



```
from google.colab import files
files.upload()
```


 No file chosen Upload widget is only available when the cell has been executed in the current browser session. Please rerun this cell to enable.
Saving kaggle.json to kaggle (1).json
{'kaggle.json': b'{"username":"dikshabhati"."key":"84f76c6a9c76c3c8d591f12b28ba6927"}'}

```
#create a new file and move the kaggle file into it
!mkdir -p ~/.kaggle
!cp kaggle.json ~/.kaggle/
!chmod 600 ~/.kaggle/kaggle.json
```


```
!ls ~/.kaggle
```

 kaggle.json


```
!kaggle competitions download -c digit-recognizer
```

 Warning: Looks like you're using an outdated API Version, please consider updating (see [here](#))
test.csv.zip: Skipping, found more recently modified local copy (use --force to force download)
sample_submission.csv: Skipping, found more recently modified local copy (use --force to force download)
train.csv.zip: Skipping, found more recently modified local copy (use --force to force download)

```
!unzip -q "../content/train.csv"
```

 [../content/train.csv]
End-of-central-directory signature not found. Either this file is not a zipfile, or it constitutes one disk of a multi-part archive. In the latter case the central directory and zipfile comment will be found on the last disk(s) of this archive.
replace train.csv? [y]es, [n]o, [A]ll, [N]one, [r]ename:

```
!unzip -q "../content/test.csv"
```

 [../content/test.csv]
End-of-central-directory signature not found. Either this file is not a zipfile, or it constitutes one disk of a multi-part archive. In the latter case the central directory and zipfile comment will be found on the last disk(s) of this archive.
replace test.csv? [y]es, [n]o, [A]ll, [N]one, [r]ename:

```
import pandas as pd
train_df=pd.read_csv("../content/train.csv")
test_df=pd.read_csv("../content/test.csv")
```

```
train_df.shape
```

```
↳ (42000, 785)
```

```
test_df.shape
```

```
↳ (28000, 784)
```

```
X_train = train_df.drop(labels = ["label"],axis = 1)
Y_train = train_df["label"]
```

```
X_train.shape
```

```
↳ (42000, 784)
```

```
Y_train
```

```
↳ 0      1
   1      0
   2      1
   3      4
   4      0
   ..
  41995   0
  41996   1
  41997   7
  41998   6
  41999   9
   Name: label, Length: 42000, dtype: int64
```

```
train_df.dtypes
```

```
↳ label      int64
   pixel0     int64
   pixel1     int64
   pixel2     int64
   pixel3     int64
   ...
  pixel779    int64
  pixel780    int64
  pixel781    int64
  pixel782    int64
  pixel783    int64
   Length: 785, dtype: object
```

```
sum(train_df.isnull().sum())
```

```
↳ 0
```

```
sum(test_df.isnull().sum())
```

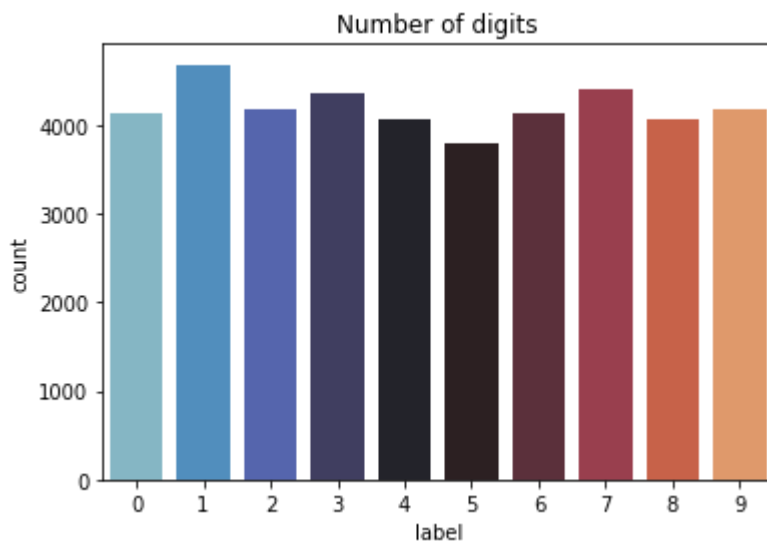
0

```
Y_train.value_counts()
```

```
1    4684
7    4401
3    4351
9    4188
2    4177
6    4137
0    4132
4    4072
8    4063
5    3795
Name: label, dtype: int64
```

```
import seaborn as sns
import matplotlib.pyplot as plt
sns.countplot(Y_train,palette='icefire')
plt.title("Number of digits")
```

```
Text(0.5, 1.0, 'Number of digits')
```



```
import seaborn as sns
import matplotlib.pyplot as plt
```

```
X_train=X_train.values.reshape(X_train.shape[0],28,28,1)
X_train.shape
```

```
(42000, 28, 28, 1)
```

```
test_df=test_df.values.reshape(test_df.shape[0],28,28,1)
test_df.shape
```

```
↳ (28000, 28, 28, 1)
```

```
from keras.utils.np_utils import to_categorical
Y_train=to_categorical(Y_train,num_classes=10)
```

