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## HYPERSPECTRAL IMAGE CLASSIFICATION USING MACHINE LEARNING

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### ABSTRACT:

Hyperspectral image (HSI) classification is a phenomenal mechanism to analyze diversified land cover in remotely sensed hyperspectral images. Remote sensing technology is improving day by day which has also increased the uses of hyperspectral imaging tremendously. Exact classification of ground features from hyperspectral images is an important and a popular research area and also has attracted widespread attention. In our research, a brief description among different classification models i.e., SVM, with PCA, has been described. The study has been carried upon one common hyperspectral datasets i.e., Indian Pines which comprise various landscape fields like dense vegetation, barren land, grasslands, etc. For noisy band reduction, PCA has been used.

### 1. INTRODUCTION

Imaging is a good way to work with materials, identify them and to define their properties is to study how light interacts with them. Spectroscopy is a study that examines how light behaves in the target & recognizes materials based on their different spectral Signatures. A human eye can see electromagnetic waves with wavelengths between 380 and 780 nanometers. Wavelengths beyond this range such as Infrared are invisible to humans and also human has 3 color receptors red, blue and green but Hyperspectral Imaging is a new analytical technique based on spectroscopy that measures the continuous spectrum of the light for each pixel of the scene with fine wavelength resolution, not only in the visible but also in the near infrared. Hyperspectral imaging (HSI) is a method for capturing images that contain information from a broader portion of the electromagnetic spectrum. This portion can start with UV light, extend through the visible spectrum, and end in the near or short-wave infrared. This extended

wavelength range can reveal properties of material composition that are not otherwise apparent. Hyperspectral Image (HSI) classification is a phenomenal mechanism to analyze diversified land cover in remotely sensed hyperspectral images. Remote sensing technology is improving day by day which has also increased the uses of hyperspectral imaging tremendously. Exact classification of ground features from hyperspectral images is an important and a popular research area and also have attracted a widespread of attention. A good classification results are achieved by many methods for the classification of hyperspectral

Imaging. Purpose Hyperspectral imaging, like other spectral imaging, collects and processes information from across the electromagnetic spectrum. The goal of hyperspectral imaging is to obtain the spectrum for each pixel in the image of a scene, with the purpose of finding objects, identifying materials, or detecting processes. Hyperspectral imaging (HSI) integrates conventional imaging and spectroscopy, to obtain both spatial and spectral information from a specimen. This

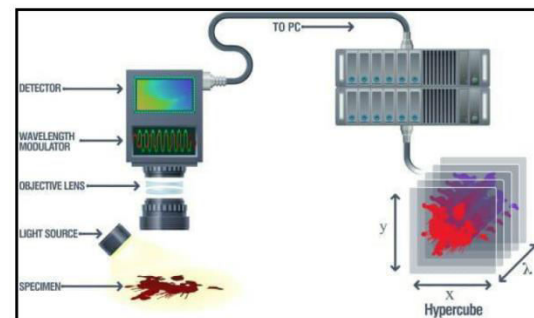
technique enables investigators to analyze the chemical composition of traces and simultaneously visualize their spatial distribution. HSI offers significant potential for the detection, visualization, identification and age estimation of forensic traces. The rapid, non-destructive and non-contact features of HSI mark its suitability as an analytical tool for forensic science.

The availability of hyperspectral remote sensing images with high spatial and spectral resolution have begun to be available in recent years. As these hyperspectral images have strong angular resolution for a spectra, these images have a wide range of applications such as in the environment, mining, military and medical fields. Imaging spectrometers are the main sources for the acquisition of hyperspectral images which are installed at particular places. The imaging spectrometer acquires images in many continuous and narrow bands due to which each pixel of the image gets a spectrum of reflected wavelength range which was used during acquiring of image. Hence hyperspectral images have high spectral resolution with many bands and plentiful data.

## 2. RELATED WORK

Hyperspectral images provide rich spectral and spatially continuous information that can be used for soil mineralogy discrimination. This paper proposes a method to evaluate the feasibility of Hyperion image in the rapid prediction of soil mineralogy. Four areas in Egypt were chosen for the current study. Preprocessing of the Hyperion data was done before applying the atmospheric correction. The minimum noise fraction transformation was used to segregate noise in the data. Various techniques were applied to the

studied areas in which mixture tuned matched filtering gave good results in a prediction of the end-members. Then, it is employed to predict soil minerals in each cell using a spectral unmixing method. Illite, chlorite, calcite, dolomite, As an emerging imaging modality for medical applications, HSI offers great potential for noninvasive disease diagnosis and surgical guidance. Light delivered to biological tissue undergoes multiple scattering from inhomogeneity of biological structures and absorption primarily in hemoglobin, melanin, and water as it propagates through the tissue. It is assumed that the absorption, fluorescence, and scattering characteristics of tissue change during the progression of disease. Therefore, the reflected, fluorescent, and transmitted

