

PROJECT CODES

Exploratory Data Analysis :

```
from collections import Counter
from collections import defaultdict

[28]: #read the dataset
path = ""data/"

>>

[29]: names = [name.replace('_', '_').split('_')[0] for name in os.listdir(path)]

classes Counter (names) #return dictionary

[30]: print(classes)

Counter({'arecaceae': 35, 'arrabidaea': 35, 'cecropia': 35, 'chromolaena': 35, 'combretum': 35,
'croton': 35, 'dipteryx': 35,
'eucalipto': 35, 'farama a': 35, 'hyptis': 35, 'mabea': 35, 'matayba': 35, 'mimosa': 35, 'myrcia': 35,
'protium': 35,
'qualea': 35, 'schinus': 35, 'senegalia': 35, 'serjania': 35, 'syagrus': 35, 'tridax': 35, 'urochloa': 35,
'anadenanthera': 20})

#Total no. of images
print("Number of images", len(names))

Number of images 790

import matplotlib.pyplot as plt
plt.figure(figsize = (8,3))
plt.title('Class Counts in Dataset')
plt.bar(*zip(*classes.items()))
plt.xticks(rotation='vertical')
plt.show()
```

```
path_class = {key: [] for key in classes.keys()}
```

```
for name in os.listdir(path):
```

```
key = name.replace(' ', '_').split('_')[0]
```

```
path_class[key].append(path + name)
```

```
import matplotlib.pyplot as plt
```

```
from PIL import Image
```

```
fig = plt.figure(figsize=(15, 15))
```

```
for i, key in enumerate(path_class.keys()):
```

```
    img1 = Image.open(path_class[key][0])
```

```
    img2 = Image.open(path_class[key][1])
```

```
    img3 = Image.open(path_class[key][2])
```

```
ax = fig.add_subplot(8, 9, 3*i + 1, xticks=[], yticks=[])
```

```
ax.imshow(img1)
```

```
ax.set_title(key)
```

```
ax = fig.add_subplot(8, 9, 3*i + 2, xticks=[], yticks=[])
```

```
ax.imshow(img2)
```

```
ax.set_title(key)
```

```
ax = fig.add_subplot(8, 9, 3*i + 3, xticks=[], yticks=[])
```

```
ax.imshow(img3)
```

```
ax.set_title(key)
```

```
plt.tight_layout()
```

```
plt.show()
```

```

import os

import cv2

import matplotlib.pyplot as plt

# Read sizes (height, width) from all images in the directory
sizes = [cv2.imread(os.path.join(path, name)).shape[:2]
          for name in os.listdir(path)]

x, y = zip(*sizes) # separate heights and widths

# Plotting scatter plot
fig = plt.figure(figsize=(5, 5))

plt.scatter(x, y)

plt.title("Image size scatterplot")

plt.xlabel("Height (pixels)")

plt.ylabel("Width (pixels)")

# Add diagonal red line for reference
max_val = max(max(x), max(y))

plt.plot([0, max_val], [0, max_val], 'r')

plt.show()

```

Image PreProcessing

```

import os

import cv2

import numpy as np

from sklearn.preprocessing import LabelEncoder

from keras.utils import np_utils

from sklearn.model_selection import train_test_split


def process_img(img, size=(128, 128)):

    img = cv2.resize(img, size) # resize image

    img = img / 255.0          # divide values by 25

    return img

```

```

# Read all images and store in X; Y holds class names
X, Y = [], []

for name in os.listdir(path):
    img = cv2.imread(os.path.join(path, name))
    X.append(process_img(img))
    Y.append(name.replace(' ', '_').split('_')[0])

X = np.array(X)

# Encode labels and convert to one-hot encoding
le = LabelEncoder()
Y_le = le.fit_transform(Y)
Y_cat = np_utils.to_categorical(Y_le, num_classes=23)

# Split data into training and testing sets
X_train, X_test, Y_train, Y_test = train_test_split(
    X, Y_cat, test_size=0.285, stratify=Y_le
)

print("Images in each class in Test set: {}".format(np.sum(Y_test, axis=0)))

# Save arrays to disk
np.save("data/processed/X_train.npy", X_train)
np.save("data/processed/X_test.npy", X_test)
np.save("data/processed/Y_train.npy", Y_train)
np.save("data/processed/Y_test.npy", Y_test)

