

***COMPARATIVE ANALYSIS OF MACHINE LEARNING
ALGORITHMS FOR
LAND COVER CLASSIFICATION USING
HYPERSPECTRAL IMAGES***

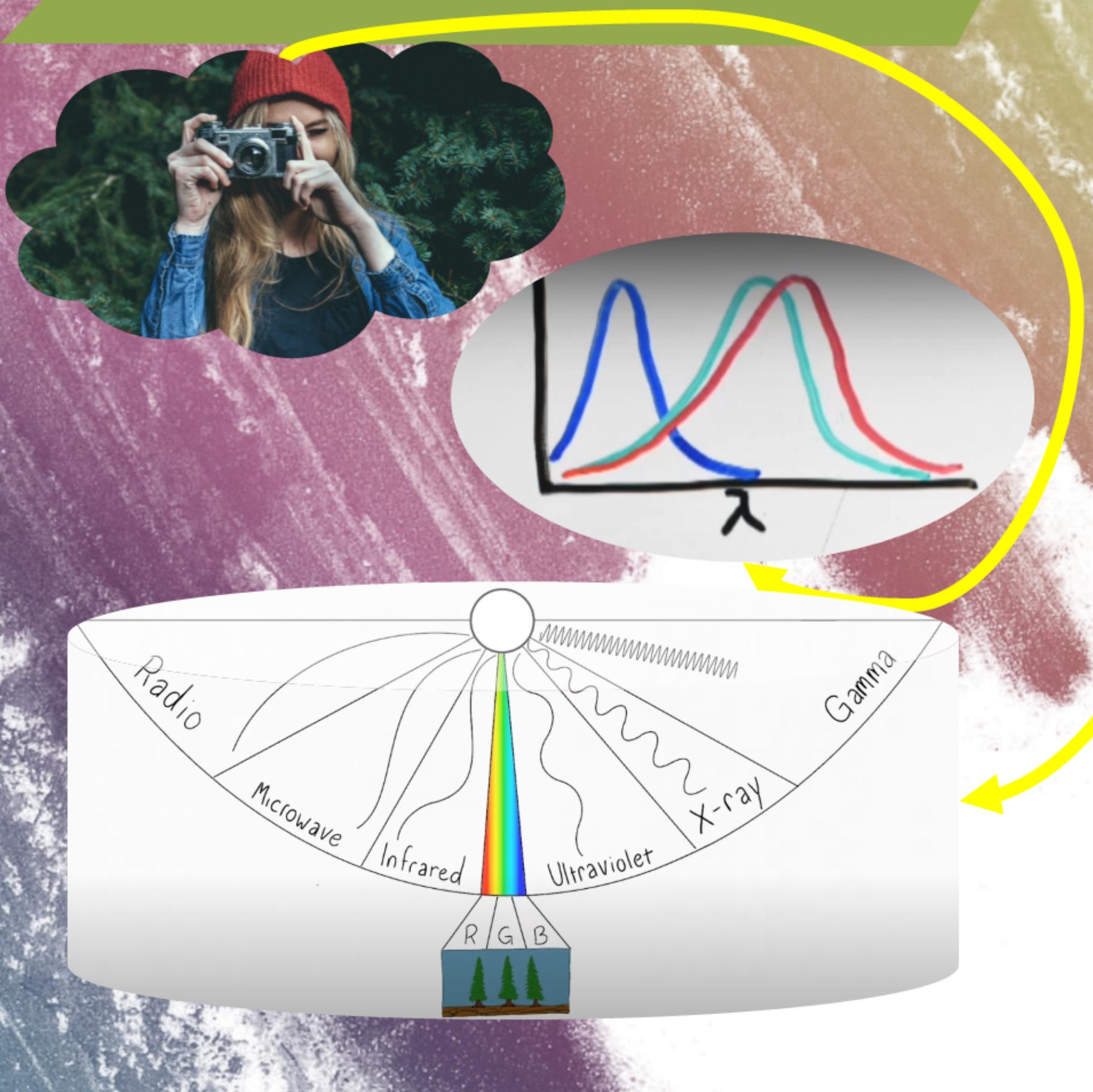
University of the Philippines - Diliman



AI 201 Mini-Project | 1st semester, A.Y. 2023-2024 | January 16, 2024

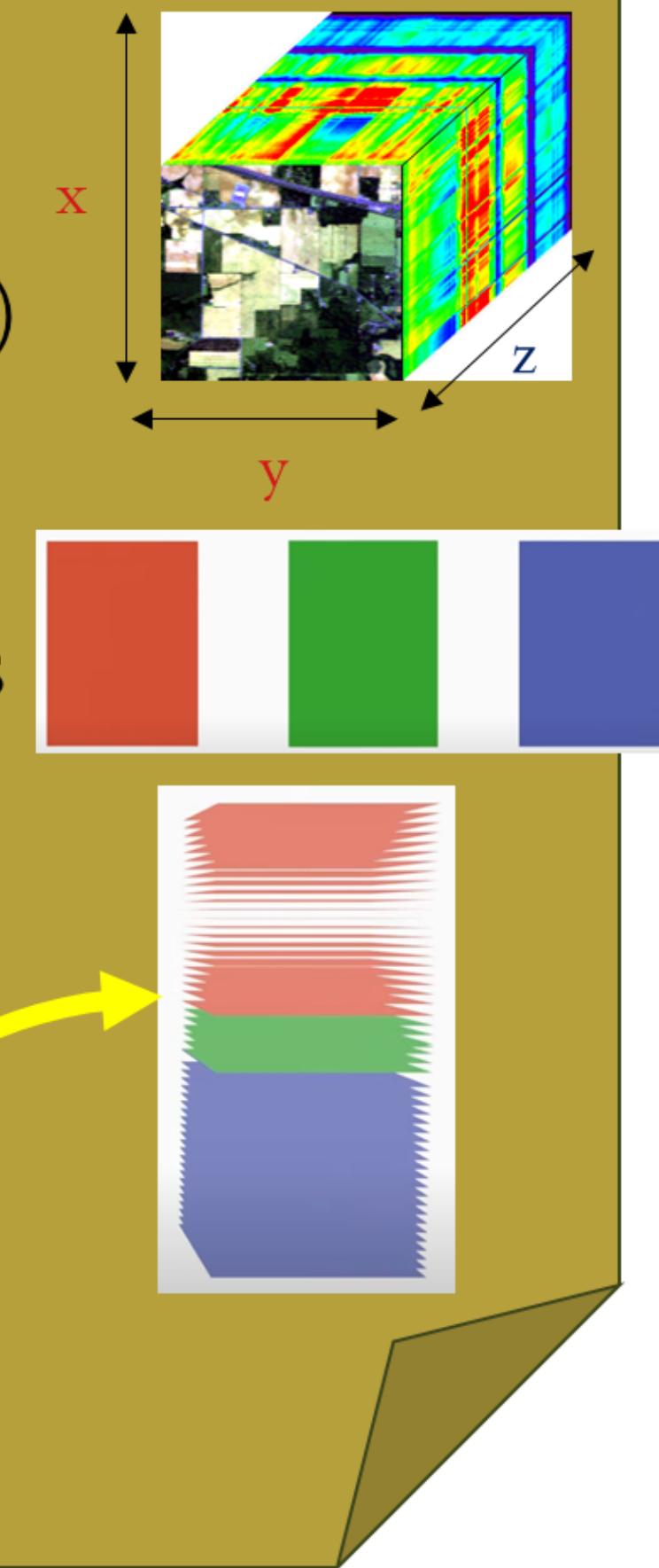
Mary Nathalie Dela Cruz | Elsa Joy Horiondo

