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| **S.NO** | **Name of the Experiment** | **Hours/CO Mapped** |
|  | **UNIT I** | **15 hours** |
| **1.1** | Write a program to implement various feature extraction techniques for image classification. | **CO1** |
| **1.2** | Write a program to assess various feature matching algorithms for object recognition. | **CO2** |
| **1.3** | Write a program to analyze the impact of refining feature detection for image segmentation. | **CO3** |
| **1.4** | Write a program to evaluate the efficacy of human-guided control point selection for image alignment. | **CO4** |
|  | **UNIT II** | **15 hours** |
| **2.1** | Write a program to compare the performance of different classification models in image recognition. | **CO5** |
| **2.2** | Write a program to interpret the effectiveness of Bag of Features in enhancing image classification performance. | **CO1** |
| **2.3** | Write a program to analyze various object detection algorithms with machine learning. | **CO2** |
|  | **UNIT III** | **15 hours** |
| **3.1** | Write a program to determine the effectiveness of incorporating optical flow analysis into object tracking algorithms. | **CO3** |
| **3.2** | Write a program to examine the performance of various pretrained deep learning models for real-time object tracking tasks. | **CO5** |
| **3.3** | Write a program to interpret the effectiveness of template matching techniques for video stabilization tasks. | **CO4** |

**EXPERIMENT NO-1.1**

**AIM:** Write a program to implement various feature extraction techniques for image classification.

**SOFTWARE REQUIRED:** Any Python IDE (e.g., PyCharm, Jupyter Notebook, Google Collab)

**RELEVANCE OF THE EXPERIMENT:** This experiment is relevant because feature extraction is a fundamental step in image classification tasks. By implementing various feature extraction techniques, students can gain hands-on experience in preprocessing images and extracting meaningful features that can be used for classification purposes. This experiment also allows students to understand the importance of selecting appropriate feature extraction techniques based on the characteristics of the dataset and the classification problem at hand.

**DESCRIPTION:**

There are several feature extraction techniques commonly used in image classification tasks. These techniques aim to capture relevant information from images and transform them into meaningful representations that can be used by machine learning algorithms for classification. Some popular feature extraction techniques are:

* Scale-Invariant Feature Transform (SIFT): SIFT is a widely used technique that identifies key points and extracts local invariant descriptors from images. It is robust to changes in scale, rotation, and illumination.
* Speeded-Up Robust Features (SURF): SURF is another technique that detects and describes local features in images. It is similar to SIFT but computationally more efficient, making it suitable for real-time applications.
* Histogram of Oriented Gradients (HOG): HOG computes the distribution of gradient orientations in an image. It captures the shape and edge information and has been particularly successful in object detection and pedestrian recognition tasks.
* Convolutional Neural Networks (CNN): CNNs are a type of deep learning model that automatically learn hierarchical features from images. They consist of multiple convolutional layers that extract low-level to high-level features. CNNs have revolutionized image classification and achieved state-of-the-art performance in various tasks.
* Color Histograms: Color histograms capture the distribution of colors in an image. They represent the color content of images by quantizing pixel colors into bins and counting their occurrences. Color histograms are simple yet effective features for certain types of image classification problems.
* Local Binary Patterns (LBP): LBP encodes the texture information by comparing each pixel's intensity value with its neighboring pixels. It is commonly used in texture analysis tasks and has shown good performance in various image classification
* n applications.
* Gabor Filters: Gabor filters are a set of linear filters that capture localized frequency and orientation information in images. They are commonly used for texture analysis and have been successfully applied in face recognition and fingerprint recognition tasks.
* Deep Convolutional Features: Instead of using pre-defined feature extraction techniques, it is also common to use pre-trained CNN models (e.g., VGG, ResNet, Inception) and extract features from intermediate layers. These deep convolutional features retain more high-level semantics and have been shown to generalize well across different image classification tasks.

The experiment involves implementing different feature extraction techniques for image classification using Python and relevant libraries. The steps involved in this experiment are as follows:

1. Import necessary libraries (e.g., OpenCV, scikit-image).
2. Load the dataset of labeled images for training and testing.
3. Preprocess the images by resizing, normalizing, or applying any necessary transformations.
4. Extract features from the images using various techniques such as:

* Histogram of Oriented Gradients (HOG)
* Scale-Invariant Feature Transform (SIFT)
* Speeded-Up Robust Features (SURF)
* Local Binary Patterns (LBP)
* Convolutional Neural Networks (CNN)
* etc.

1. Split the dataset into training and testing sets.
2. Train a classifier (e.g., Support Vector Machine, Random Forest, etc.) using the extracted features and the corresponding labels.
3. Evaluate the performance of the classifier on the testing set by calculating metrics like accuracy, precision, recall, and F1-score.
4. Compare the performance of different feature extraction techniques by analyzing the evaluation results.
5. Repeat steps 4-8 for different combinations of feature extraction techniques and classifiers to explore the impact on classification performance.
6. Document the observations and conclusions drawn from the experiment.