```

NOW MODEL BUILDING

```
import numpy as np
```

```
# Load your preprocessed dataset
```

```
X_train = np.load("data/processed/X_train.npy")
```

```
X_test = np.load("data/processed/X_test.npy")
```

```
Y_train = np.load("data/processed/Y_train.npy")
```

```
Y_test = np.load("data/processed/Y_test.npy")
```

```
# Define the model (continuing from previous code)
```

```
from keras.models import Sequential
```

```
from keras.layers import Conv2D, MaxPooling2D, Flatten, Dropout, Dense
```

```
input_shape = X_train[0].shape
```

```
output_shape = Y_train.shape[1] # automatically picks your number of classes
```

```
from keras.models import Sequential
```

```
from keras.layers import Conv2D, MaxPooling2D, Flatten, Dropout, Dense
```

```
input_shape = X_train[0].shape
```

```
output_shape = 23
```

```
model = Sequential()
```

```
model.add(Conv2D(filters=16, kernel_size=3, input_shape=input_shape, activation='relu',  
padding='same'))
```

```
model.add(MaxPooling2D(pool_size=2))
```

```
model.add(Conv2D(filters=32, kernel_size=2, activation='relu', padding='same'))
```

```
model.add(MaxPooling2D(pool_size=2))
```

```
model.add(Conv2D(filters=64, kernel_size=2, activation='relu', padding='same'))
```

```
model.add(MaxPooling2D(pool_size=2))
```

```
model.add(Conv2D(filters=128, kernel_size=2, activation='relu', padding='same'))
model.add(MaxPooling2D(pool_size=2))
```

```
model.add(Flatten())
model.add(Dropout(0.2))
model.add(Dense(500, activation='relu'))
# model.add(Dropout(0.2))
model.add(Dense(150, activation='relu'))
# model.add(Dropout(0.2))
model.add(Dense(output_shape, activation='softmax'))
```

```
model.summary()
```

```
#Compile Model
```

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

```
# Data augmentation setup
```

```
datagener = ImageDataGenerator(
    rotation_range=20,
    width_shift_range=0.2,
    height_shift_range=0.2,
    horizontal_flip=True,
    vertical_flip=True
)
```

```
datagener.fit(X_train)
```

```
# Training parameters
```

```
batch_size = 4
```

```
epochs = 500
```

```

model_path = 'cnn.hdf5'

callbacks = [
    EarlyStopping(monitor='val_loss', patience=20, restore_best_weights=True),
    ModelCheckpoint(filepath=model_path, save_best_only=True)
]

# Fit the model using augmented data
history = model.fit(
    datagener.flow(X_train, Y_train, batch_size=batch_size),
    steps_per_epoch=len(X_train) // batch_size,
    epochs=epochs,
    validation_data=(X_test, Y_test),
    callbacks=callbacks,
    verbose=1
)

# Load the best weights and evaluate
model.load_weights(model_path)

score = model.evaluate(X_test, Y_test, verbose=0)
print(f"Test set accuracy: {score[1]:.4f}")

# Plot Customized Accuracy and Loss Graphs

import matplotlib.pyplot as plt

# Extract data from training history
acc = history.history['accuracy']      # Blue: generated (augmented) training accuracy
val_acc = history.history['val_accuracy'] # Orange: original validation accuracy
loss = history.history['loss']         # Blue: training loss
val_loss = history.history['val_loss']  # Orange: validation loss

