0.      0.      0.
0.      0.      0.      0.      0.      0.
0.      0.18431373 0.80784315 0.9764706 0.9764706 0.85490197
0.30980393 0.01176471 0.      0.      ]
[0.      0.01568628 0.49019608 0.9764706 0.9843137 0.54901963
0.03529412 0.      0.      0.      0.      0.
0.      0.      0.      0.      0.      0.
0.      0.02745098 0.3019608 0.49019608 0.49019608 0.3019608
0.03137255 0.      0.      0.      ]
[0.      0.01568628 0.49803922 0.98039216 0.9764706 0.49019608
0.01568628 0.      0.      0.      0.      0.
0.      0.      0.      0.      0.      0.
0.      0.      0.00784314 0.01568628 0.01568628 0.00784314
0.      0.      0.      0.      ]
[0.      0.01568628 0.44705883 0.9607843 0.9137255 0.32156864
0.00784314 0.      0.      0.      0.      0.
0.      0.      0.      0.      0.      0.
0.      0.      0.      0.      0.      0.
0.      0.      0.      0.      ]
[0.      0.      0.1764706 0.7411765 0.7254902 0.12941177
0.      0.      0.      0.      0.      0.
0.      0.      0.      0.      0.      0.
0.      0.      0.      0.      0.      0.
0.      0.      0.      0.      ]
[0.      0.      0.01960784 0.1254902 0.1254902 0.01960784
0.      0.      0.      0.      0.      0.
0.      0.      0.      0.      0.      0.
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0.      0.      0.      0.      ]

```

```
[0.  0.  0.  0.  0.  0.
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 0.  0.  0.  0.  0.  0.
 0.  0.  0.  0.  0.  0.]
[0.  0.  0.  0.  0.  0.
 0.  0.  0.  0.  0.  0.
 0.  0.  0.  0.  0.  0.
 0.  0.  0.  0.  0.  0.
 0.  0.  0.  0.  0.  0.]
```

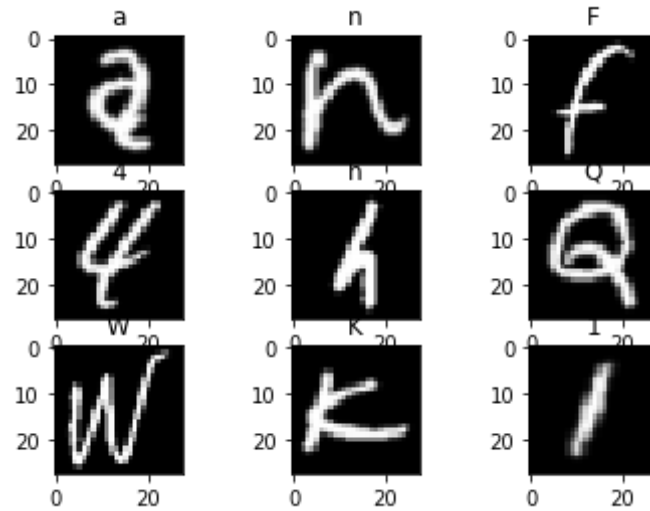
Out[11]: <matplotlib.image.AxesImage at 0x7f2ffe3346a0>



In [12]: `chr(mapp[train_y[1]])`

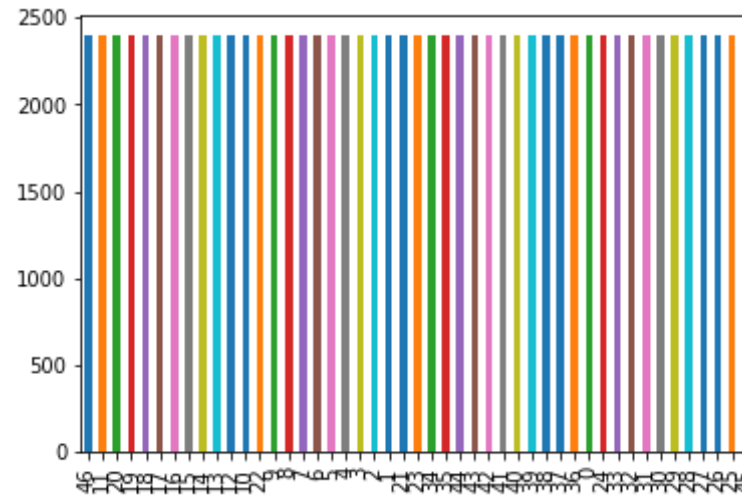
Out[12]: 'n'