INTRODUCTION

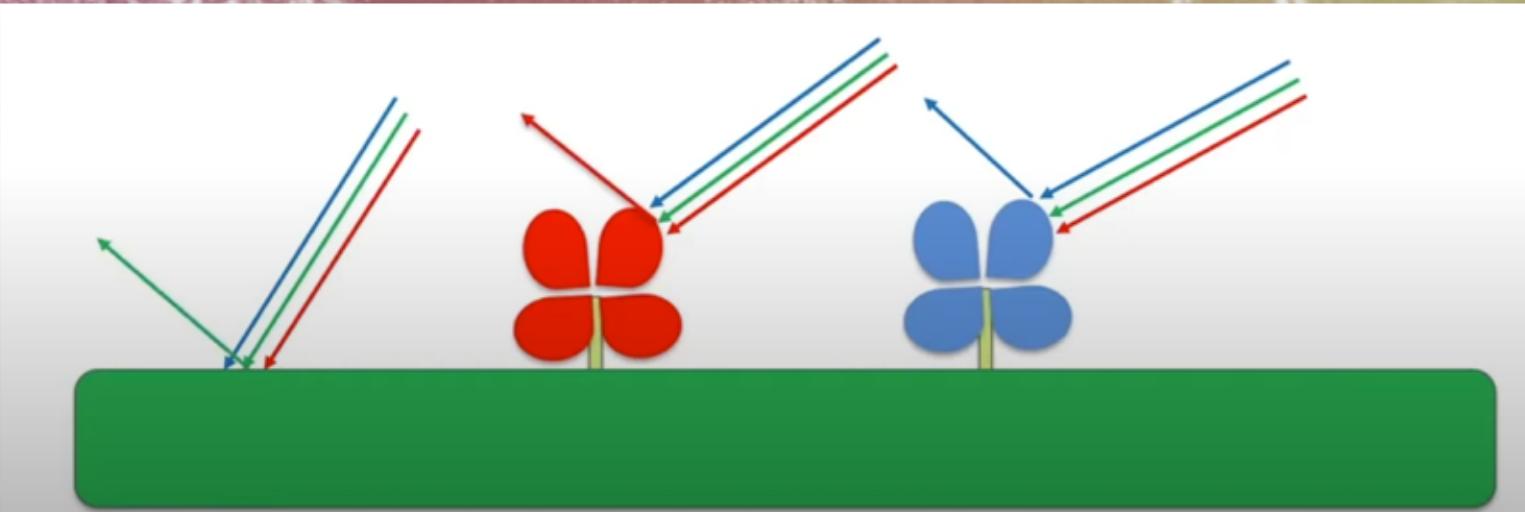
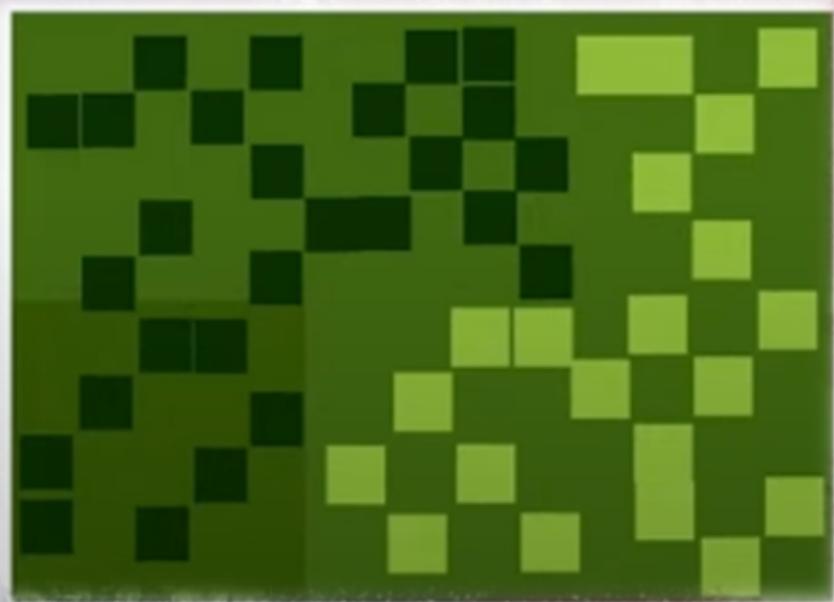


Hyperspectral Imaging (HSI)

- Imaging + Spectroscopy
- Data Cube
 - Spatial (x- and y-axis)
 - Wavelength (z-axis)
- Principles
 - EM spectrum records thousands of unique wavelengths
 - Grouped into bands
 - Uses hundreds of bands

