

# multiclass

June 28, 2024

```
[ ]: import tensorflow as tf
      from tensorflow import keras
      from tensorflow.keras import layers
      from tensorflow.keras.preprocessing.image import ImageDataGenerator
      # Define image size and batch size
      IMG_SIZE = 224
      BATCH_SIZE = 32
```

```
[ ]:
```

```
[ ]: # Define data generators for train, validation and test sets
      train_datagen = ImageDataGenerator(rescale=1./255,validation_split=0.2)

      train_generator = train_datagen.flow_from_directory(
          r"/content/drive/MyDrive/weather/Multi-class Weather Dataset",
          target_size=(IMG_SIZE, IMG_SIZE),
          batch_size=BATCH_SIZE,
          class_mode='categorical',
          subset='training'
      )

      val_generator = train_datagen.flow_from_directory(
          r"/content/drive/MyDrive/weather/Multi-class Weather Dataset",
          target_size=(IMG_SIZE, IMG_SIZE),
          batch_size=BATCH_SIZE,
          class_mode='categorical',
          subset='validation'
      )
```

Found 901 images belonging to 4 classes.

Found 224 images belonging to 4 classes.

```
[ ]: # Get the class indices from the training generator
      class_indices = train_generator.class_indices

      # Extract class names
      class_names = list(class_indices.keys())
```

```
print("Class indices:", class_indices)
print("Class names:", class_names)
```

```
Class indices: {'Cloudy': 0, 'Rain': 1, 'Shine': 2, 'Sunrise': 3}
Class names: ['Cloudy', 'Rain', 'Shine', 'Sunrise']
```

```
[ ]: # Define a Sequential model
model = keras.Sequential([
    layers.Conv2D(32, (3,3), activation='relu',
    ↪input_shape=(IMG_SIZE,IMG_SIZE,3)),
    layers.MaxPooling2D((2,2)),
    layers.Conv2D(64, (3,3), activation='relu'),
    layers.MaxPooling2D((2,2)),
    layers.Conv2D(128, (3,3), activation='relu'),
    layers.MaxPooling2D((2,2)),
    layers.Flatten(),
    layers.Dense(128, activation='relu'),
    layers.Dense(4, activation='softmax')
])
```

```
[ ]: # Compile the model
model.compile(optimizer='adam', loss='categorical_crossentropy',
    ↪metrics=['accuracy'])
```

```
[ ]: model.fit(train_generator, validation_data=val_generator, epochs=10)
```

```
Epoch 1/10
29/29 [=====] - 190s 7s/step - loss: 0.6077 - accuracy:
0.7858 - val_loss: 0.7867 - val_accuracy: 0.7455
Epoch 2/10
29/29 [=====] - 118s 4s/step - loss: 0.3147 - accuracy:
0.8868 - val_loss: 0.5195 - val_accuracy: 0.8170
Epoch 3/10
29/29 [=====] - 118s 4s/step - loss: 0.2620 - accuracy:
0.8990 - val_loss: 0.3383 - val_accuracy: 0.8616
Epoch 4/10
29/29 [=====] - 118s 4s/step - loss: 0.2210 - accuracy:
0.9245 - val_loss: 0.4685 - val_accuracy: 0.8304
Epoch 5/10
29/29 [=====] - 113s 4s/step - loss: 0.1719 - accuracy:
0.9390 - val_loss: 0.3048 - val_accuracy: 0.8839
Epoch 6/10
29/29 [=====] - 115s 4s/step - loss: 0.1277 - accuracy:
0.9600 - val_loss: 0.4090 - val_accuracy: 0.8527
Epoch 7/10
29/29 [=====] - 118s 4s/step - loss: 0.1093 - accuracy:
```

```

0.9567 - val_loss: 0.6851 - val_accuracy: 0.8527
Epoch 8/10
29/29 [=====] - 113s 4s/step - loss: 0.1326 - accuracy:
0.9534 - val_loss: 0.3721 - val_accuracy: 0.8482
Epoch 9/10
29/29 [=====] - 117s 4s/step - loss: 0.0985 - accuracy:
0.9656 - val_loss: 0.3870 - val_accuracy: 0.8482
Epoch 10/10
29/29 [=====] - 115s 4s/step - loss: 0.0591 - accuracy:
0.9834 - val_loss: 0.2817 - val_accuracy: 0.8795

