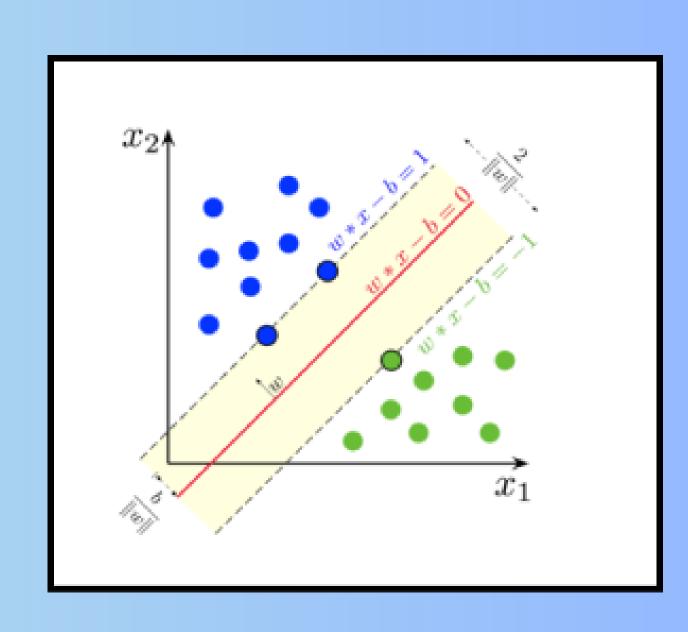
# Human Activity Recognition using Kernelized SVMs

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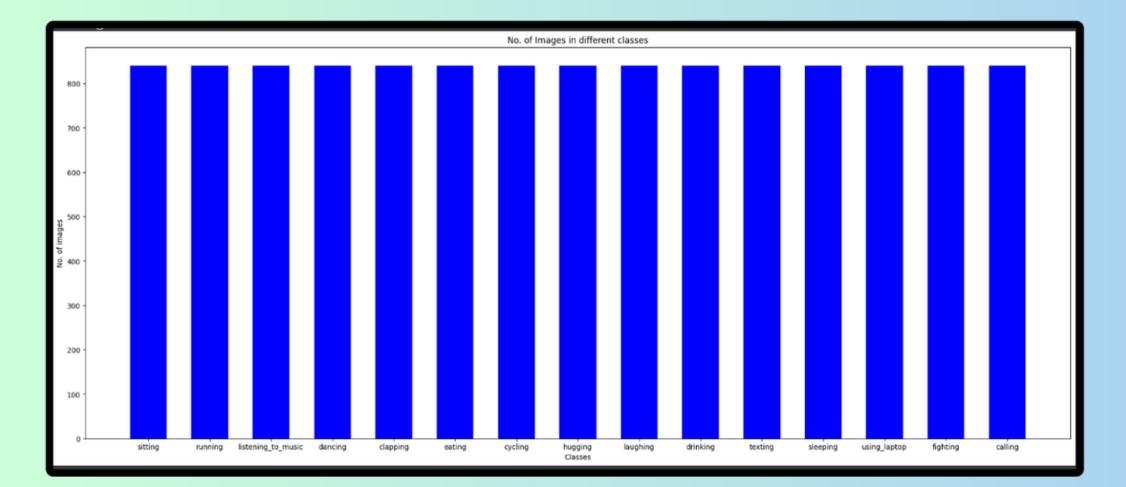


#### **Problem Statement**

Build an Image Classification Model using SVMs that classifies to which class of activity a human is performing

#### **About the Dataset**

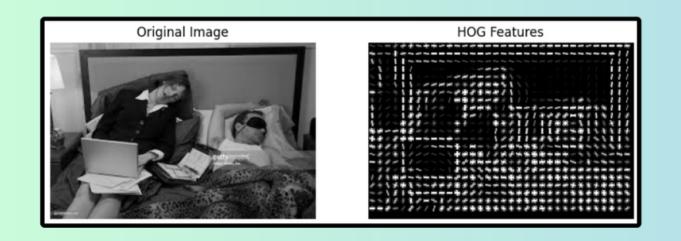
- The dataset features 15 different classes of Human Activities.
- The dataset contains about 15k+ labelled images including the validation images.
- Each image has only one human activity category and are saved in separate folders of the labelled classes



All the 15 classes in the HAR dataset have the same number of images, hence, there will be no class imbalance based on the no.

of images per class

#### EDA and Data Preprocessing

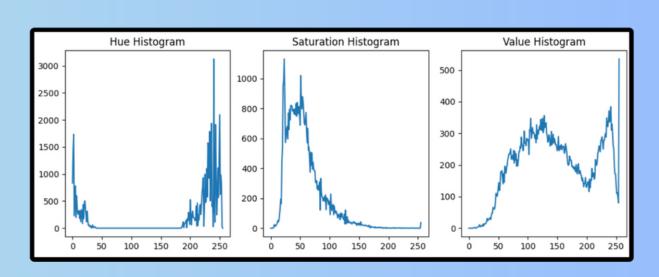


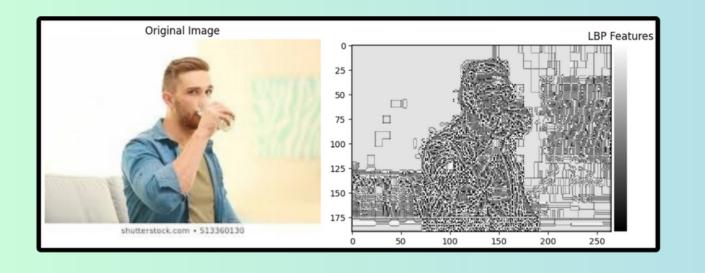
#### HISTOGRAM OF ORIENTED GRADIENTS

HOG, is a feature descriptor which is used in for the purpose of object detection. The technique counts occurrences of gradient orientation in the localized portion of an image. For the regions of the image it generates histograms using the magnitude and orientations of the gradient.

#### **HSV FEATURES**

The HSV features are well-suited for color analysis and feature extraction. By separating color information into hue, saturation, and value components, you can perform tasks like color-based object detection and tracking. This can be done by using colour histograms.





#### LOCAL BINARY PATTERNS

LBPs compute a local representation of texture. This is constructed by looking at points surrounding a central point and testing whether the surrounding points are greater than or less than the central point (i.e. gives a binary result).

#### EDA and Data Preprocessing - II



#### **SOBELS FILTER**

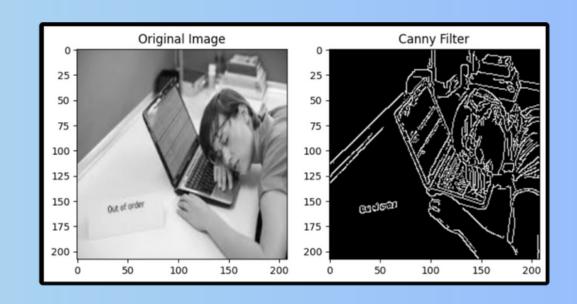
Sobel's Edge detection filter is used to detect edges in the images.

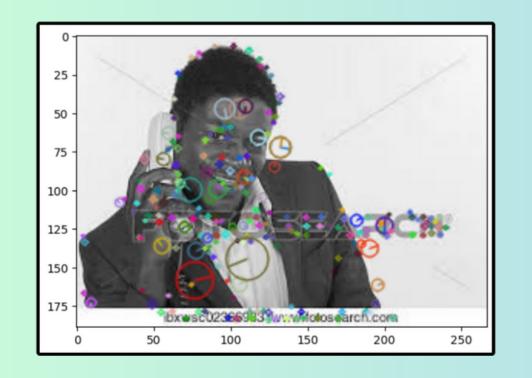
These can be converted to statistical features like mean, median etc.

or flattened to convert into feature vectors.

#### **CANNY FILTER**

The Canny edge detector is an edge detection operator that uses a multistage algorithm to extract useful structural information from different vision objects and dramatically reduce the amount of data to be processed.