```
In [13]: # plot image
for i in range(0, 9):
    plt.subplot(330 + (i+1))
    plt.imshow(train_x[i], cmap=plt.get_cmap('gray'))
    plt.title(chr(mapp[train_y[i]]))
```



```
In [14]: train_y.value_counts().plot(kind='bar')
```

```
Out[14]: <matplotlib.axes._subplots.AxesSubplot at 0x7f2fff1c5780>
```



```
In [15]: # number of classes
num_classes = train_y.nunique()
```



```
In [16]: # One hot encoding
train_y = np_utils.to_categorical(train_y, num_classes)
test_y = np_utils.to_categorical(test_y, num_classes)
print("train_y: ", train_y.shape)
print("test_y: ", test_y.shape)

train_y: (112799, 47)
test_y: (18799, 47)
```

```
In [17]: print ("train_x:",train_x.shape)

train_x: (112799, 28, 28)
```

```
In [18]: # Reshape image for CNN
train_x = train_x.reshape(-1, HEIGHT, WIDTH, 1)
test_x = test_x.reshape(-1, HEIGHT, WIDTH, 1)
```

```
In [19]: print ("train_x:",train_x.shape)

train_x: (112799, 28, 28, 1)
```

```
In [20]: # partition to train and val
train_x, val_x, train_y, val_y = train_test_split(train_x, train_y, test_size= 0.10, random_state=7)
```

```
In [21]: # Building model
# ((w - k + 2P)/S) + 1
model = Sequential()

model.add(Conv2D(filters=128, kernel_size=(5,5), padding = 'same', activation='relu',\
                input_shape=(HEIGHT, WIDTH,1)))
model.add(MaxPooling2D(pool_size=(2,2), strides=(2,2)))
model.add(Conv2D(filters=64, kernel_size=(3,3) , padding = 'same', activation='relu'))
model.add(MaxPooling2D(pool_size=(2,2)))
```

```

model.add(Flatten())
model.add(Dense(units=128, activation='relu'))
model.add(Dropout(.5))
model.add(Dense(units=num_classes, activation='softmax'))

model.summary()

```

Layer (type)	Output Shape	Param #
conv2d_1 (Conv2D)	(None, 28, 28, 128)	3328
max_pooling2d_1 (MaxPooling2D)	(None, 14, 14, 128)	0
conv2d_2 (Conv2D)	(None, 14, 14, 64)	73792
max_pooling2d_2 (MaxPooling2D)	(None, 7, 7, 64)	0
flatten_1 (Flatten)	(None, 3136)	0
dense_1 (Dense)	(None, 128)	401536
dropout_1 (Dropout)	(None, 128)	0
dense_2 (Dense)	(None, 47)	6063
Total params: 484,719		
Trainable params: 484,719		
Non-trainable params: 0		

```

In [22]: model.compile(loss='categorical_crossentropy', optimizer='adam', metrics=['accuracy'])

```

```

In [ ]: history = model.fit(train_x, train_y, epochs=300, batch_size=512, verbose=1, \
                             validation_data=(val_x, val_y))

```

```
In [ ]: def show_result(epochs, acc, val_acc):
        plt.plot(epochs, acc, 'y')
        plt.plot(epochs, val_acc, 'b')
        plt.title('Model accuracy', fontsize = 18)
        plt.ylabel('Accuracy', fontsize = 18)
        plt.xlabel('Epoch', fontsize = 18)
        plt.legend(['Train', 'Val'], loc='best')
        plt.savefig('res.png')
        plt.show()
```

```
In [ ]: history.history.keys()
```

```
In [ ]: train = history.history['acc']
        val = history.history['val_acc']
        epochs = range(1, len(train)+1)
```

```
In [ ]: show_result(epochs, train, val)
```

```
In [ ]: #Save the model
        # serialize model to JSON
        model_json = model.to_json()
        with open("model2.json", "w") as json_file:
            json_file.write(model_json)
        # serialize weights to HDF5
        model.save_weights("model2.h5")
        print("Saved model to disk")
```

```
In [ ]: #from joblib import dump
```

```
In [ ]: #dump(model, 'model.pkl')
```

```
In [ ]: #!ls ../
```

```
In [ ]:
```

In []: