

notebook

May 31, 2024

1 Part I: Dataset loading & preprocessing

```
[ ]: !rm -rf /content/drive/MyDrive/cv_hw2_data
```

```
[ ]: !wget -c https://rgbd-dataset.cs.washington.edu/dataset/rgbd-dataset/
      ↪rgbd-dataset.tar
      !tar -xf rgbd-dataset.tar
      !rm rgbd-dataset.tar
```

```
--2024-05-28 12:03:03-- https://rgbd-dataset.cs.washington.edu/dataset/rgbd-
dataset/rgbd-dataset.tar
```

```
Resolving rgbd-dataset.cs.washington.edu (rgbd-dataset.cs.washington.edu)...
```

```
128.208.3.117, 2607:4000:200:12:3eec:efff:fe5e:6f68
```

```
Connecting to rgbd-dataset.cs.washington.edu (rgbd-
dataset.cs.washington.edu)|128.208.3.117|:443... connected.
```

```
HTTP request sent, awaiting response... 200 OK
```

```
Length: 5067799243 (4.7G) [application/x-tar]
```

```
Saving to: 'rgbd-dataset.tar'
```

```
rgbd-dataset.tar    100%[=====>]    4.72G  19.4MB/s    in 2m 53s
```

```
2024-05-28 12:05:56 (28.0 MB/s) - 'rgbd-dataset.tar' saved
```

```
[5067799243/5067799243]
```

```
[ ]: import os
import cv2
import shutil
import random
import matplotlib.pyplot as plt

def create_dataset(src_folder, dest_folder, selected_categories,
    ↪image_size=(224, 224)):
    # Ensure the destination folder exists
    if not os.path.exists(dest_folder):
        os.makedirs(dest_folder)
```

```

for category in selected_categories:
    # Source folder for the first instance of the category
    obj_folder = os.path.join(src_folder, category, f"{category}_1")
    dest_category_folder = os.path.join(dest_folder, category)

    # Ensure the destination category folder exists
    if not os.path.exists(dest_category_folder):
        os.makedirs(dest_category_folder)

    for filename in os.listdir(obj_folder):
        # Process only the cropped RGB images
        if filename.endswith('_crop.png'):
            img_path = os.path.join(obj_folder, filename)
            img = cv2.imread(img_path)

            if img is not None:
                # Resize the image to a fixed size if specified
                if image_size:
                    img = cv2.resize(img, image_size)

                # Save the loaded and resized image to the destination
                ↪ folder
                dest_path = os.path.join(dest_category_folder, filename)
                cv2.imwrite(dest_path, img)

# Function to display random images from each category
def display_random_images(dest_folder, selected_categories):
    plt.figure(figsize=(15, 10))

    for i, category in enumerate(selected_categories):
        category_folder = os.path.join(dest_folder, category)
        if os.path.exists(category_folder):
            images = [img for img in os.listdir(category_folder) if img.
            ↪ endswith('_crop.png')]
            if images:
                random_image = random.choice(images)
                img_path = os.path.join(category_folder, random_image)
                img = cv2.imread(img_path)
                img_rgb = cv2.cvtColor(img, cv2.COLOR_BGR2RGB) # Convert BGR
                ↪ to RGB for display

                plt.subplot(2, 5, i+1)
                plt.imshow(img_rgb)
                plt.title(category)
                plt.axis('off')

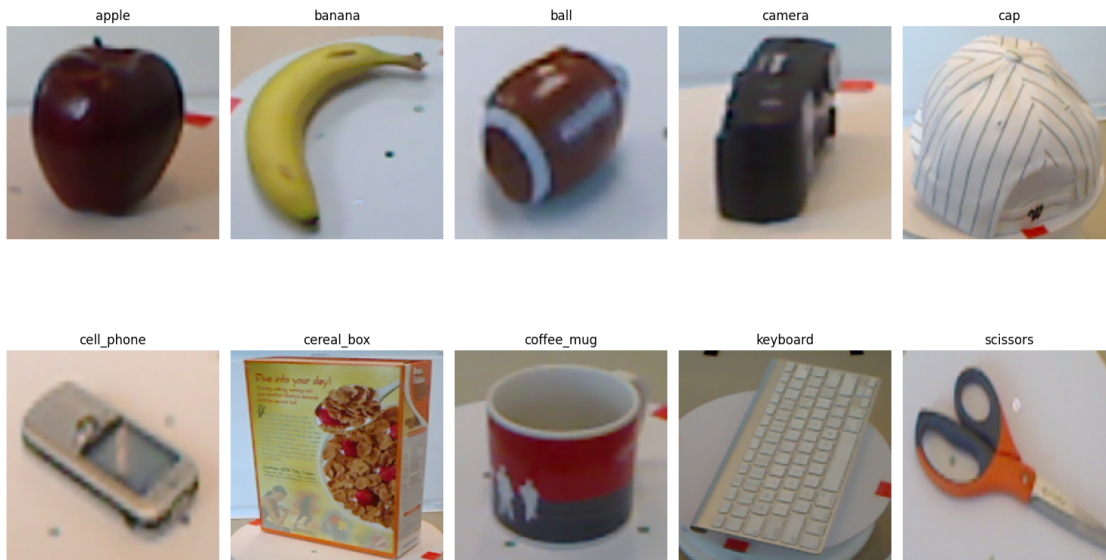
    plt.tight_layout()

