```
import warnings
warnings.filterwarnings("ignore")
import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
from tensorflow.keras.models import Sequential, Model
from tensorflow.keras.layers import
Dense, Activation, Dropout, Conv2D, Flatten, MaxPool2D, Reshape, GlobalAveragePooling2D, Input
Layer
from tensorflow.keras.applications.resnet50 import preprocess input
from tensorflow. keras. preprocessing import image
from tensorflow. keras. preprocessing. image import
ImageDataGenerator, load_img, img_to_array
from tensorflow.keras.callbacks import EarlyStopping,ReduceLROnPlateau
IMAGE_SIZE=[229, 229]
train path='../input/natural-disaster-intensity/dataset/train set'
test_path='../input/natural-disaster-intensity/dataset/test_set'
train data gen=ImageDataGenerator(rescale=1./255, shear range=0.2, zoom range=0.2, horizo
ntal flip=True, validation split=0.30)
test data gen=ImageDataGenerator(rescale=1./255, validation split=0.30)
training set=train data gen. flow from directory (train path, target size=(229, 229), batch
_size=100, class_mode='categorical', shuffle=True, color_mode='rgb', subset='training')
testing_set=test_data_gen.flow_from_directory(test_path, target_size=(229, 229), batch_si
ze=100, class mode='categorical', shuffle=True, color mode='rgb', subset='validation')
Found 521 images belonging to 4 classes.
Found 58 images belonging to 4 classes.
model=Sequential()
model.add(Conv2D(32, (3, 3), input shape=(229, 229, 3), activation='relu'))
model.add(MaxPool2D(pool size=(2,2)))
model.add(Conv2D(32, (3, 3), activation='relu'))
model.add(MaxPool2D(pool size=(2,2)))
model.add(Flatten())
model.add(Dense(units=128, activation='relu'))
model.add(Dense(units=4, activation='softmax'))
model.summary()
Model: "sequential 18"
Layer (type)
                              Output Shape
                                                         Param #
conv2d_6 (Conv2D)
                              (None, 227, 227, 32)
                                                         896
```

```
max pooling2d 2 (MaxPooling2 (None, 113, 113, 32)
conv2d_7 (Conv2D)
                          (None, 111, 111, 32)
                                                  9248
max pooling2d 3 (MaxPooling2 (None, 55, 55, 32)
                                                  0
flatten 1 (Flatten)
                          (None, 96800)
                                                  ()
dense (Dense)
                          (None, 128)
                                                  12390528
dense 1 (Dense)
                          (None, 4)
                                                  516
Total params: 12,401,188
Trainable params: 12,401,188
Non-trainable params: 0
model.compile(optimizer='adam', loss='categorical_crossentropy', metrics=['accuracy'])
model.fit generator(
   generator=training_set, steps_per_epoch=len(training_set),
   epochs=20, validation_data=testing_set, validation_steps=len(testing_set))
Epoch 1/20
6/6 [================== ] - 40s 6s/step - loss: 7.4400 - accuracy: 0.2745 -
val loss: 4.7555 - val accuracy: 0.1379
Epoch 2/20
6/6 [================= ] - 30s 6s/step - loss: 1.6385 - accuracy: 0.3186 -
val_loss: 1.4745 - val_accuracy: 0.3103
Epoch 3/20
val_loss: 1.3408 - val_accuracy: 0.4138
Epoch 4/20
6/6 [========
                  ========] - 31s 5s/step - loss: 1.2657 - accuracy: 0.4088 -
val_loss: 1.2716 - val_accuracy: 0.6034
Epoch 5/20
6/6 [================= ] - 31s 5s/step - loss: 1.1531 - accuracy: 0.5125 -
val_loss: 1.2474 - val_accuracy: 0.4138
Epoch 6/20
6/6 [================== ] - 31s 5s/step - loss: 1.0124 - accuracy: 0.5336 -
val_loss: 1.1508 - val_accuracy: 0.5862
Epoch 7/20
6/6 [================== ] - 31s 5s/step - loss: 0.8680 - accuracy: 0.6468 -
val_loss: 1.0622 - val_accuracy: 0.6034
Epoch 8/20
6/6 [======= ] - 30s 6s/step - loss: 0.8234 - accuracy: 0.6449 -
val loss: 0.9700 - val accuracy: 0.5862
Epoch 9/20
6/6 [=================== ] - 30s 5s/step - loss: 0.7506 - accuracy: 0.7179 -
val loss: 0.8130 - val accuracy: 0.7586
Epoch 10/20
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6/6 [================== ] - 31s 5s/step - loss: 0.7150 - accuracy: 0.7351 -
val loss: 0.9505 - val accuracy: 0.7241
Epoch 11/20
6/6 [================= ] - 30s 5s/step - loss: 0.7464 - accuracy: 0.7102 -
val loss: 1.0083 - val accuracy: 0.5517
Epoch 12/20
6/6 [================== ] - 30s 5s/step - loss: 0.6719 - accuracy: 0.7562 -
val_loss: 0.7746 - val_accuracy: 0.7586
Epoch 13/20
6/6 [============] - 30s 6s/step - loss: 0.5585 - accuracy: 0.7946 -
val loss: 0.9316 - val accuracy: 0.7414
Epoch 14/20
6/6 [===========] - 30s 5s/step - loss: 0.5536 - accuracy: 0.7774 -
val_loss: 1.1241 - val_accuracy: 0.6552
Epoch 15/20
6/6 [==================] - 31s 6s/step - loss: 0.5133 - accuracy: 0.8157 -
val_loss: 1.0897 - val_accuracy: 0.6897
Epoch 16/20
6/6 [=============] - 33s 6s/step - loss: 0.5082 - accuracy: 0.8081 -
val_loss: 0.9379 - val_accuracy: 0.7069
Epoch 17/20
6/6 [================= ] - 33s 6s/step - loss: 0.4945 - accuracy: 0.8157 -
val loss: 1.0962 - val accuracy: 0.6379
Epoch 18/20
val_loss: 1.2129 - val_accuracy: 0.5862
Epoch 19/20
6/6 [================= ] - 33s 6s/step - loss: 0.4238 - accuracy: 0.8196 -
val loss: 1.0063 - val accuracy: 0.7241
Epoch 20/20
6/6 [=================== ] - 33s 5s/step - loss: 0.3634 - accuracy: 0.8676 -
val loss: 1.0335 - val accuracy: 0.7586
model. save ("nature. h5")
from tensorflow.keras.models import load model
from keras. preprocessing import image
model = load model("nature.h5")
a = ['Cyclone', 'Earthquake', 'Flood', 'Wildfire']
img=image.load_img('.../input/natural-disaster-
intensity/dataset/test set/Cyclone/876.jpg',
                target size=(229, 229))
x=image.img_to_array(img)
x=np. expand dims(x, axis=0)
pred=np. argmax (model. predict(x))
a[pred]
'Cyclone'
testing set. class indices
{'Cyclone': 0, 'Earthquake': 1, 'Flood': 2, 'Wildfire': 3}
```