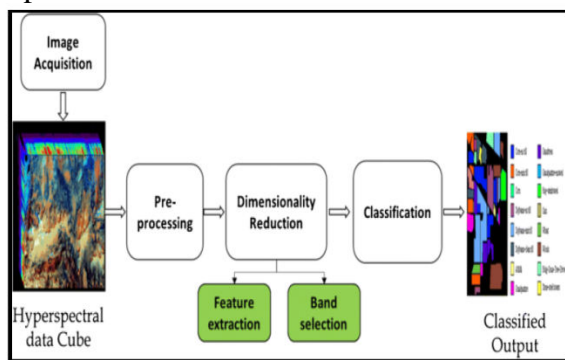


light from tissue captured by HSI carries quantitative diagnostic information about tissue pathology. Hyperspectral imaging (HSI) integrates conventional imaging and spectroscopy, to obtain both spatial and spectral information from a specimen. This technique enables investigators to analyze the chemical composition of traces and simultaneously visualize their spatial distribution. HSI offers significant potential for the detection, visualization, identification and age estimation of forensic traces. The rapid, non-destructive and non-contact features of HSI mark its suitability as an analytical tool for forensic science. The preliminary investigation

examines project feasibility, the likelihood the application will be useful to the user. The main objective of the feasibility study is to test the Technical, Operational, and Economical feasibility for adding new modules and debugging traditional desktop-centric applications, and porting them to mobile devices. All systems are feasible if they are given unlimited resources and infinite time.

### 3. METHODOLOGY:

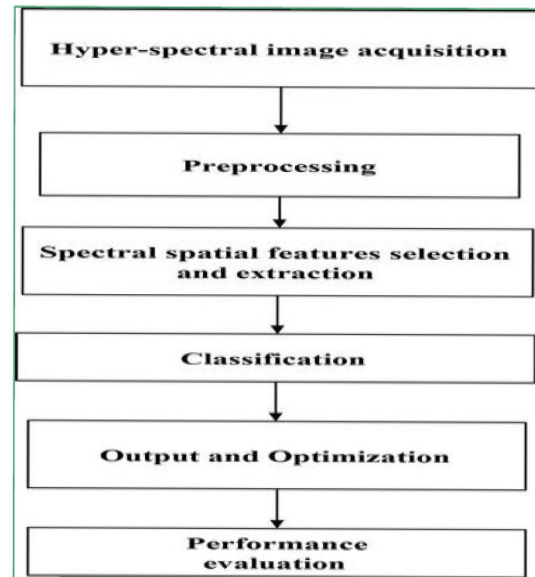
System design is the process or art of defining the architecture, components, modules, interfaces, and data for a system to satisfy specified requirements. One could see it as the application of systems theory to product development. There is some overlap and synergy with the disciplines of systems analysis, systems architecture, and systems engineering. This is the architecture that describes the Hyperspectral image classification process with a step by step representation.



**Figure 1: Architecture**

#### Data flow diagram

This explains the way of representing the flow of data through a process or a system. It provides information about the outputs and inputs of each entity and process itself.



**Figure 2: Data flow diagram**

Support vector Machine is a set of supervised learning methods used for classification, regression and outliers detection. The advantages of support vector machines are: Effective in high dimensional spaces. Still effective in cases where the number of dimensions is greater than the number of samples. Uses a subset of training points in the decision function (called support vectors), so it is also memory efficient. Versatile: different Kernel functions can be specified for the decision function. Common kernels are provided, but it is also possible to specify custom kernels. The disadvantages of support vector machines include: If the number of features is much greater than the number of samples, avoiding over-fitting in choosing Kernel functions and regularization terms is crucial. SVMs do not directly provide probability estimates, these are calculated using an expensive five-fold cross-validation (see Scores and probabilities, below). The support vector machines in scikit-learn support both dense (numpy.ndarray

and convertible to that by numpy.asarray) and sparse (any scipy.sparse) sample vectors as input. However, to use an SVM to make predictions for sparse data, it must have benefit on such data. For optimal performance, use C-ordered numpy.ndarray (dense) or scipy.sparse.csr\_matrix (sparse) with dtype=float64.