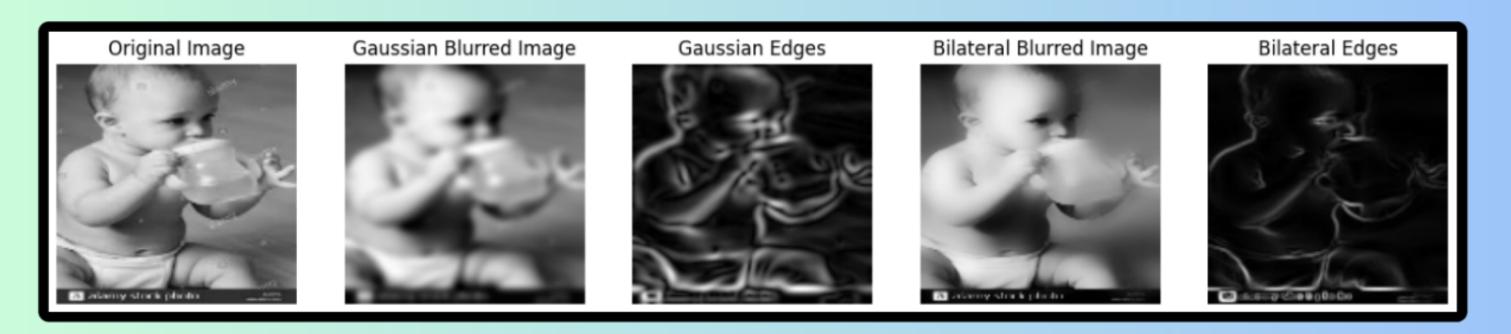


#### SIFT FEATURES

The SIFT features are local and based on the appearance of the object at particular interest points, and are invariant to image scale and rotation. They are also robust to changes in illumination, noise, and minor changes in viewpoint.

#### EDA and Data Preprocessing - III

#### GAUSSIAN AND BILATERAL FILTER FOR NOISE REMOVAL



The Gaussian filter and bilateral filter are both commonly employed for noise removal in image processing, but they differ in their approaches and outcomes. The Gaussian filter, uniformly blurs an image, effectively reducing noise but potentially sacrificing edge details. On the other hand, the bilateral filter, excels in preserving edges by considering both spatial proximity and intensity differences.

#### OTHER TECHNIQUES

- Min-max scaling: For a constant scale as SVMs are sensitive to the scale of the features
- Template Matching / Patch Similarity: It finds predefined patterns in images, aiding in object recognition and feature localization.
- Contours: It identify edges and shapes in images, pivotal for object detection, recognition, and boundary analysis.
- Median filters: Has high capablity of reducing noise, due to which edges got lost.

#### METHODOLOGY AND RESULTS

#### 1) HOG + HSV + LBP features

We extracted HOG features from grayscale images and concatenated them with HSV features from coloured images. After this we joined them with LBP features to get a total of around 4500 features. This combined feature matrix was passed to the SVM classifier where we used Polynomial kernel with a degree of 6 for optimal results.

• Min- max scaler boosted accuracy by 4%

**Accuracy on Cross validation: 35%** 

#### 2) <u>Bilateral Filter (for noise removal)</u>, <u>Sobel filter (for BG seperation)</u> <u>then HOG + HSV + LBP features</u>

Here we reduced all the extra noise and then used sobel filter to highlight only imp edges in the image. All HOG feature extraction done on this was expected to increase accuracy by a lot, because of less noise. However, the accuracy reduced.

