

Train the model on IBM Cloud

Team ID	PNT2022TMID27942
Project Name	Project - Natural Disasters Intensity Analysis and Classification using Artificial Intelligence

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Model Building

1.Convolutional Neural Network (CNN)

Import Packages & Libraries

In [1]:

```
#import the necessary Libraries
import tensorflow
from tensorflow import keras
from keras import Sequential
from keras.layers import Convolution2D,MaxPooling2D,Flatten,Dense,Dropout
from keras.preprocessing.image import ImageDataGenerator as idm
```

In [2]:

```
import numpy as np
import warnings
warnings.filterwarnings('ignore')
```

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Create model

In [10]:

```
#Defining model
model=Sequential()
```

Build and configure model

In [11]:

```
#Convolution Layer 1
model.add(Convolution2D(32,(3,3),activation='relu',input_shape=(76,76,3)))
model.add(MaxPooling2D(pool_size=(2,2)))
#Convolution Layer 2
model.add(Convolution2D(32,(3,3),activation='relu',input_shape=(76,76,3)))
model.add(MaxPooling2D(pool_size=(2,2)))
#Convolution Layer 3
model.add(Convolution2D(64,(3,3),activation='relu',input_shape=(76,76,3)))
model.add(MaxPooling2D(pool_size=(2,2)))
#Convolution Layer 4
model.add(Convolution2D(64,(3,3),activation='relu',input_shape=(76,76,3)))
model.add(MaxPooling2D(pool_size=(2,2)))
model.add(Flatten())
model.add(Dense(300,activation='relu'))
model.add(Dense(150,activation='relu'))
model.add(Dense(3,activation='softmax'))
```

Training the model

```
In [87]: #Model fitting - training and validation
history = model.fit_generator(Xtrain, steps_per_epoch= len(Xtrain), epochs=15, validation_data=Xtest, validation_steps= len(Xtest))

Epoch 1/15
35/35 [=====] - 144s 4s/step - loss: 0.9885 - accuracy: 0.4835 - val_loss: 0.8942 - val_accuracy: 0.5806
Epoch 2/15
35/35 [=====] - 128s 4s/step - loss: 0.7582 - accuracy: 0.6337 - val_loss: 0.8916 - val_accuracy: 0.6000
Epoch 3/15
35/35 [=====] - 131s 4s/step - loss: 0.6990 - accuracy: 0.6578 - val_loss: 0.6299 - val_accuracy: 0.6710
Epoch 4/15
35/35 [=====] - 129s 4s/step - loss: 0.6197 - accuracy: 0.7175 - val_loss: 0.6409 - val_accuracy: 0.7548
Epoch 5/15
35/35 [=====] - 135s 4s/step - loss: 0.5430 - accuracy: 0.7698 - val_loss: 1.0394 - val_accuracy: 0.5484
Epoch 6/15
35/35 [=====] - 133s 4s/step - loss: 0.5848 - accuracy: 0.7537 - val_loss: 0.7078 - val_accuracy: 0.7677
Epoch 7/15
35/35 [=====] - 128s 4s/step - loss: 0.4550 - accuracy: 0.8128 - val_loss: 0.4577 - val_accuracy: 0.7677
Epoch 8/15
35/35 [=====] - 127s 4s/step - loss: 0.4582 - accuracy: 0.8154 - val_loss: 0.4633 - val_accuracy: 0.8194
Epoch 9/15
35/35 [=====] - 126s 4s/step - loss: 0.5447 - accuracy: 0.7646 - val_loss: 0.6192 - val_accuracy: 0.7742
Epoch 10/15
35/35 [=====] - 139s 4s/step - loss: 0.4137 - accuracy: 0.8374 - val_loss: 0.3813 - val_accuracy: 0.8194
Epoch 11/15
35/35 [=====] - 128s 4s/step - loss: 0.5409 - accuracy: 0.7722 - val_loss: 0.6150 - val_accuracy: 0.7677
Epoch 12/15
```

```
35/35 [=====] - 134s 4s/step - loss: 0.4355 - accuracy: 0.8277 - val_loss: 0.3717 - val_accuracy: 0.8516
```

Testing the model

```
In [88]: from tensorflow.keras.preprocessing import image
```

```
In [89]: test_img=image.load_img('/content/drive/MyDrive/IBM_DATASET/test_set/Earthquake/Copy of 22.jpg',target_size=(76,76))
test_img
```

```
Out[89]:
```



```
In [90]: x=image.img_to_array(test_img)
x=np.expand_dims(x,axis=0)
predicted=np.argmax(model.predict(x))
Prediction_category=['Cyclone','Earthquake','Flood']
Prediction_category[predicted]
```

```
1/1 [=====] - 0s 193ms/step
```

```
Out[90]: 'Earthquake'
```

```
In [91]: test_img1=image.load_img('/content/drive/MyDrive/IBM_DATASET/train_set/Cyclone/1.jpg',target_size=(76,76))
test_img1
```

```
Out[91]:
```



```
In [92]: x=image.img_to_array(test_img1)
x=np.expand_dims(x,axis=0)
predicted=np.argmax(model.predict(x))
Prediction_category=['Cyclone','Earthquake','Flood']
Prediction_category[predicted]
```

```
1/1 [=====] - 0s 39ms/step
```

```
Out[92]: 'Cyclone'
```

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Saving the model

In [101.]

#saving the Model
model.save('CNN_Model_for_Disaster_Classification.h5')

Plotting Accuracy Graph

In [102.]

#importing dependencies to plot the graph
import matplotlib.pyplot as plt
import seaborn as sns
import cv2

In [105.]

#Training and Validation Accuracy Plots
epochs_range = range(15)

plt.figure(figsize=(6,6))
plt.plot(epochs_range, history.history['accuracy'], label='Training Accuracy')
plt.plot(epochs_range, history.history['val_accuracy'], label='Validation Accuracy')
plt.legend()
plt.title('Training and Testing Accuracy')
plt.show()

Epoch	Training Accuracy	Validation Accuracy
0	0.48	0.58
1	0.62	0.60
2	0.65	0.65
3	0.72	0.75
4	0.78	0.55
5	0.75	0.75
6	0.80	0.78
7	0.82	0.80
8	0.78	0.75
9	0.82	0.80
10	0.78	0.75
11	0.82	0.80
12	0.80	0.78
13	0.82	0.75
14	0.85	0.82

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Confusion Matrix

In [107.]

from sklearn.metrics import classification_report, confusion_matrix

In [110.]

#Confusion Matrix and Classification Report
Y_pred = model.predict_generator(Xtest,500 // 100)
y_pred = np.argmax(Y_pred, axis=1)
print('Confusion Matrix')
print(confusion_matrix(Xtest.classes, y_pred))

WARNING:tensorflow:Your input ran out of data; interrupting training. Make sure that your dataset or generator can generate at least `steps_per_epoch * epochs` batches (in this case, 5 batches). You may need to use the repeat() function when building your dataset.

Confusion Matrix
[[20 22 13]
[16 22 12]
[16 14 20]]

Classification Report

In [111.]

print('Classification Report')
target_names = ['Cyclone', 'Earthquake', 'Flood']
print(classification_report(Xtest.classes, y_pred, target_names=target_names))

Classification Report

precision recall f1-score support

Cyclone 0.38 0.36 0.37 55

Earthquake 0.38 0.44 0.41 50

Flood 0.44 0.40 0.42 50

accuracy 0.40 0.40 0.40 155

macro avg 0.40 0.40 0.40 155

weighted avg 0.40 0.40 0.40 155