```

```
plt.show()
```

```
[ ]: src_folder = '/content/rgbd-dataset'
dest_folder = '/content/drive/MyDrive/cv_hw2_data'
selected_categories = ['apple', 'banana', 'ball', 'camera', 'cap',
    ↪ 'cell_phone', 'cereal_box', 'coffee_mug', 'keyboard', 'scissors']
create_dataset(src_folder, dest_folder, selected_categories)

# Display random images from each category
display_random_images(dest_folder, selected_categories)
```



2 Part II: Object recognition based on feature detector and descriptor

```
[4]: import numpy as np
from sklearn.base import BaseEstimator, TransformerMixin
from sklearn.cluster import KMeans
from sklearn.preprocessing import StandardScaler

class FeatureExtractor(BaseEstimator, TransformerMixin):
    def __init__(self, n_clusters=100, method='SIFT'):
        self.n_clusters = n_clusters
        self.method = method
        self.kmeans = KMeans(n_clusters=self.n_clusters)
        self.scaler = StandardScaler()
        self.detector = self.init_detector()
```

```

def init_detector(self):
    if self.method == 'SIFT':
        return cv2.SIFT_create()
    elif self.method == 'ORB':
        return cv2.ORB_create()
    elif self.method == 'BRISK':
        return cv2.BRISK_create()
    else:
        raise ValueError(f"Unsupported method: {self.method}")

def fit(self, X, y=None):
    features = self.extract_features(X)
    self.kmeans.fit(features)
    return self

def transform(self, X):
    # Transform the images into histograms of visual words
    histograms = self.create_histograms(X)
    # Standardize the histograms
    return self.scaler.fit_transform(histograms)

def extract_features(self, images):
    descriptors_list = []
    for img in images:
        if self.method == 'SIFT' or self.method == 'ORB' or self.method ==   

↪ 'BRISK':
            keypoints, descriptors = self.detector.detectAndCompute(img,   

↪ None)
            if descriptors is not None:
                descriptors_list.extend(descriptors)
            else:
                raise ValueError(f"Unsupported method: {self.method}")
    return np.array(descriptors_list)

def create_histograms(self, images):
    histograms = []
    for img in images:
        if self.method == 'SIFT' or self.method == 'ORB' or self.method ==   

↪ 'BRISK':
            keypoints, descriptors = self.detector.detectAndCompute(img,   

↪ None)
            else:
                raise ValueError(f"Unsupported method: {self.method}")
            if descriptors is not None:
                # Predict the cluster for each descriptor and create a histogram
                histogram = np.zeros(self.n_clusters)
                cluster_result = self.kmeans.predict(descriptors)

```

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        for i in cluster_result:
            histogram[i] += 1
        histograms.append(histogram)
    else:
        histograms.append(np.zeros(self.n_clusters))
    return np.array(histograms)

```

```

[ ]: import os
import cv2
import numpy as np
import random
from sklearn.model_selection import train_test_split

def load_dataset(folder, n_samples=100, seed=42):
    images = []
    labels = []

    random.seed(seed)

    for obj in os.listdir(folder):
        obj_folder = os.path.join(folder, obj)
        # Randomly select images from the current directory
        filenames = os.listdir(obj_folder)
        random.shuffle(filenames)
        for filename in filenames[:n_samples]:
            img_path = os.path.join(obj_folder, filename)
            img = cv2.imread(img_path)
            if img is not None:
                images.append(img)
                labels.append(obj)
    return np.array(images), np.array(labels)

```