Y_train

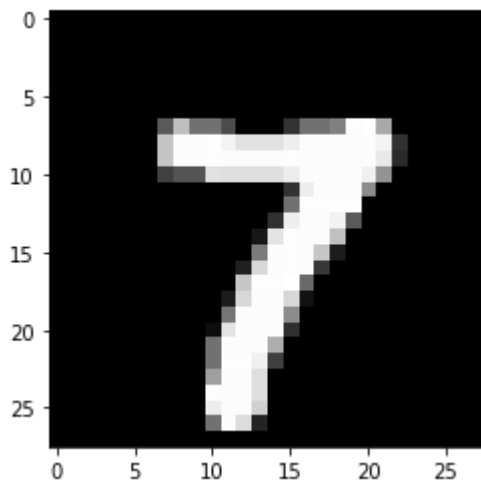
```
↳ array([[0., 1., 0., ..., 0., 0., 0.],
        [1., 0., 0., ..., 0., 0., 0.],
        [0., 1., 0., ..., 0., 0., 0.],
        ...,
        [0., 0., 0., ..., 1., 0., 0.],
        [0., 0., 0., ..., 0., 0., 0.],
        [0., 0., 0., ..., 0., 0., 1.]], dtype=float32)
```

```
X_train = X_train / 255.0
test_df = test_df / 255.0
```

```
from sklearn.model_selection import train_test_split
X_train, X_val, Y_train, Y_val = train_test_split(X_train, Y_train, test_size = 0.1, random_s
```

```
plt.imshow(X_train[0][:,:,0],cmap="gray")
```

```
↳ <matplotlib.image.AxesImage at 0x7f7cb8424160>
```



```
plt.imshow(X_train[4][:,:,0],cmap="gray")
```

```
↳
```