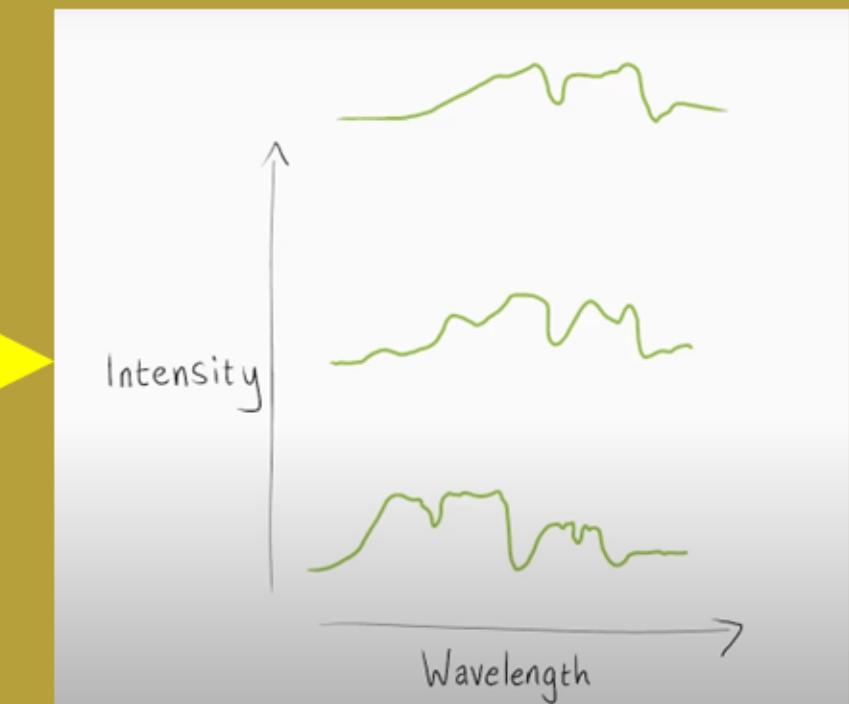


INTRODUCTION

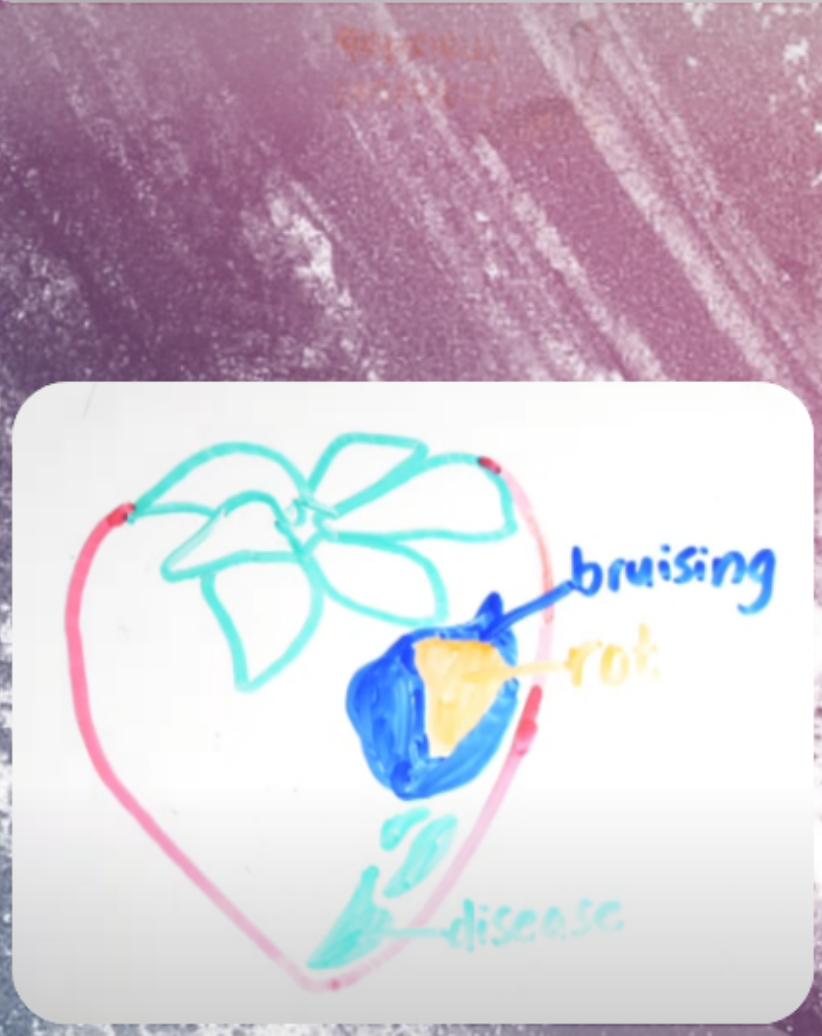


Hyperspectral Imaging (HSI)

- Spectral Fingerprint
 - Objects reflect and absorb (vegetation, minerals, building materials)
- Hyperspectral sensors can precisely detect these spectra



INTRODUCTION

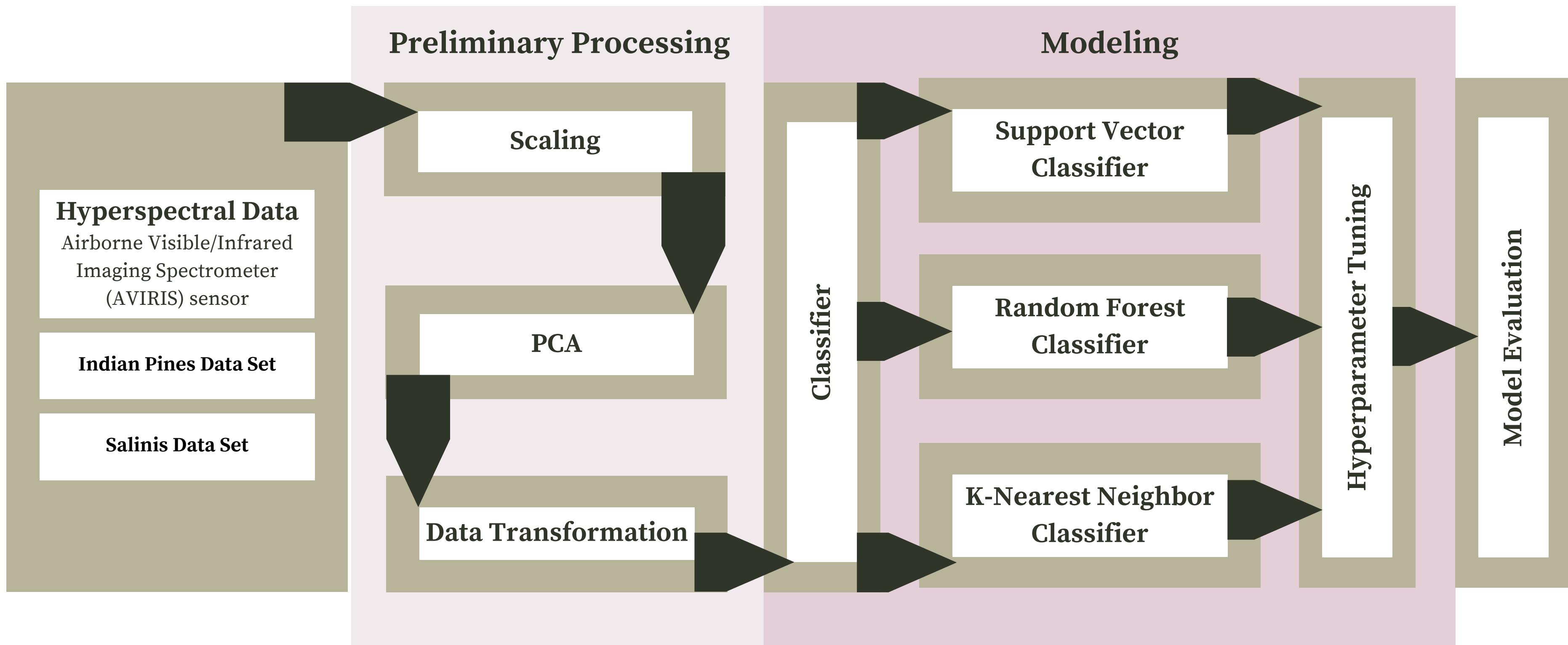


Hyperspectral Imaging (HSI)

- Relevance
 - Agriculture
 - Art conservation
 - Medicine
- Pollution Mapping
- Surveillance
- Geological mapping

OBJECTIVES

- To analyze HSI of Indian Pines and Salinas dataset using three classification methods
- To determine the optimal hyperparameters for the models
- To gauge the algorithms thru evaluation metrics (Precision, Recall, F1)