```

[ ]: dataset_folder = '/content/drive/MyDrive/cv_hw2_data'
selected_categories = ['apple', 'banana', 'ball', 'camera', 'cap', '
    ↪ 'cell_phone', 'cereal_box', 'coffee_mug', 'keyboard', 'scissors']

images, labels = load_dataset(dataset_folder, n_samples=200)

# Split the dataset into training and testing sets
X_train, X_test, y_train, y_test = train_test_split(images, labels, test_size=0.
    ↪ 1, stratify=labels, random_state=42)

```

```

[ ]: import warnings
warnings.filterwarnings("ignore")

from sklearn.metrics import confusion_matrix
import seaborn as sns

```

```

import matplotlib.pyplot as plt
from sklearn.pipeline import make_pipeline
from sklearn.svm import SVC
from sklearn.metrics import accuracy_score

# Function to display confusion matrix
def display_confusion_matrix(y_true, y_pred, categories):
    cm = confusion_matrix(y_true, y_pred)
    plt.figure(figsize=(10, 8))
    sns.heatmap(cm, annot=True, fmt='d', cmap='Blues', xticklabels=categories,
    ↪yticklabels=categories)
    plt.xlabel('Predicted Labels')
    plt.ylabel('True Labels')
    plt.title('Confusion Matrix')
    plt.show()

# Train and evaluate the classifier for each feature extraction method
methods = ['SIFT', 'ORB', 'BRISK']

# Loop through each feature extraction method
for method in methods:
    # Create a pipeline with FeatureExtractor and SVM classifier
    # FeatureExtractor: extracts features from images using the specified method
    # SVC: SVM classifier with a linear kernel
    pipeline = make_pipeline(FeatureExtractor(method=method),
    ↪SVC(kernel='linear'))

    # Fit the pipeline on the training data
    pipeline.fit(X_train, y_train)

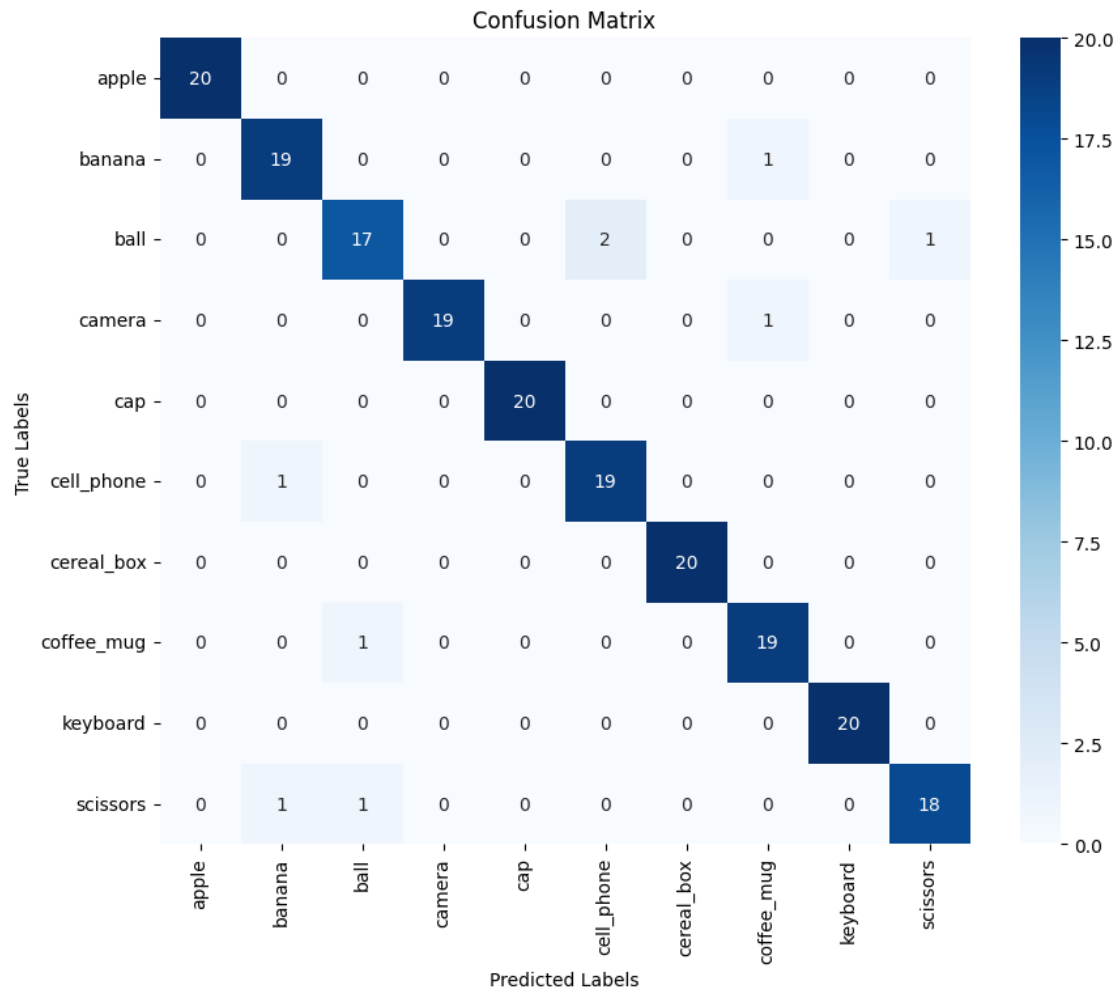
    # Predict the labels for the test data
    y_pred = pipeline.predict(X_test)