```

```
[ ]: <keras.src.callbacks.History at 0x78afb1de9b40>
```

```
[ ]: model.save('weather.h5')
```

```

/usr/local/lib/python3.10/dist-packages/keras/src/engine/training.py:3103:
UserWarning: You are saving your model as an HDF5 file via `model.save()`. This
file format is considered legacy. We recommend using instead the native Keras
format, e.g. `model.save('my_model.keras')`.
    saving_api.save_model(

```

```

[ ]: from tensorflow.keras.models import load_model
    from tensorflow.keras.preprocessing import image
    import numpy as np
    model = load_model("/content/drive/MyDrive/weather/weather.h5")
    print("Model Loaded")

```

Model Loaded

```

[ ]: # Load and view the image
    from matplotlib import pyplot as plt
    test_image_path = r"/content/drive/MyDrive/weather/Multi-class Weather Dataset/
    ↪Rain/rain1.jpg"
    img = image.load_img(test_image_path, target_size=(224, 224))

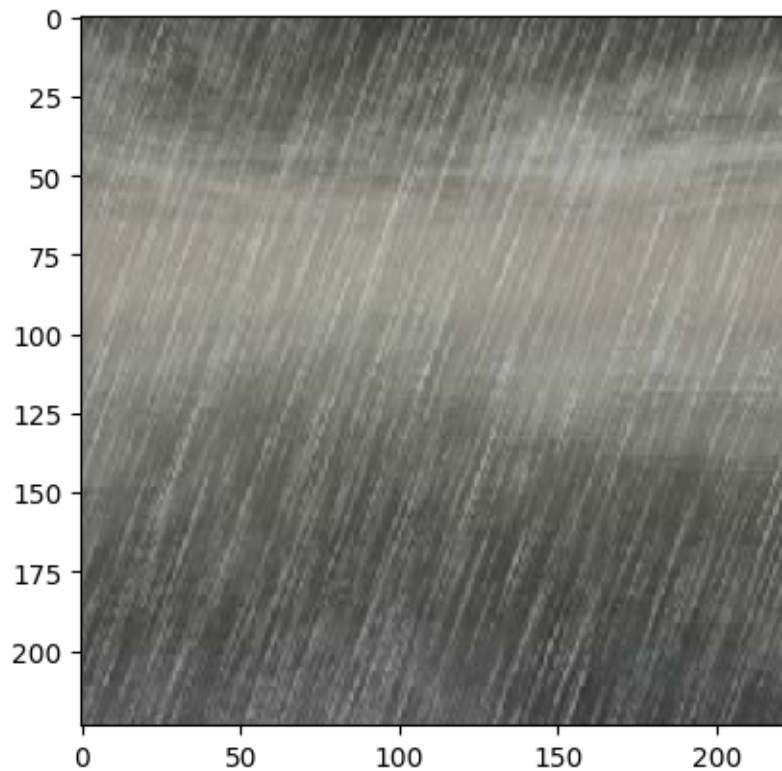
    plt.imshow(img)
    plt.axis()
    plt.show()

    #convert image into array
    img_array = image.img_to_array(img)
    img_array = np.expand_dims(img_array, axis=0)
    img_array /= 255. # Normalize the pixel values

    # Make predictions
    prediction = model.predict(img_array)
    # Print the prediction

```

```
print(prediction)
```



```
1/1 [=====] - 0s 86ms/step  
[[6.9388881e-04 9.9921572e-01 8.3931773e-05 6.4895585e-06]]
```

```
[ ]: #interpret the results  
prediction = model.predict(img_array)  
ind = np.argmax(prediction[0])  
print(class_names[ind])
```

```
1/1 [=====] - 0s 52ms/step  
Cloudy
```