Accuracy on Cross validation: 30%

#### METHODOLOGY AND RESULTS-II

#### 3) Ensemble of HOG, HSV

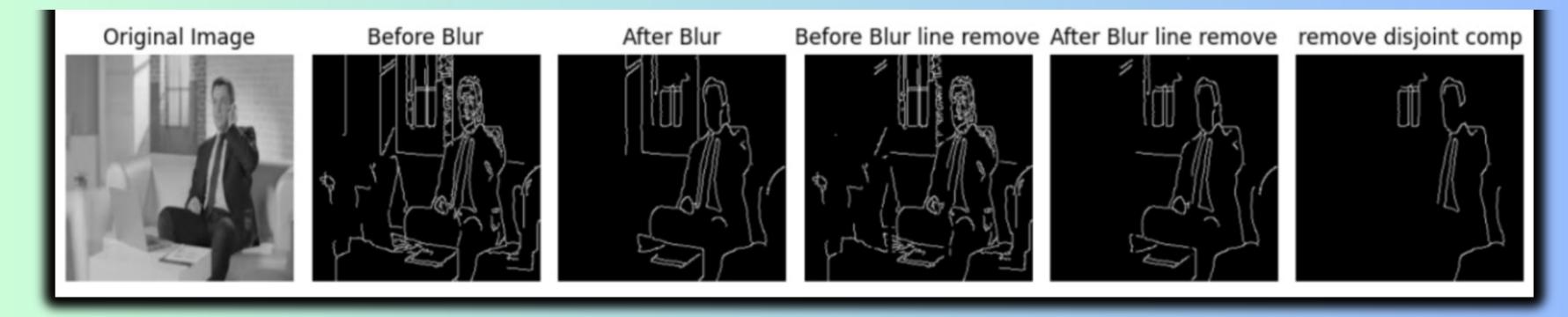
This time we created an ensemble of HOG and HSV features . 2 SVMs was trained on scaled versions of the two independently. Then at the time of classification we calculated the confidence score of the two SVMs the one which gave better was finally chosen for prediction.

Accuracy on Cross validation: 31%

#### 4) SIFT FEATURES + HSV + LBP

SIFT, HSV, and LBP combined for richer image analysis, leveraging keypoints, color details, and local textures. Their fusion enhanced model understanding for more nuanced image interpretation. We tried with only one row of SIFT features for each image because it was giving extra features for each iteration.

Accuracy on Cross validation: 25 %



#### METHODOLOGY AND RESULTS-III

## Results on Different Kernels Used (for the best model)

• Linear Kernel : 28%

• RBF Kernel (Gaussian) : 34%

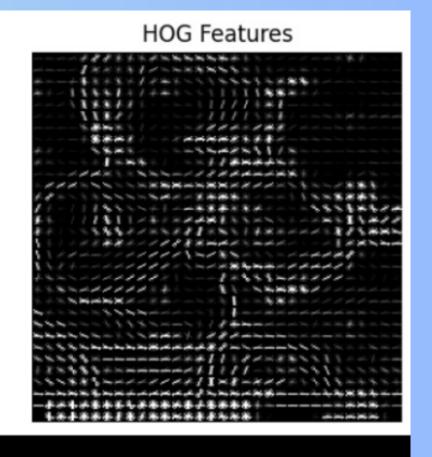
• Polynomial Kernel (deg-6) : 35%

• Sigmoid Kernel : 14%

#### Other Techniques Used

- Increased the No. of HOG features and used mean, var to reduce dimensionality (PCA not working as expected)
- Z-score Normalization









#### PROBLEMS AND CONCLUSION

- Noise in the dataset: Traditional feature extraction methods involve handcrafted operations on the image. While these methods (filters and thresholds) may capture certain local features, they do not perform noise removal in a learned and adaptive way as they are manual. Ties in the dataset added to the woes.
- Imperfect Feature extraction: Handcrafted techniques like HOG, HSV, and LBP have limited capacity to represent and generalize complex and high-dimensional patterns inherent in human activities.
- Thus, traditional ML techniques and Non-Deep learning-based feature engineering do not lead to high accuracy in this task. Existing analysis also showed a majority of DL techniques (like OpenPose, CNNs, and RNNs being used to solve this problem. CNNs are capable of learning hierarchical representations of data and reducing noise inherently due to multiple layers.

Note: Even using CNNs only an accuracy of 70% could be achieved

### THANK YOU!