    accuracy = accuracy_score(y_test, y_pred)
    print(f"Accuracy using {method}: {accuracy:.2f}")

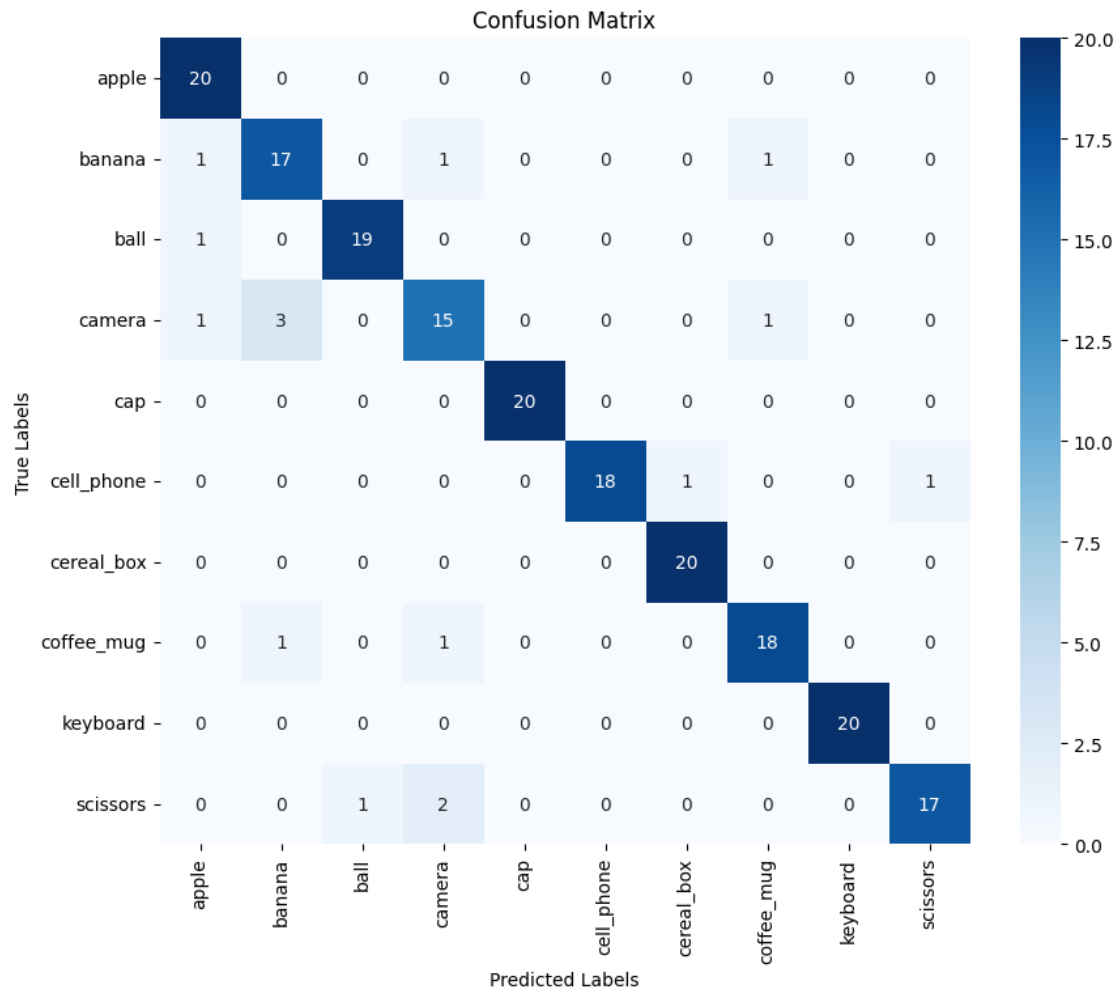
    display_confusion_matrix(y