<matplotlib.image.AxesImage at 0x7f7cb8e76fd0>



```
from keras.models import Sequential
from keras.layers import Dense,Conv2D,Flatten,Dropout,MaxPool2D
from keras.layers.normalization import BatchNormalization

model=Sequential()

model.add(Conv2D(filters = 64, kernel_size = (5,5),padding = 'Same',activation ='relu', input_shape=(28,28,1)))
model.add(Conv2D(filters = 64, kernel_size = (5,5),padding = 'Same',activation ='relu'))
model.add(BatchNormalization())
model.add(MaxPool2D(pool_size=(2,2)))
model.add(Dropout(0.25))

model.add(Conv2D(filters = 128, kernel_size = (3,3),padding = 'Same', activation ='relu'))
model.add(Conv2D(filters = 128, kernel_size = (3,3),padding = 'Same', activation ='relu'))
model.add(MaxPool2D(pool_size=(2,2), strides=(2,2)))
model.add(BatchNormalization())
model.add(Dropout(0.25))

model.add(Conv2D(filters = 64, kernel_size = (3,3),padding = 'Same', activation ='relu'))
model.add(Conv2D(filters = 64, kernel_size = (3,3),padding = 'Same', activation ='relu'))
model.add(MaxPool2D(pool_size=(2,2), strides=(2,2)))
model.add(BatchNormalization())

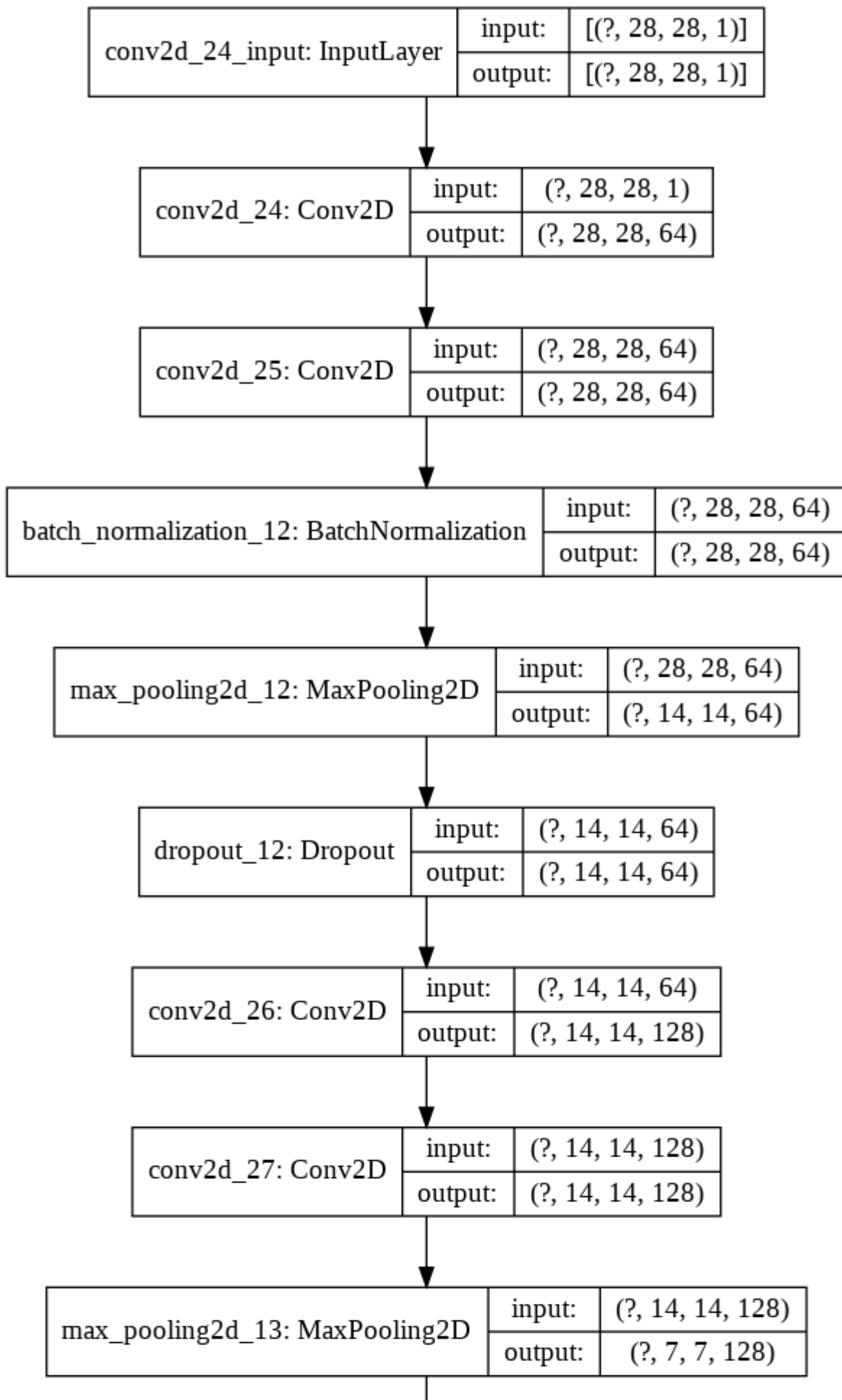
model.add(Flatten())
model.add(Dense(256, activation = "relu"))
model.add(Dropout(0.5))
model.add(Dense(10, activation = "softmax"))

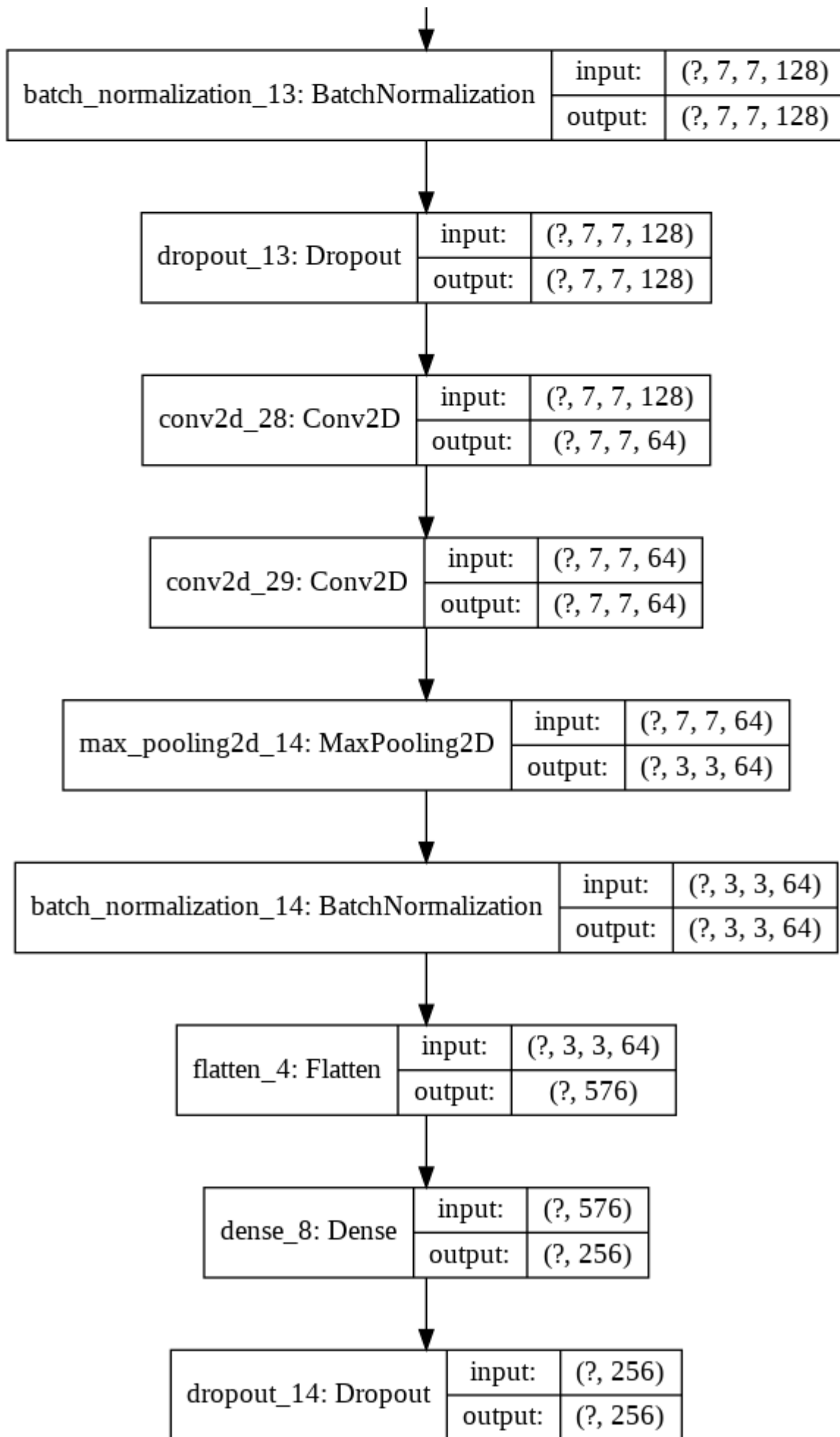
from tensorflow import keras
adam=keras.optimizers.Adam(learning_rate=0.001,epsilon=1e-08,beta_1=0.9,beta_2=0.999)

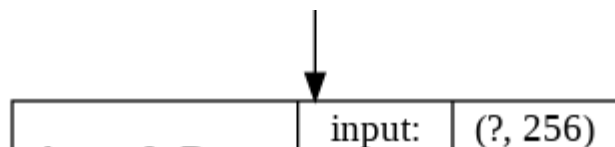
model.compile(optimizer = adam , loss = "categorical_crossentropy", metrics=["accuracy"])

from keras.utils import plot_model
plot_model(model, show_shapes=True, show_layer_names=True)
```









```
model.summary()
```

Model: "sequential_4"

Layer (type)	Output Shape	Param #
conv2d_24 (Conv2D)	(None, 28, 28, 64)	1664
conv2d_25 (Conv2D)	(None, 28, 28, 64)	102464
batch_normalization_12 (Batch Normalization)	(None, 28, 28, 64)	256
max_pooling2d_12 (MaxPooling2D)	(None, 14, 14, 64)	0
dropout_12 (Dropout)	(None, 14, 14, 64)	0
conv2d_26 (Conv2D)	(None, 14, 14, 128)	73856
conv2d_27 (Conv2D)	(None, 14, 14, 128)	147584
max_pooling2d_13 (MaxPooling2D)	(None, 7, 7, 128)	0
batch_normalization_13 (Batch Normalization)	(None, 7, 7, 128)	512
dropout_13 (Dropout)	(None, 7, 7, 128)	0
conv2d_28 (Conv2D)	(None, 7, 7, 64)	73792
conv2d_29 (Conv2D)	(None, 7, 7, 64)	36928
max_pooling2d_14 (MaxPooling2D)	(None, 3, 3, 64)	0
batch_normalization_14 (Batch Normalization)	(None, 3, 3, 64)	256
flatten_4 (Flatten)	(None, 576)	0
dense_8 (Dense)	(None, 256)	147712
dropout_14 (Dropout)	(None, 256)	0
dense_9 (Dense)	(None, 10)	2570
Total params: 587,594		
Trainable params: 587,082		
Non-trainable params: 512		

```
from keras.preprocessing.image import ImageDataGenerator
import tensorflow as tf
```

```
data_aug = tf.keras.preprocessing.image.ImageDataGenerator(featurewise_center=False,
```



```
samplewise_center=False,  
featurewise_std_normalization=False,  
samplewise_std_normalization=False,  
zca_whitening=False,  
rotation_range=10,  
zoom_range=0.1,  
width_shift_range=0.1,  
height_shift_range=0.1,  
horizontal_flip=False,  
vertical_flip=False)
```

```
data_aug.fit(X_train)
```

```
epochs=10  
batch_size=20
```

```
from keras.optimizers import RMSprop  
from sklearn.metrics import confusion_matrix  
from keras.callbacks import ReduceLROnPlateau
```

```
learning_rate_reduction=ReduceLROnPlateau(monitor='val_acc',patience=1,verbose=2,factor=0.6,m
```

```
hist = model.fit_generator(data_aug.flow(X_train,Y_train,batch_size=batch_size),  
                           steps_per_epoch=X_train.shape[0]//batch_size,  
                           epochs=epochs,  
                           validation_data=(X_val,Y_val),  
                           validation_steps=10000//batch_size,  
                           callbacks=[learning_rate_reduction])
```



```

Epoch 1/10
1890/1890 [=====] - ETA: 0s - loss: 0.2824 - accuracy: 0.9126W/
WARNING:tensorflow:Reduce LR on plateau conditioned on metric `val_acc` which is not av
1890/1890 [=====] - 794s 420ms/step - loss: 0.2824 - accuracy:
Epoch 2/10
1890/1890 [=====] - ETA: 0s - loss: 0.1216 - accuracy: 0.9656W/
1890/1890 [=====] - 766s 405ms/step - loss: 0.1216 - accuracy:
Epoch 3/10
1890/1890 [=====] - ETA: 0s - loss: 0.0942 - accuracy: 0.9731W/
1890/1890 [=====] - 766s 405ms/step - loss: 0.0942 - accuracy:
Epoch 4/10
1890/1890 [=====] - ETA: 0s - loss: 0.0823 - accuracy: 0.9764W/
1890/1890 [=====] - 766s 405ms/step - loss: 0.0823 - accuracy:
Epoch 5/10
1890/1890 [=====] - ETA: 0s - loss: 0.0705 - accuracy: 0.9801W/
1890/1890 [=====] - 764s 404ms/step - loss: 0.0705 - accuracy:
Epoch 6/10
 667/1890 [=====>.....] - ETA: 8:23 - loss: 0.0603 - accuracy: 0.9836

```

KeyboardInterrupt

Traceback (most recent call last):

```
print(hist.history.keys())
```

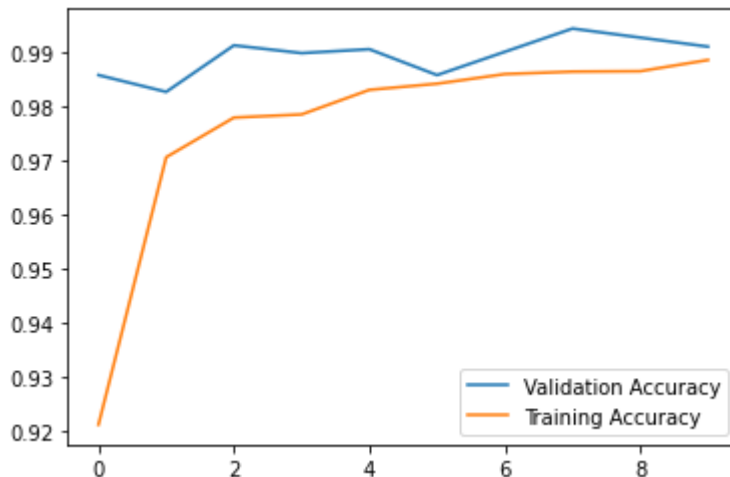
```
dict_keys(['loss', 'accuracy', 'val_loss', 'val_accuracy', 'lr'])
```

```

plt.plot(hist.history["val_accuracy"],label="Validation Accuracy")
plt.plot(hist.history["accuracy"],label="Training Accuracy")
plt.legend()
plt.plot()

```

```
[ ]
```

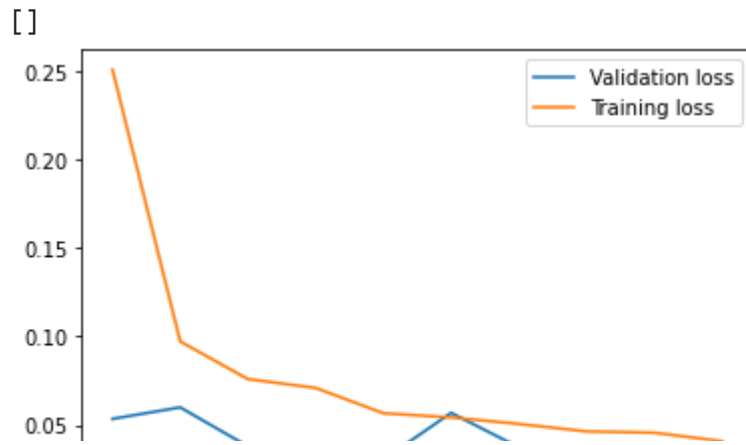


```

plt.plot(hist.history["val_loss"],label="Validation loss")
plt.plot(hist.history["loss"],label="Training loss")
plt.legend()
plt.plot()

```

```
[ ]
```



```
Y_pred=model.predict(test_df)
```

```
Y_pred.shape
```

```
↳ (28000, 10)
```

```
Y_pred[0]
```

```
↳ array([2.3740344e-13, 5.3405846e-11, 1.0000000e+00, 1.9546807e-08,
        1.9335039e-12, 5.3650389e-14, 3.8102445e-13, 1.3427028e-09,
        4.2735740e-10, 5.5917705e-12], dtype=float32)
```

```
import numpy as np
results=np.argmax(Y_pred,axis=1)
results=pd.Series(results,name='Label')
```

```
submission = pd.concat([pd.Series(range(1,28001),name = "ImageId"),results],axis = 1)
submission.to_csv('submission_datarecognizer1.csv',index=False)
```

```
i=9
print("Image label is: ", results[i])
plt.imshow(test_df[i][:,:,0])
```

```
↳
```

Image label is: 3

<matplotlib.image.AxesImage at 0x7f7cc4cae860>