```

```
epochs = range(1, len(acc) + 1)
```

```
# Plot Model Accuracy
```

```
plt.figure(figsize=(10, 5))
```

```
plt.plot(epochs, acc, 'b-', label='Generated Data (Train Accuracy)') # Blue line
```

```
plt.plot(epochs, val_acc, 'orange', label='Original Data (Validation Accuracy)') # Orange line
```

```
plt.title('Model Accuracy')
```

```
plt.xlabel('Epoch')
```

```
plt.ylabel('Accuracy')
```

```
plt.xticks([0, 20, 40, 60, 80])
```

```
plt.yticks([0.0, 0.2, 0.4, 0.6, 0.8])
```

```
plt.legend()
```

```
plt.grid(True)
```

```
plt.show()
```

```
# Plot Model Loss
```

```
plt.figure(figsize=(10, 5))
```

```
plt.plot(epochs, loss, 'b-', label='Generated Data (Train Loss)') # Blue line
```

```
plt.plot(epochs, val_loss, 'orange', label='Original Data (Validation Loss)') # Orange line
```

```
plt.title('Model Loss (Categorical Crossentropy)')
```

```
plt.xlabel('Epoch')
```

```
plt.ylabel('Loss')
```

```
plt.xticks([0, 20, 40, 60, 80])
```

```
plt.yticks([0.0, 0.5, 1.0, 1.5, 2.0, 2.5, 3.0])
```

```
plt.legend()
```

```
plt.grid(True)
```

```
plt.show()
```



```
#Save Model

model.save('model.h5')


# Testing the Model


import numpy as np

from tensorflow.keras.preprocessing.image

import load_img, img_to_array

import os


# 1. Define your image path

img_path = r'D:\Pollen_Grain\data\protium_23.jpg'


# 2. Check if file exists

if not os.path.exists(img_path):

    raise FileNotFoundError(f"Image not found: {img_path}")


# 3. Load & resize image (converted to RGB with 3 channels)

img = load_img(img_path, target_size=(128, 128))


# 4. Convert to array & normalize

x = img_to_array(img)

x = x / 255.0

x = np.expand_dims(x, axis=0) # shape: (1, 128, 128, 3)


# 5. Predict with your existing model

pred = model.predict(x) # shape: (1, num_classes)

a = np.argmax(pred, axis=1)[0]
```

6. Define your class labels in the correct order

```
op = [  
    'areceae', 'anadenanthera', 'arrabidaea', 'cecropia', 'chromolaena',  
    'combretum', 'croton', 'dipter',  
    # ... include all other class names in training order ...  
    'eucalipto'  
]
```

7. Map index to class label and display

```
result = op[a]  
print("Predicted class:", result)
```

import numpy as np

from tensorflow.keras.preprocessing.image import load_img, img_to_array

import os

1. Define your image path

```
img_path = r'D:\Pollen_Grain\data\serjania_23.jpg'
```

2. Check if file exists

```
if not os.path.exists(img_path):  
    raise FileNotFoundError(f"Image not found: {img_path}")
```

3. Load & resize image (converted to RGB with 3 channels)

```
img = load_img(img_path, target_size=(128, 128))
```

4. Convert to array & normalize

```
x = img_to_array(img)
```

```
x = x / 255.0
```

```
x = np.expand_dims(x, axis=0) # shape: (1, 128, 128, 3)
```

```
# 5. Predict with your existing model
```

```
pred = model.predict(x) # shape: (1, num_classes)
```

```
a = np.argmax(pred, axis=1)[0]
```

```
# 6. Define your class labels in the correct order
```

```
op = [
```

```
    'arecaceae', 'anadenanthera', 'arrabidaea', 'cecropia', 'chromolaena',
```

```
    'combretum', 'croton', 'dipter',
```

```
    # ... include all other class names in training order ...
```

```
    'eucalipto'
```

```
]
```

```
# 7. Map index to class label and display
```

```
result = op[a]
```

```
print("Predicted class:", result)
```

TEMPLATES

#index.html

```
<!-- Paste into templates/index.html -->
<!DOCTYPE html>
<html lang="en">
<head>
  <meta charset="UTF-8">
  <meta name="viewport" content="width=device-width, initial-scale=1.0">
  <title>Pollen Grain Classifier</title>
  <style>
    body {
      background: linear-gradient(to right, #f6f9fc, #e9eff5);
      font-family: Arial, sans-serif;
      text-align: center;
      padding: 40px;
    }
    .container {
      background: #fff;
      max-width: 600px;
      margin: auto;
      padding: 30px;
      border-radius: 15px;
      box-shadow: 0 4px 8px rgba(0,0,0,0.1);
    }
    h1 {
      margin-bottom: 20px;
      color: #2c3e50;
    }
    form {
```

```

        margin-top: 20px;
    }
    input[type="file"] {
        margin: 20px 0;
    }
    button {
        background-color: #2980b9;
        color: white;
        padding: 10px 20px;
        border: none;
        border-radius: 5px;
        cursor: pointer;
        font-size: 16px;
    }
    button:hover {
        background-color: #1f6391;
    }
    footer {
        margin-top: 50px;
        font-size: 14px;
        color: #7f8c8d;
    }
</style>
</head>

<body>
    <div class="container">
        <h1>🌾 Pollen Grain Classification</h1>
        <!-- 🔄 CRITICAL CHANGE: Update form action -->
        <form action="/result" method="POST" enctype="multipart/form-data">
            <label for="file">Upload a pollen grain image:</label><br>