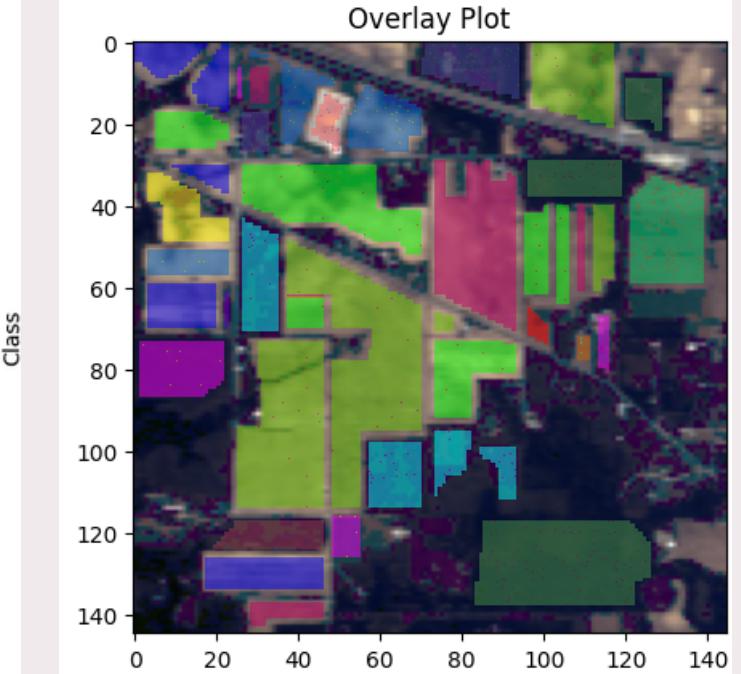
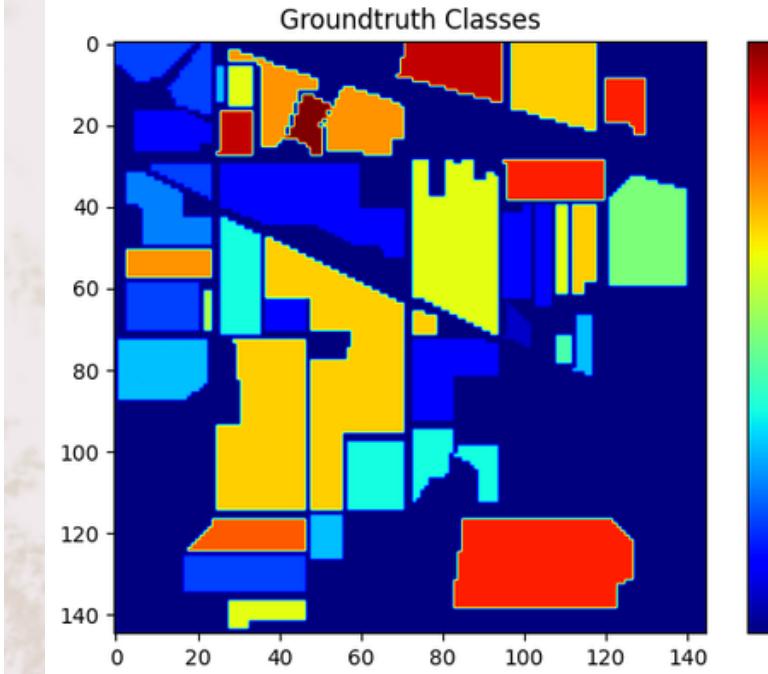
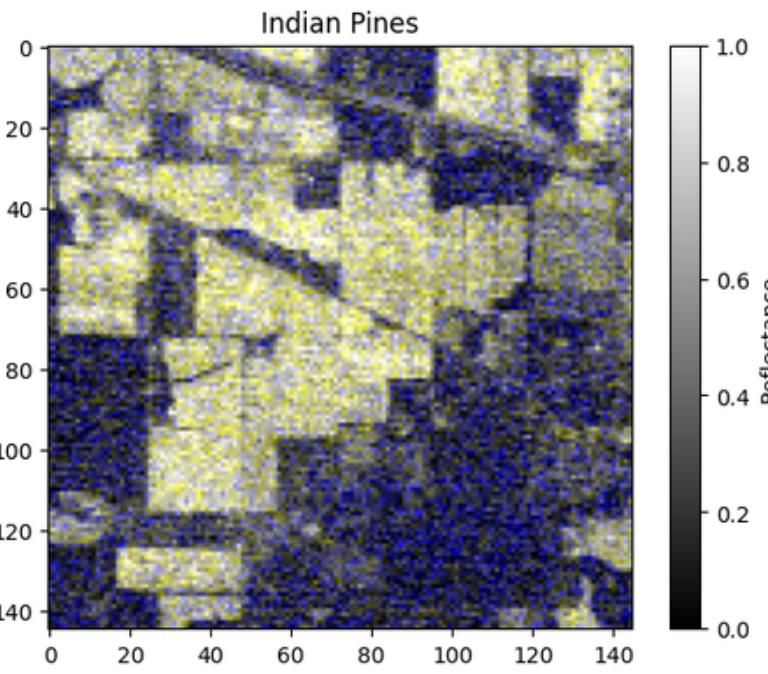
HYPERSPECTRAL DATA

from AVIRIS Sensor

Indian Pines Data Set

Location: Indian Pines test site in North-Western Indiana, U.S.A.
spatial extent of 145×145 pixels per band with 200 bands

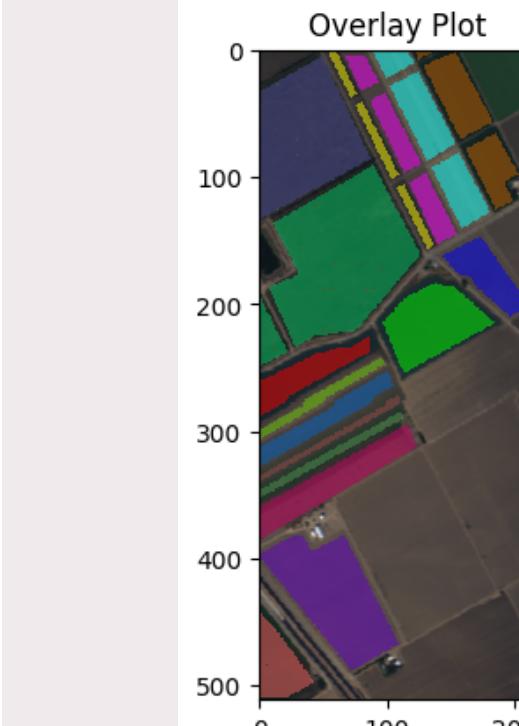
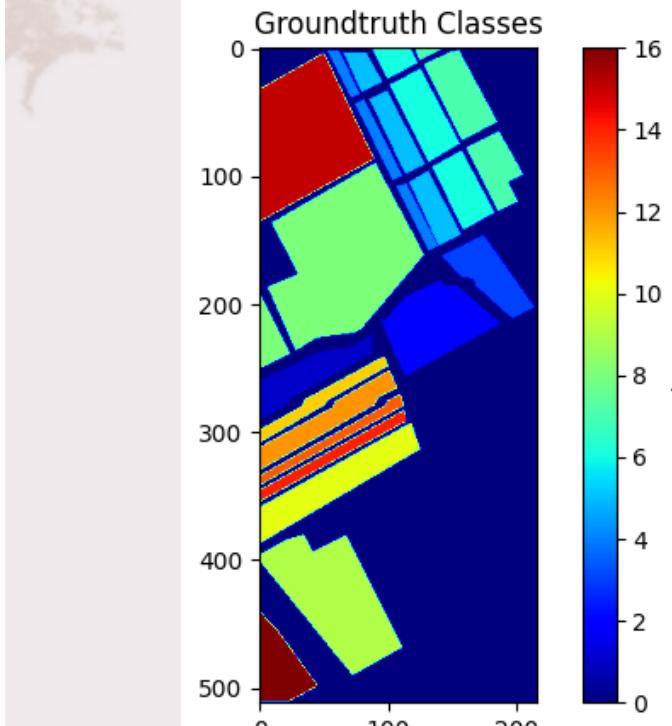
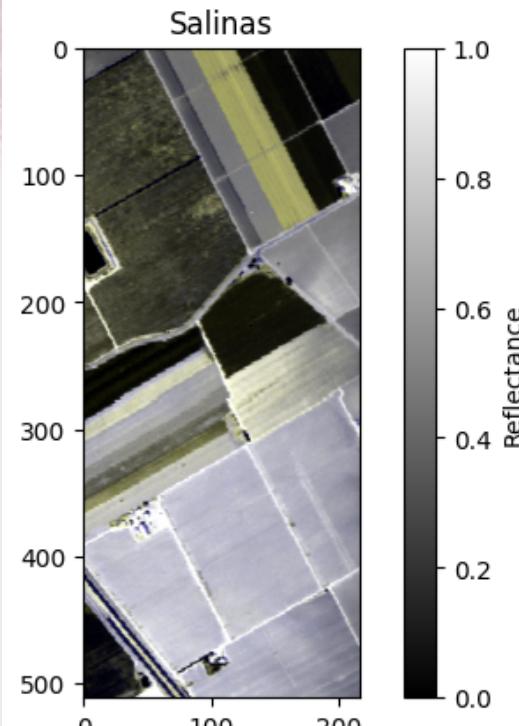
Class	1	2	3	4	5	6	7	8
Name	Alfalfa	Corn notill	Corn mintill	Corn	Grass pasture	Grass trees	Grass pasture mowed	Hay windrowed
Samples	46	1428	830	237	483	730	28	478
Class	9	10	11	12	13	14	15	16
Name	Oats	Soybean notill	Soybean mintill	Soybean clean	Wheat	Woods	Buildings grass trees drives	Stone steel towers
Samples	20	972	2455	593	205	1265	386	93



Salinas Data Set

Location: Salinas Valley, California, U.S.A.
spatial extent of 512×217 pixels per band with 204 bands

Class	1	2	3	4	5	6	7	8
Name	Broccoli green weeds 1	Broccoli green weeds 2	Fallow	Fallow rough plough	Fallow smooth	Stubble	Celery	Grapes untrained
Samples	2009	3726	1976	1394	2678	3959	3579	11271
Class	9	10	11	12	13	14	15	16
Name	Soil vineyard develop	Corn senesced green weeds	Lettuce romaine 4 wk	Lettuce romaine 5wk	Lettuce romaine 6wk	Lettuce romaine 7wk	Vineyard untrained	Vineyard vertical trellis
Samples	6203	3278	1068	1927	916	1070	7268	1807



PRE-PROCESSING

SCALING

$$z = (x - u) / s$$

x is the original feature vector
 u is the mean of feature vector
 s is the standard deviation of vector

PCA

Covariance Computation
 $C_{ij} = 1/(n-1) * \sum (x_i - u_i) * (x_j - u_j)$

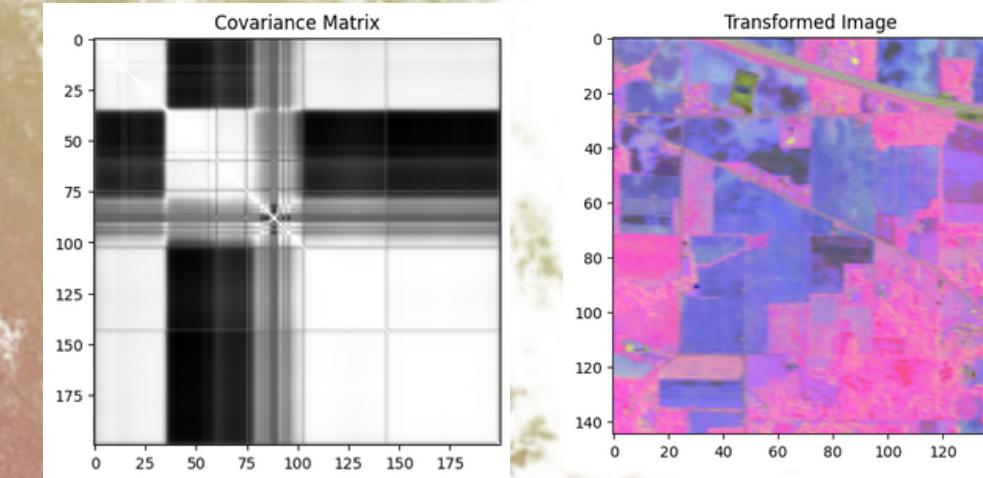
x_i and x_j are data vectors for bands i and j
 u_i and u_j are their respective means
 n is the number of pixels

Eigendecomposition

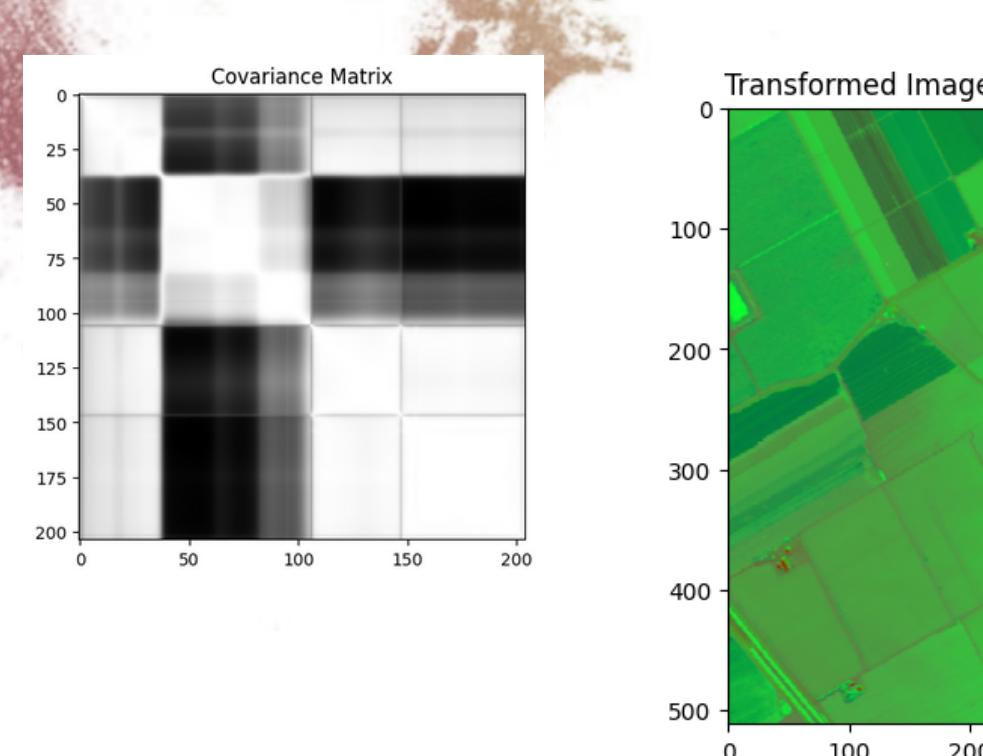
Data Transformation
 $y = \text{dot}(E^T, x)$

