

Prepare and explore the Data set, (Fruits Classes by kagle)

this note book is made to prepare and explore the data set

```
In [2]: import numpy as np
import matplotlib.pyplot as plt
```

Install The data set

install the data set from kaggle hub into the local host move it from the default locatoion into the current directory

Library:

Animals-10:

<https://www.kaggle.com/datasets/alessiocorrado99/animals10>

took from it the spiders photos

Animal Image Dataset:

<https://www.kaggle.com/datasets/ashishsaxena2209/animal-image-datasetdog-cat-and-panda>

took from it the cats and pandas photos

```
In [ ]: import kagglehub

# Download latest version
path = kagglehub.dataset_download("alessiocorrado99/animals10")
print("Path to dataset files:", path)
```

```
In [ ]: import shutil
target_path = r".."

# Move the whole dataset folder
shutil.move(path, target_path)

print("Moved dataset to:", target_path)
```

```
In [ ]: import kagglehub

# Download latest version
path = kagglehub.dataset_download("ashishsaxena2209/animal-image-datasetdog-")

print("Path to dataset files:", path)
```

```
In [ ]: import shutil
        target_path = r"."

        # Move the whole dataset folder
        shutil.move(path, target_path)

        print("Moved dataset to:", target_path)
```

Displaying Some Images

```
In [3]: from PIL import Image
        target_path = r"."

        # Define image paths for each class
        cat_img_path = fr"{target_path}/Animals/cats/cats_00038.jpg"

        panda_img_path = fr"{target_path}/Animals/panda/panda_00034.jpg"

        spider_img_path = fr"{target_path}/Animals/spiders/e830b2062dfc023ed1584d05f

        # Open images
        cat_img = Image.open(cat_img_path)
        panda_img = Image.open(panda_img_path)
        spider_img = Image.open(spider_img_path)

        # Plot them side by side
        plt.figure(figsize=(12, 4))

        # Cat
        plt.subplot(1, 3, 1)
        plt.imshow(cat_img)
        plt.title("Label: cat")
        plt.axis('off')

        # Dog
        plt.subplot(1, 3, 2)
        plt.imshow(panda_img)
        plt.title("Label: panda")
        plt.axis('off')

        # Panda
        plt.subplot(1, 3, 3)
        plt.imshow(spider_img)
        plt.title("Label: spider")
        plt.axis('off')

        plt.tight_layout()
        plt.show()
```



Load images features

here we load each image in each class, then prepare them and save into `.npy` files to load them quickly later on

```
In [4]: from PIL import Image
        from skimage.feature import hog

        def extract_combined_features(image_path, image_size=(64, 64)):
            try:
                # Open the image from the given file path and resize it
                image_rgb = Image.open(image_path).resize(image_size)

                # Convert the image to grayscale for HOG feature extraction
                image_gray = image_rgb.convert('L')

                # Convert the grayscale image to a NumPy array
                gray_array = np.array(image_gray)

                # HOG (Histogram of Oriented Gradients) Features
                # Extract texture and edge information from the grayscale image
                hog_features = hog(
                    gray_array,
                    orientations=9, # Number of orientation bins
                    pixels_per_cell=(8, 8), # Size of the cell in pixels
                    cells_per_block=(2, 2), # Number of cells per block
                    block_norm='L2-Hys' # Normalization method
                )

                # Convert the RGB image to a NumPy array
                rgb_array = np.array(image_rgb)

                # Color Histogram Features
                # Compute normalized histograms for R, G, B channels separately
                hist_features = []
                for i in range(3): # Loop over Red, Green, Blue channels
                    hist, _ = np.histogram(
                        rgb_array[:, :, i], # Select the color channel
                        bins=32, # Number of bins for the histogram
                        range=(0, 256), # Pixel intensity range
                        density=True # Normalize the histogram
                    )
                    hist_features.extend(hist) # Add histogram data to the list
```

```

        # Combine HOG and Histogram Features
        combined = np.concatenate((hog_features, hist_features))

    return combined

except Exception as e:
    # Print and skip any image that causes an error
    print(f"Skipping {image_path}: {e}")
    return None

```

```

In [5]: import os

def load_images_with_combined_features(folder_path, image_size=(64, 64), limit_per_class=10):
    X, y = [], []
    label_map = {}
    current_label = 0

    for class_name in sorted(os.listdir(folder_path)): # List the three classes
        class_path = os.path.join(folder_path, class_name) # For each folder
        if not os.path.isdir(class_path): # if the path doesn't exist, ignore
            continue

        label_map[current_label] = class_name
        count = 0

        # for all images in this class
        for filename in os.listdir(class_path):
            if filename.lower().endswith(('.png', '.jpg', '.jpeg')):
                if count >= limit_per_class: # take the same range for all classes
                    break
                image_path = os.path.join(class_path, filename)
                features = extract_combined_features(image_path, image_size=(image_size[0], image_size[1]))
                if features is not None:
                    X.append(features)
                    y.append(current_label)
                    count += 1

        current_label += 1

    return np.array(X), np.array(y), label_map

```

Save the pre-processed data - Training

save the data that we processed to an appropriate extension to load them later

```

In [6]: import pickle
X, y, label_map = load_images_with_combined_features(r"Animals")

np.save("X.npy", X) #save
np.save("y.npy", y) #save
with open("label_map.pkl", "wb") as f:
    pickle.dump(label_map, f) #write

```

Skipping Animals\panda\panda_00049.jpg: too many indices for array: array is 2-dimensional, but 3 were indexed
Skipping Animals\panda\panda_00347.jpg: too many indices for array: array is 2-dimensional, but 3 were indexed
Skipping Animals\panda\panda_00723.jpg: too many indices for array: array is 2-dimensional, but 3 were indexed
Skipping Animals\panda\panda_00781.jpg: too many indices for array: array is 2-dimensional, but 3 were indexed
Skipping Animals\panda\panda_00895.jpg: too many indices for array: array is 2-dimensional, but 3 were indexed
Skipping Animals\panda\panda_00914.jpg: too many indices for array: array is 2-dimensional, but 3 were indexed
Skipping Animals\panda\panda_00923.jpg: too many indices for array: array is 2-dimensional, but 3 were indexed

Data Set features summary

```
In [7]: X = np.load("X.npy")
y = np.load("y.npy")

with open("label_map.pkl", "rb") as f:
    label_map = pickle.load(f)

print(f"Total images loaded: {len(X)}")
print(f"Shape of each image vector: {X[0].shape}")
print(f"Label distribution: {np.bincount(y)}")
print(f"Label map: {label_map}")
```

```
Total images loaded: 2993
Shape of each image vector: (1860,)
Label distribution: [1000  993 1000]
Label map: {0: 'cats', 1: 'panda', 2: 'spiders'}
```

```
In [ ]:
```