#### 4. STUDY OF RESULTS:

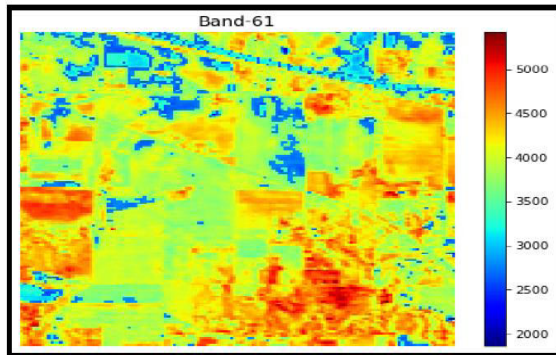


Figure 3: Visualizing the bands of the Hyperspectral image

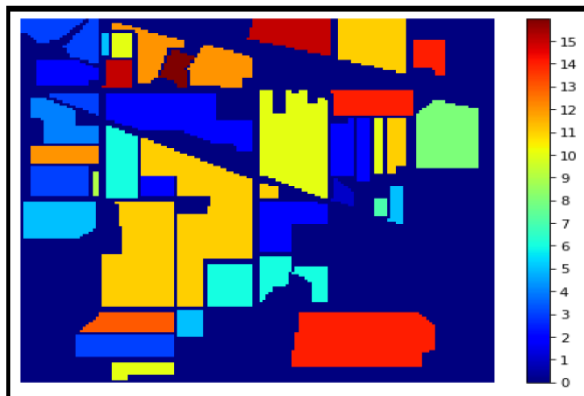


Figure 4: Visualizing ground truth of the image.

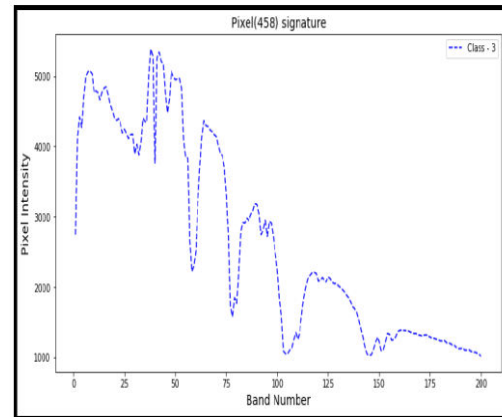


Figure 5: Visualizing spectral signatures

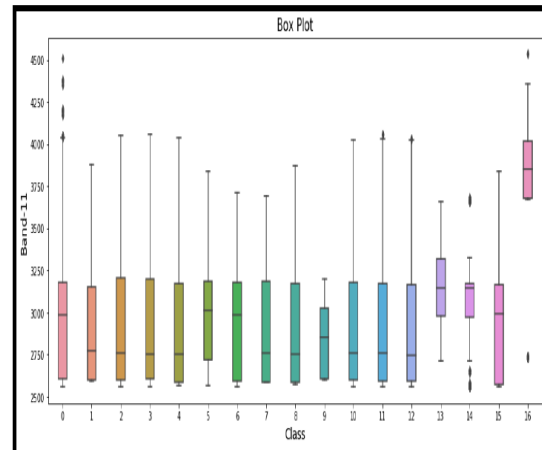


Figure 6: Box plot w.r.t bands of the HSI

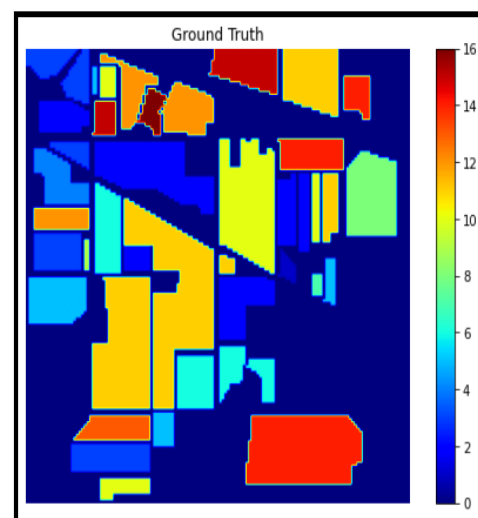
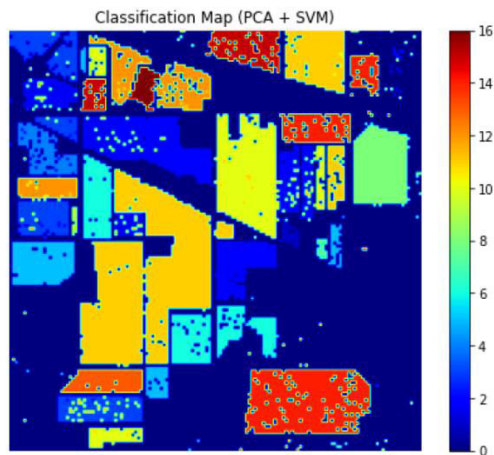


Figure 7: Classification map



**Figure 8: Classification map (PCA+SVM)**

## 5.CONCLUSION :

A spectral and spatial classification method was developed for the hyperspectral image classification. The proposed approach uses the SVM classifier with kernels for the classification. The results show that the RBF kernel will have the highest overall accuracy. Hyperspectral image contains many numbers of bands so more computational time is required to compute the kernel parameters and classification maps. In order to minimize the computational time, spectral features are reduced using PCA and only a few principal components are used for the classification. Furthermore, the classification accuracy of the SVM is increased by using the guided image filter with the first principal component having a guided image. From the experimental results it can be inferred that overall accuracy is improved from 81.15% of SVM to 94.22% for RBF kernels with the same training samples. In the future scope, computational time of the classification can be further reduced by employing

supervised dimensionality reduction algorithms.

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