E is the matrix of eigenvectors
 x is the original data vector
 y is the transformed data vector

Indian Pine Dataset
200 bands > 108 bands



Salinas Dataset
204 bands > 19 bands



Data Transformation

3D Data Cube

2D Data Frame

Train-Test Split

pixel_row	pixel_col	band-1	band-108	class
0	0	-10.573985	-0.083758	3
0	1	-13.157707	-0.060955	3
0	2	-14.070350	-0.139894	3
0	3	-13.356123	-0.064724	3
0	4	-11.591195	-0.114516	3

MODEL DEVELOPMENT

INDIAN PINES

Support Vector
Classifier

Random Forest
Classifier

K-Nearest Neighbor
Classifier

Optimal
hyperparameters

C = 100
gamma = scale
kernel = rbf
time = 83.1 s

n_estimators = 200
time = 185.3 s

metric = manhattan
n_neighbors = 9
weights = distance
time = 32.7 s

Accuracy output

83.85%

71.85%

73.41%

MODEL DEVELOPMENT

SALINAS

Support Vector
Classifier

Random Forest
Classifier

K-Nearest Neighbor
Classifier

Optimal
hyperparameters

C = 1000
kernel = rbf
score = .920
time = 6.0 mins

n_estimators = 230
time = 4.0 mins

metric = manhattan
n_neighbors = 11
weights = distance
time = 6.5 mins

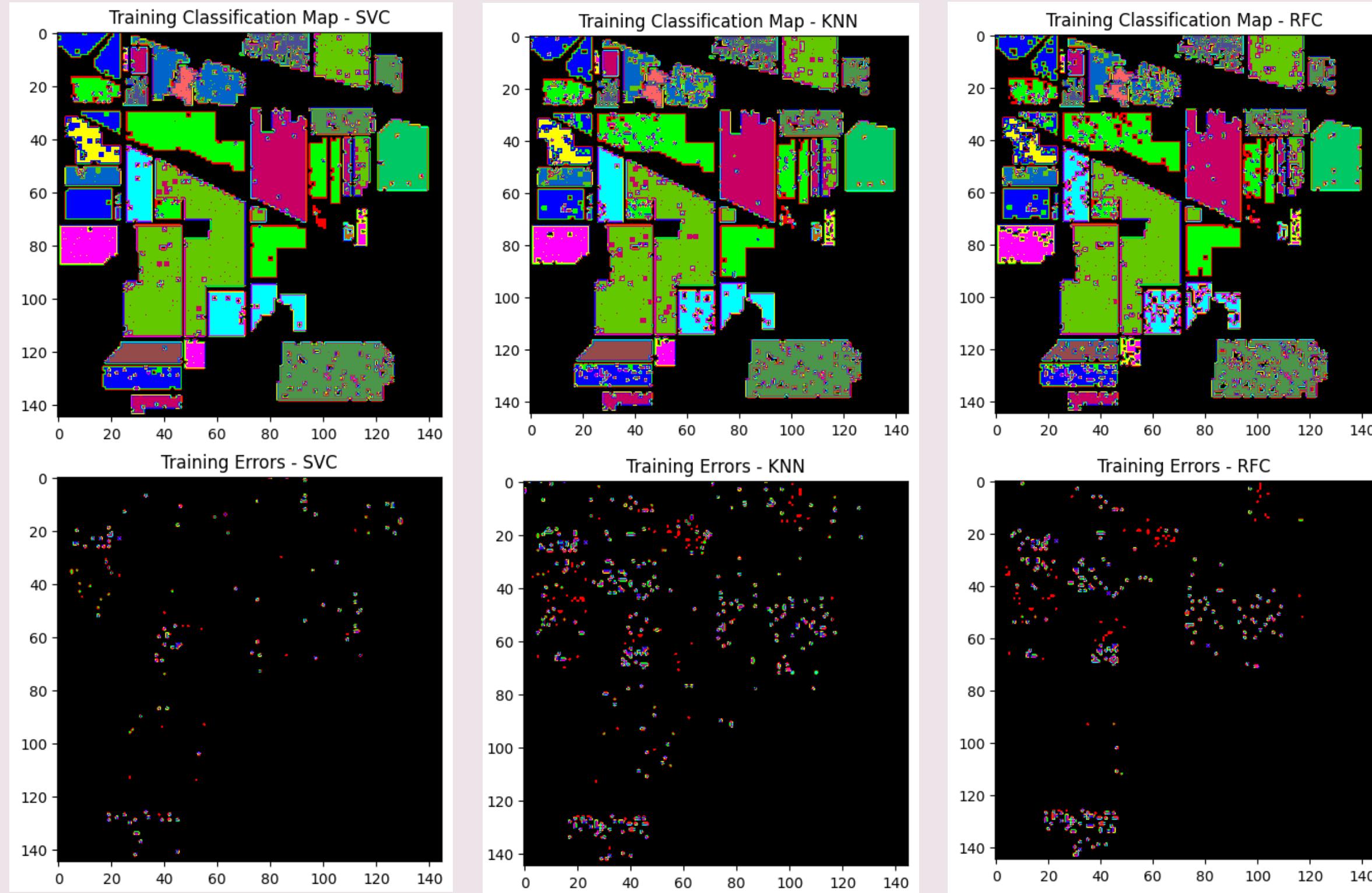
Accuracy output

91.98%

91.95%

90.75%

INDIAN PINES RESULTS



INDIAN PINES RESULTS

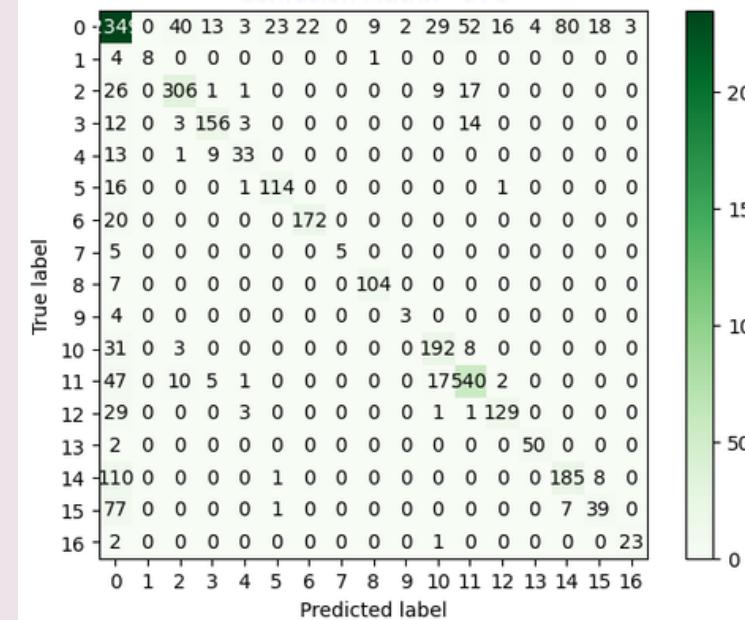
- **SVC** surpassed its counterparts in terms of:
 - 83.8% accuracy
 - 83% precision
 - 84% recall

Classification Report - SVC

Accuracy: 83.85 %

	precision	recall	f1-score	support
0	0.85	0.88	0.87	2663
1	1.00	0.62	0.76	13
2	0.84	0.85	0.85	360
3	0.85	0.83	0.84	188
4	0.73	0.59	0.65	56
5	0.82	0.86	0.84	132
6	0.89	0.90	0.89	192
7	1.00	0.50	0.67	10
8	0.91	0.94	0.92	111
9	0.60	0.43	0.50	7
10	0.77	0.82	0.80	234
11	0.85	0.87	0.86	622
12	0.87	0.79	0.83	163
13	0.93	0.96	0.94	52
14	0.68	0.61	0.64	304
15	0.60	0.31	0.41	124
16	0.88	0.88	0.88	26
accuracy		0.84		5257
macro avg	0.83	0.74	0.77	5257
weighted avg	0.83	0.84	0.83	5257

Confusion Matrix - SVC

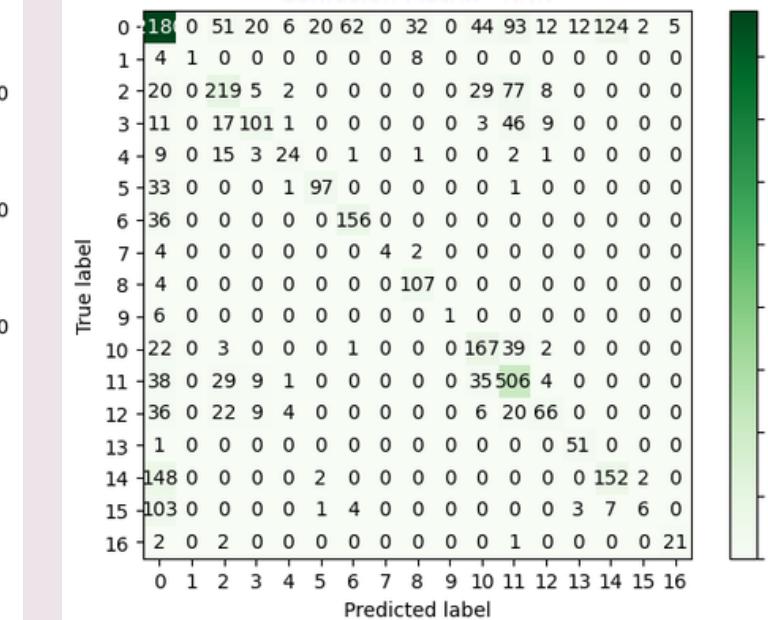


Classification Report - KNN

Accuracy: 73.41 %

	precision	recall	f1-score	support
0	0	0.82	0.82	2663
1	1	1.00	0.08	13
2	2	0.61	0.61	360
3	3	0.69	0.54	188
4	4	0.62	0.43	56
5	5	0.81	0.73	132
6	6	0.70	0.81	192
7	7	1.00	0.40	10
8	8	0.71	0.96	111
9	9	1.00	0.14	7
10	10	0.59	0.71	234
11	11	0.64	0.81	622
12	12	0.65	0.40	163
13	13	0.77	0.98	52
14	14	0.54	0.50	304
15	15	0.60	0.05	124
16	16	0.81	0.81	26
accuracy		0.73		5257
macro avg	0.74	0.58	0.59	5257
weighted avg	0.73	0.73	0.72	5257

Confusion Matrix - KNN

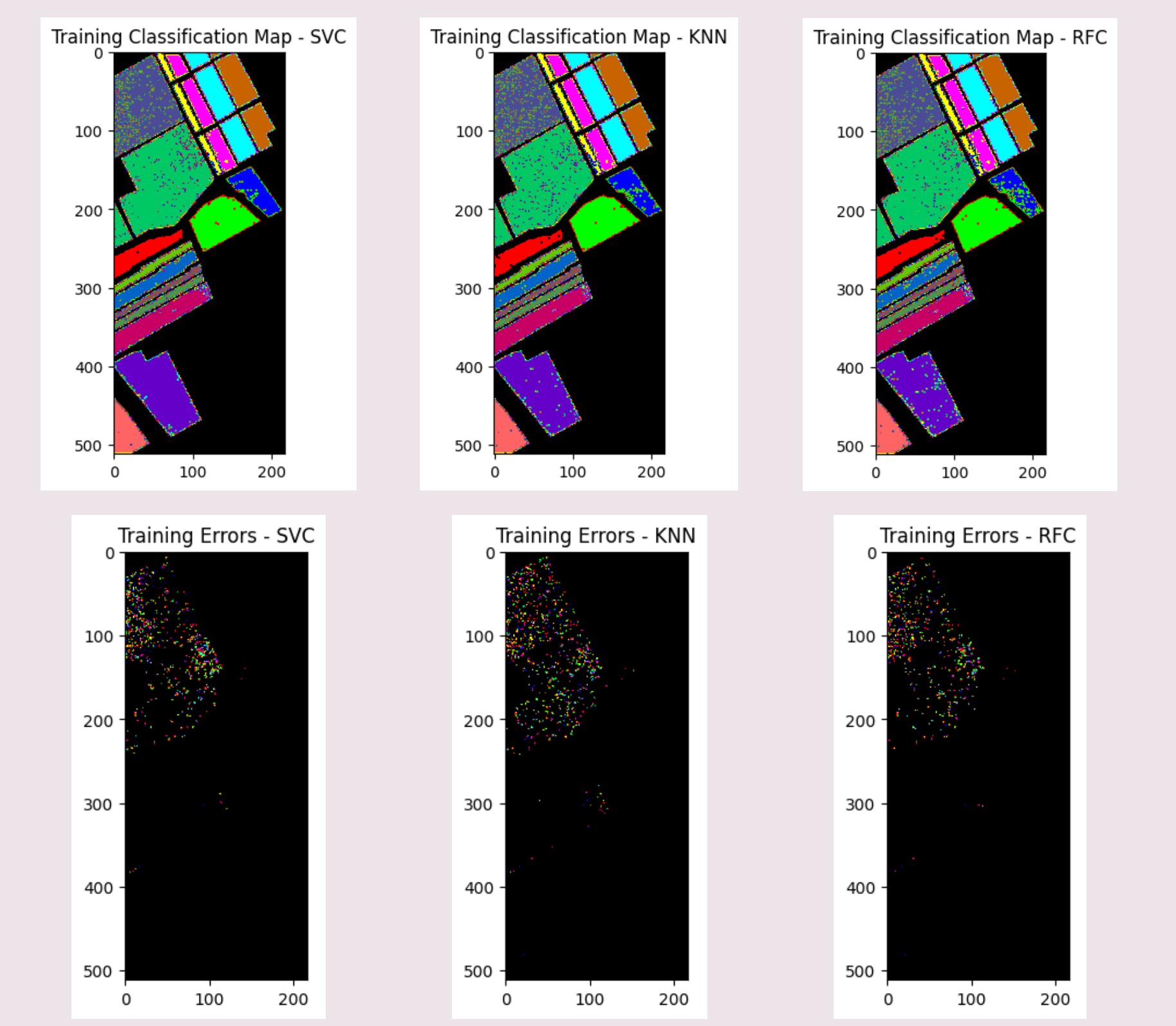


Classification Report - RFC

Accuracy: 71.85 %

	precision	recall	f1-score	support
0	0	0.72	0.93	2663
1	1	0.00	0.00	13
2	2	0.65	0.55	360
3	3	0.84	0.38	188
4	4	0.90	0.16	56
5	5	0.97	0.50	132
6	6	0.89	0.35	192
7	7	0.00	0.00	10
8	8	0.8		

SALINAS RESULTS



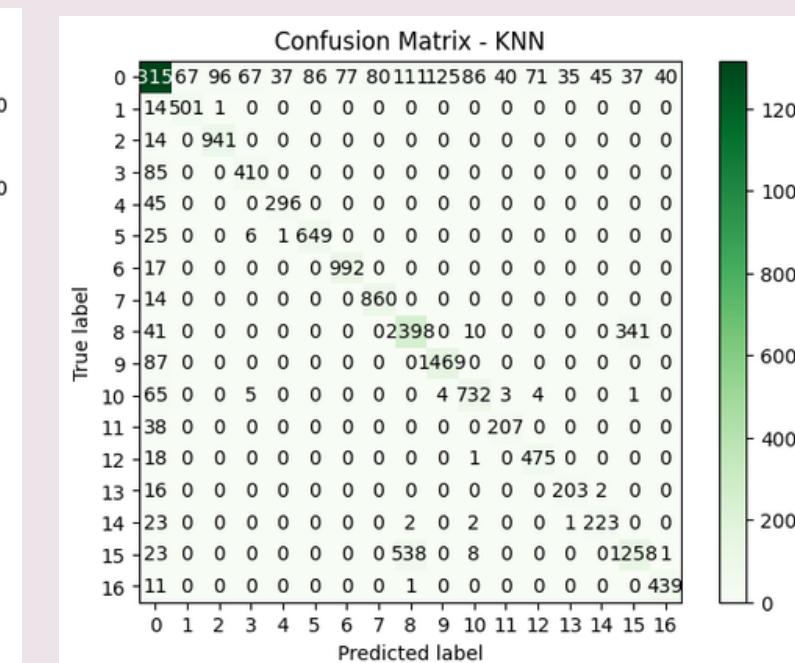
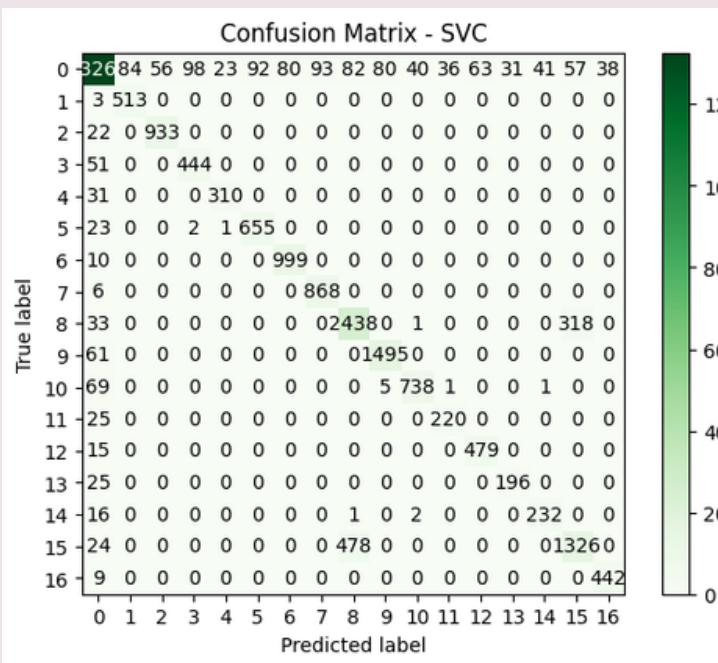
SALINAS RESULTS

There's not much any significant difference in terms of the evaluation metrics for the Salinas dataset

Classification Report - SVC				
	precision	recall	f1-score	support
0	0.97	0.93	0.95	14255
1	0.86	0.99	0.92	516
2	0.94	0.98	0.96	955
3	0.82	0.90	0.85	495
4	0.93	0.91	0.92	341
5	0.88	0.96	0.92	681
6	0.93	0.99	0.96	1009
7	0.90	0.99	0.95	874
8	0.81	0.87	0.84	2790
9	0.95	0.96	0.95	1556
10	0.94	0.91	0.93	814
11	0.86	0.90	0.88	245
12	0.88	0.97	0.92	494
13	0.86	0.89	0.88	221
14	0.85	0.92	0.88	251
15	0.78	0.73	0.75	1828
16	0.92	0.98	0.95	451
accuracy			0.92	27776
macro avg	0.89	0.93	0.91	27776
weighted avg	0.92	0.92	0.92	27776

Classification Report - KNN				
	precision	recall	f1-score	support
0	0.96	0.92	0.94	14255
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16	0.91	0.97	0.94	451
accuracy			0.91	27776
macro avg	0.87	0.91	0.89	27776
weighted avg	0.91	0.91	0.91	27776

Classification Report - RFC				
	precision	recall	f1-score	support
0	0.94	0.94	0.94	14255
1	0.88	0.99	0.93	516
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Conclusion

- **Scaling and Dimensionality Reduction Via PCA**
 - Indian Pines: 200 bands to 108 bands
 - Salinas: 204 bands to 19 bands
- **Hyperparameter Tuning via Grid Search**
- **Training Errors**
 - KNN for both Indian Pines and Salinas dataset
- **Evaluation metrics**
 - SVC performed best in the classification of Indian Pines dataset
 - No classifier is significantly better than others in the classification of Salinas dataset