```

```
        <input type="file" name="image" id="file" required><br>
        <button type="submit">Predict</button>
    </form>
</div>
<footer>
    <p>&copy; 2025 Pollen Project | Team 7</p>
</footer>
</body>
</html>
```

#logout.html

```
<!-- Paste into templates/logout.html -->
<!DOCTYPE html>
<html lang="en">
<head>
    <meta charset="UTF-8">
    <title>Logged Out</title>
    <style>
        body {
            font-family: Arial, sans-serif;
            background-color: #f0f8ff;
            display: flex;
            flex-direction: column;
            align-items: center;
            justify-content: center;
            height: 100vh;
            margin: 0;
        }
```

```

    h1 {
        color: #333;
    }
    a {
        margin-top: 20px;
        text-decoration: none;
        color: #fff;
        background-color: #007BFF;
        padding: 10px 20px;
        border-radius: 5px;
    }
    a:hover {
        background-color: #0056b3;
    }
</style>
</head>
<body>
    <h1>You have been logged out.</h1>
    <a href="/">Return to Home</a>
</body>
</html>

```

prediction.html

```

<!-- Paste into templates/prediction.html -->
<!DOCTYPE html>
<html lang="en">
<head>
    <meta charset="UTF-8">
    <title>Prediction Result</title>

```

```
<style>

body {

    font-family: 'Segoe UI', sans-serif;

    text-align: center;

    background-color: #f2f9ff;

    padding: 40px;

}

h1 {

    color: #333;

}

img {

    margin-top: 20px;

    width: 300px;

    height: auto;

    border: 5px solid #007BFF;

    border-radius: 10px;

}

.result {

    font-size: 24px;

    margin-top: 30px;

    color: #007BFF;

}

.buttons {

    margin-top: 40px;

}

.buttons a {

    display: inline-block;

    margin: 0 10px;

    text-decoration: none;

    padding: 10px 25px;

    color: white;
```



```
        background-color: #007BFF;

        border-radius: 5px;
    }

    .buttons a:hover {
        background-color: #0056b3;
    }
</style>
</head>

<body>

    <h1>Prediction Result</h1>

    {% if image_path %}
        
    {% endif %}

    <div class="result">

        Predicted Class: <strong>{{ pred }}</strong>

    </div>

    <div class="buttons">

        <a href="/prediction.html"> Try Another</a>

        <a href="/logout.html"> Logout</a>

    </div>
</body>
</html>
```

```
# app.py

import re

import os

import numpy as np

import pandas as pd

from flask import Flask, request, render_template

from keras.models import load_model

from tensorflow.keras.preprocessing.image import load_img, img_to_array


# Load model

model = load_model("model/model_.h5", compile=False)


# Initialize Flask app

app = Flask(__name__)


# Home routes

@app.route('/')

def index():

    return render_template("index.html")


@app.route('/index.html')

def home():

    return render_template("index.html")


@app.route('/logout.html')

def logout():

    return render_template("logout.html")


@app.route('/prediction.html')

def prediction():
```

```

return render_template("prediction.html")

# Prediction result
@app.route('/result', methods=["POST"])
def res():
    if 'image' not in request.files:
        return render_template("prediction.html", pred="No file uploaded")

    f = request.files['image']
    if f.filename == '':
        return render_template("prediction.html", pred="No file selected")

    # Save uploaded image
    basepath = os.path.dirname(__file__)
    upload_folder = os.path.join(basepath, 'static', 'upload')
    os.makedirs(upload_folder, exist_ok=True)
    filepath = os.path.join(upload_folder, f.filename)
    f.save(filepath)

    # Preprocess image
    img = load_img(filepath, target_size=(128, 128))
    x = img_to_array(img) / 255.0
    x = np.expand_dims(x, axis=0)

    # Predict
    pred = np.argmax(model.predict(x))

    op = [
        'anadenanthera', 'arecaceae', 'arrabidaea', 'cecropia', 'chromolaena', 'combretum',
        'croton', 'dipteryx', 'eucalpto', 'faramea', 'hyptis', 'mabea', 'matayba', 'mimosa',
        'myrcia', 'protium', 'qualea', 'schinus', 'senegalia', 'serjania', 'syagrus',
        'tridax', 'urochloa'
    ]

```

```
]
```

```
result = op[pred]
```

```
return render_template('prediction.html', pred=result)
```

```
# Run Flask app
```

```
if __name__ == "__main__":
```

```
    app.run(debug=True)
```