_test, y_pred, selected_categories)

```

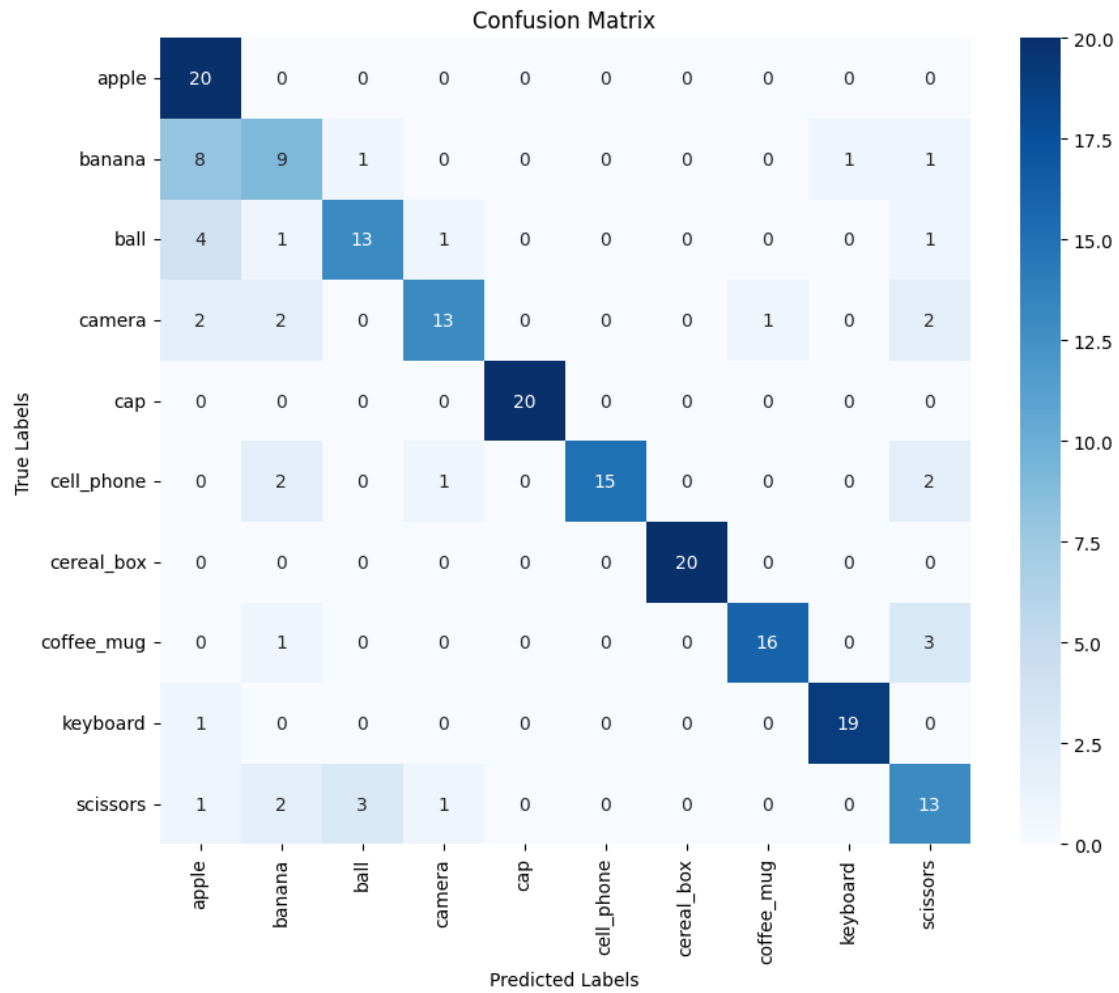
Accuracy using SIFT: 0.95



Accuracy using ORB: 0.92



Accuracy using BRISK: 0.79



[]: