

# Machine Learning and Multispectral Unmanned Aerial Vehicle Imagery Data for Agriculture.

Georgios Batsis

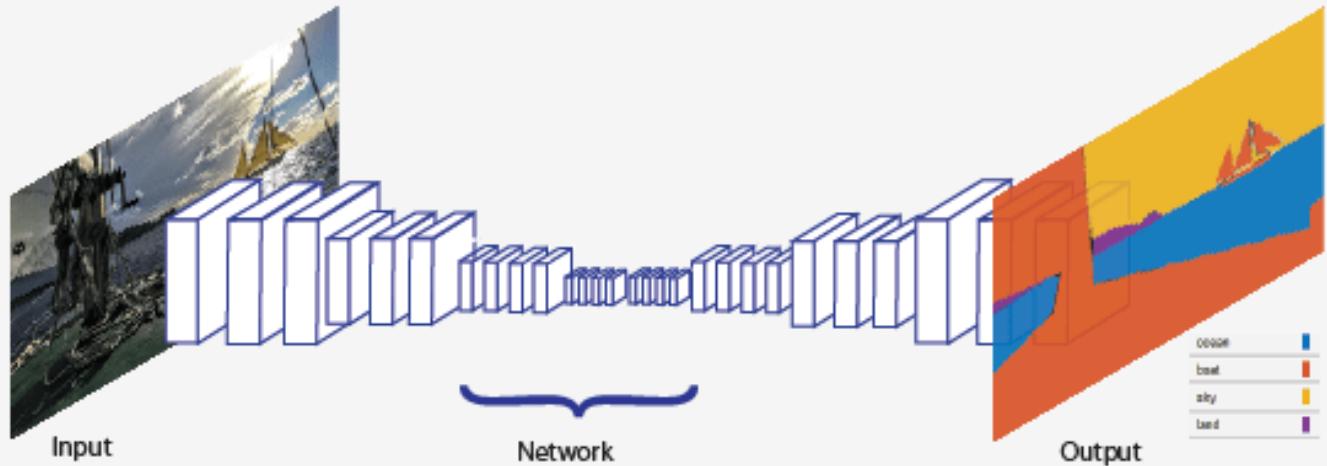
<https://github.com/gbatsis/VYSegML>

# Image Segmentation

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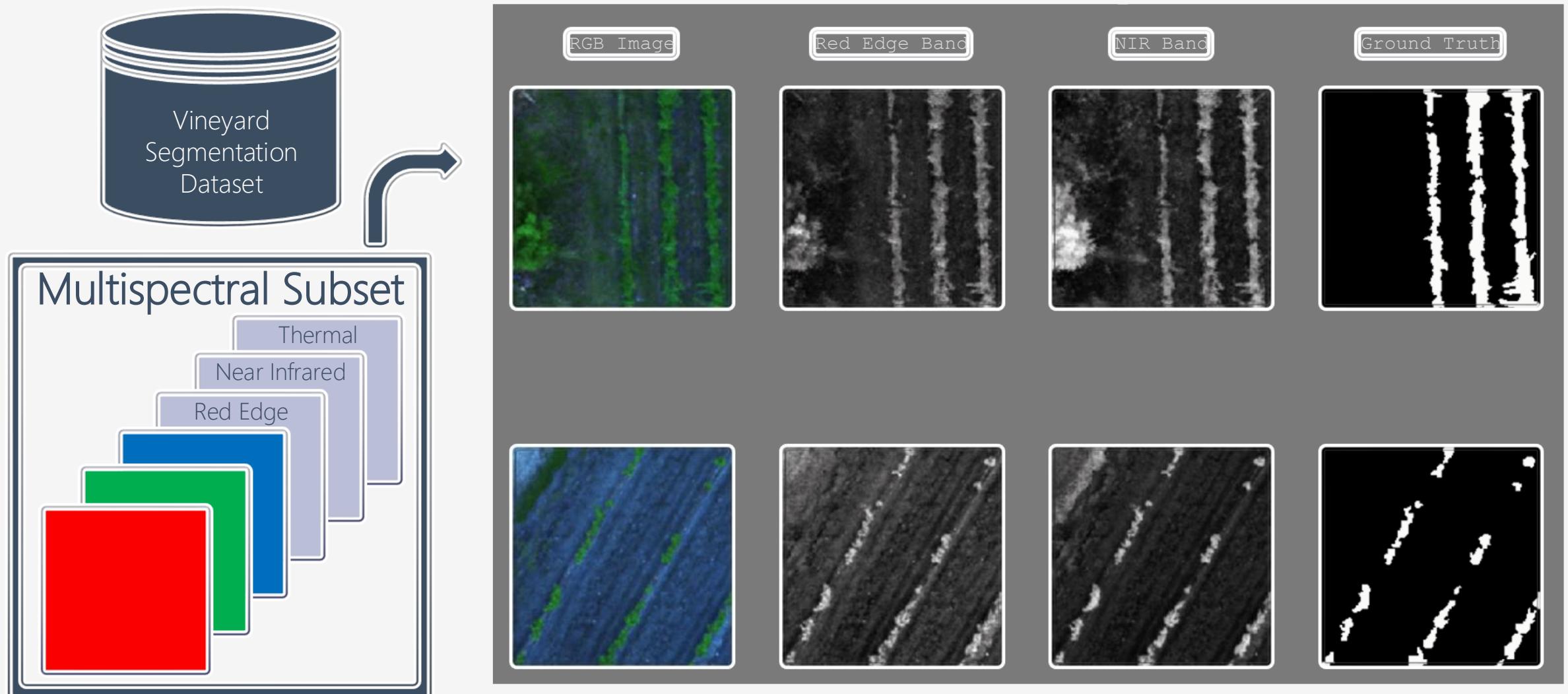
**Image segmentation →** Image is broken down into subgroups (Segments) → Assigning labels to pixels

**Image segmentation – Deep Learning →**



**Image segmentation – Machine (Non Deep) Learning →** Direct Pixel classification using informative Features.

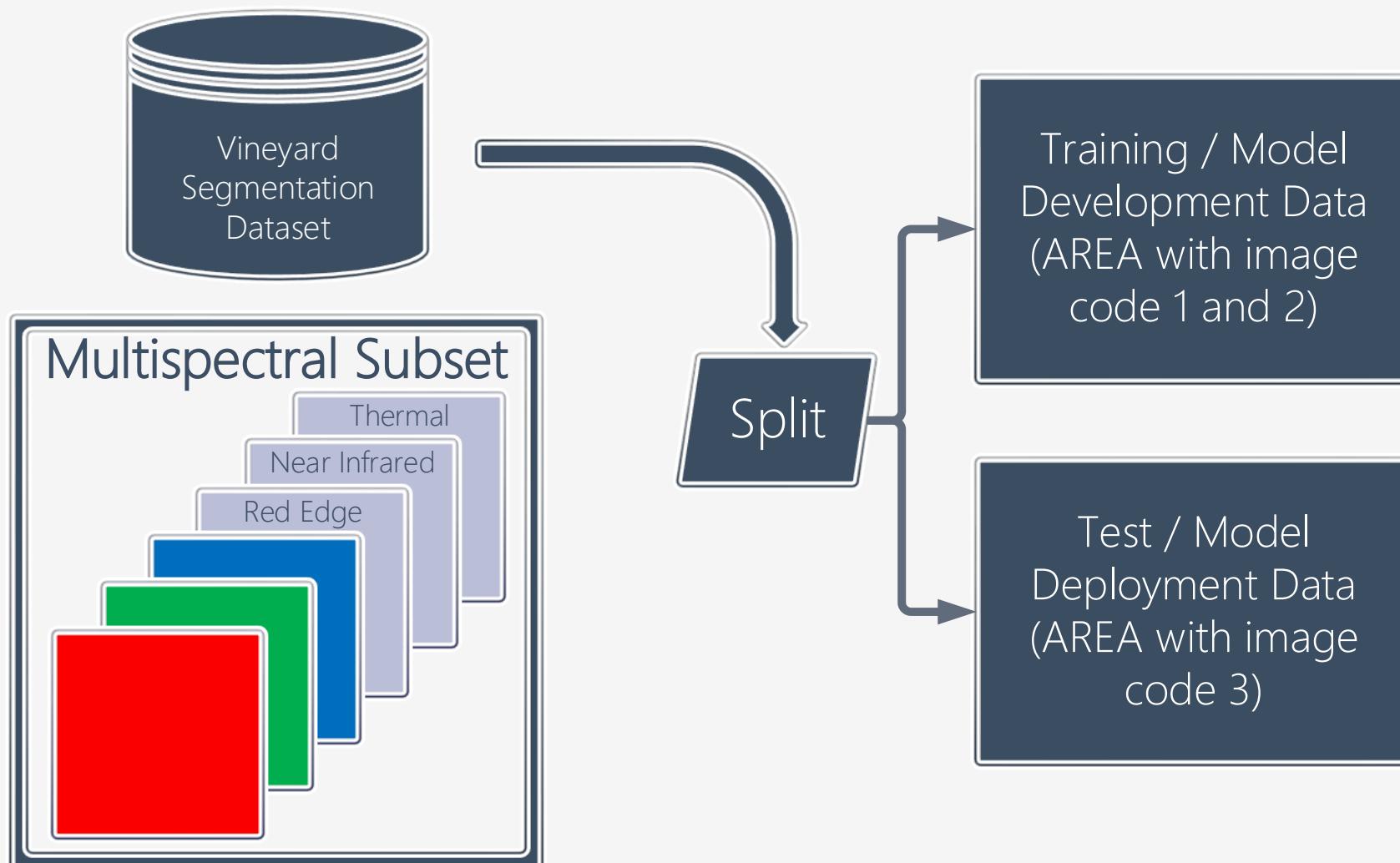
# Dataset



[https://github.com/Cybonic/DL\\_vineyard\\_segmentation\\_study](https://github.com/Cybonic/DL_vineyard_segmentation_study)

# Dataset

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<https://github.com/Cybonic/DL vineyard segmentation study.git>

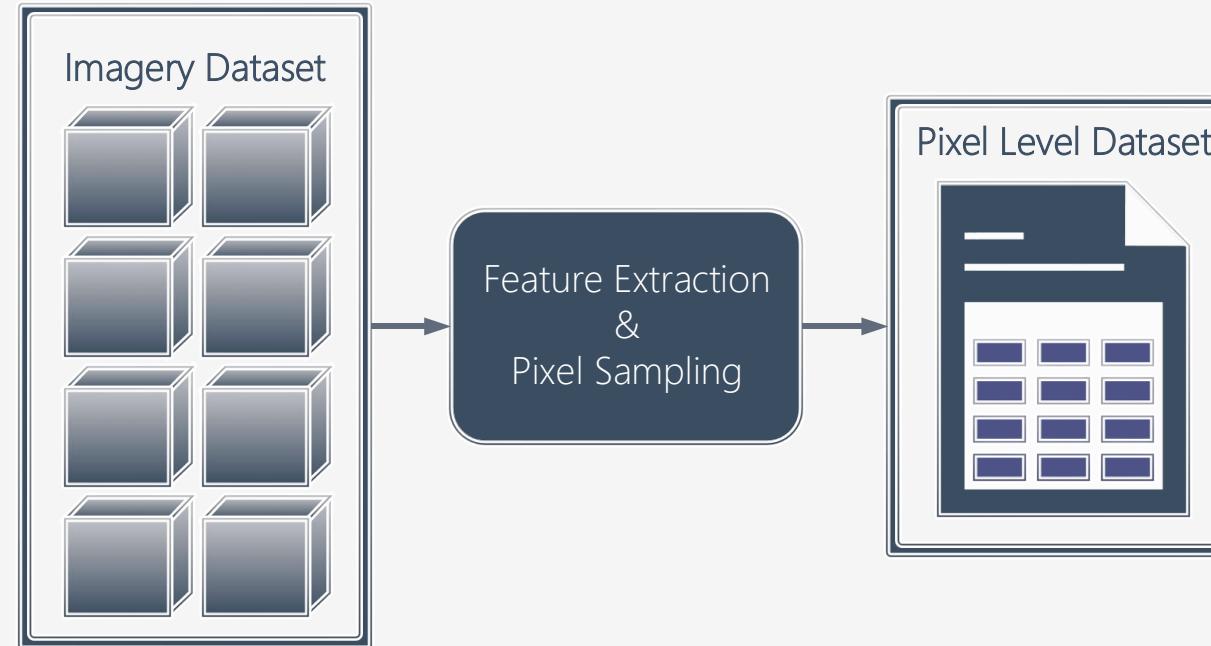
# Changing Form of Dataset

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**Image segmentation – Machine (Non Deep) Learning →** Direct Pixel classification using informative Features.

It is essential that we should change the form of Dataset to fit Machine Learning Classifiers and to perform Pixel Level Classification.

Extract a Pixel – Level Dataset from original Imagery Dataset.



# Feature Extraction

## Vegetation Indices

- A Vegetation Index (VI) is a mathematical combination of several spectral Bands which provides information about relationship between vegetation or crop patterns.
- A Vegetation Index (VI) is a spectral transformation metric for measuring the presence and state of vegetation.
- After its calculation, VI simulates a single channel image.

Red Edge Normalized Difference Vegetation index:  $\frac{NIR - RedEdge}{NIR + RedEdge}$

Normalized Difference Vegetation index:  $\frac{NIR - Red}{NIR + Red}$

Green Normalized Difference Vegetation index:  $\frac{NIR - Green}{NIR + Green}$

Blue Normalized Difference Vegetation index:  $\frac{NIR - Blue}{NIR + Blue}$

RedEdge-based Indices:

- $\frac{RedEdge - Red}{RedEdge + Red}$
- $\frac{RedEdge - Green}{RedEdge + Green}$
- $\frac{RedEdge - Blue}{RedEdge + Blue}$

Normalized Difference Green-Red index:  $\frac{Green - Red}{Green + Red}$

Normalized Difference Green-Blue index:  $\frac{Green - Blue}{Green + Blue}$

# Feature Extraction

## Texture Information

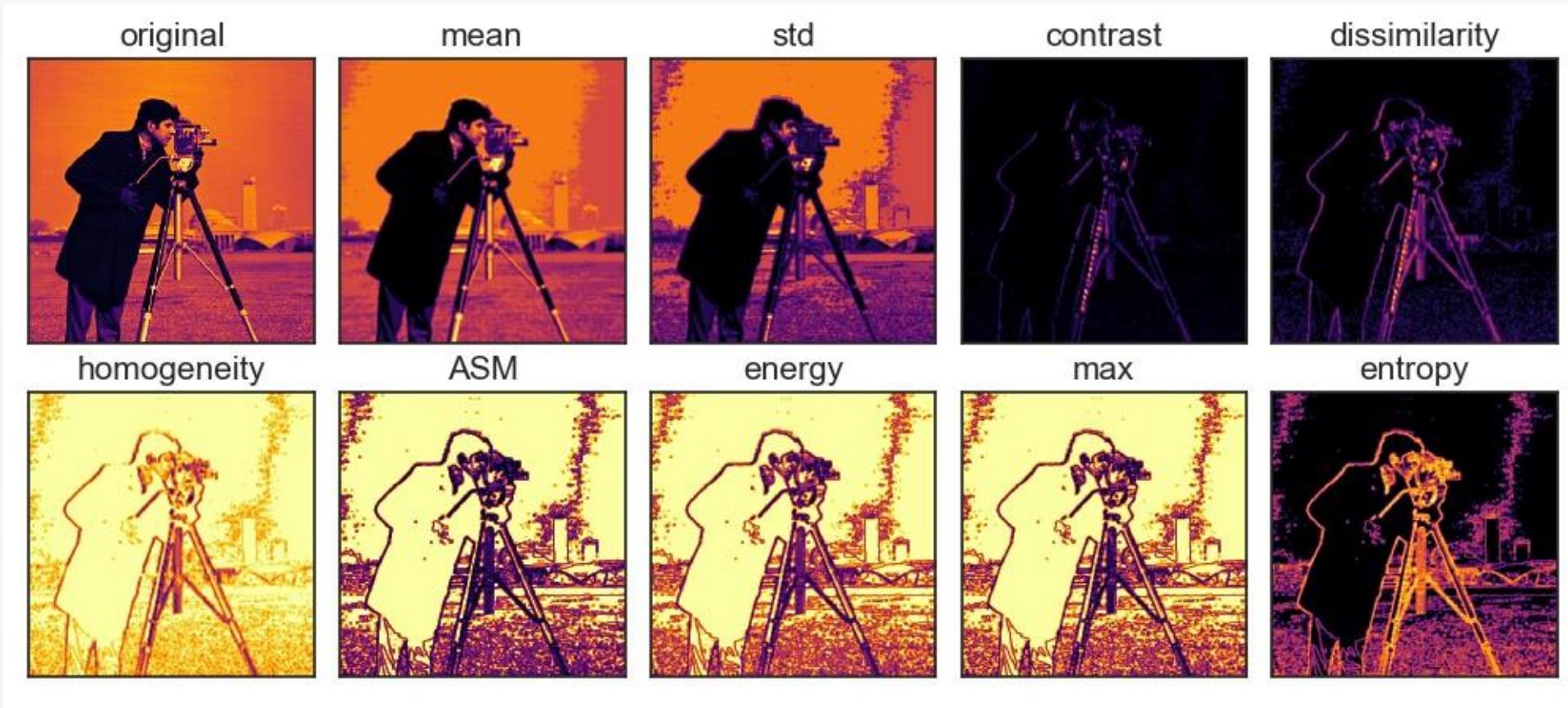
- *Gray Level Co-occurrence Matrix: Tabulation of how often pairs of pixel with specific values and in a specified spatial relationship occur in an image.*
- GLCM contains information about the relationship of intensity of a pixel and its neighborhood defined using a window.
- Texture Information are statistical measures extracted from the aforementioned matrix.
- Each measure simulates a single pixel image.

- Mean
- Standard Deviation
- Max
- Entropy
- Contrast =  $\sum_{i=0}^{N-1} \sum_{j=0}^{N-1} (i - j)^2$
- Homogeneity =  $\sum_{i=0}^{N-1} \sum_{j=0}^{N-1} \frac{P(i,j)}{1+(i-j)^2}$
- Angular second Moment (ASM)  $\sum_{i=0}^{N-1} \sum_{j=0}^{N-1} P(i, j)^2$
- Energy  $\sqrt{\sum_{i=0}^{N-1} \sum_{j=0}^{N-1} P(i, j)^2}$

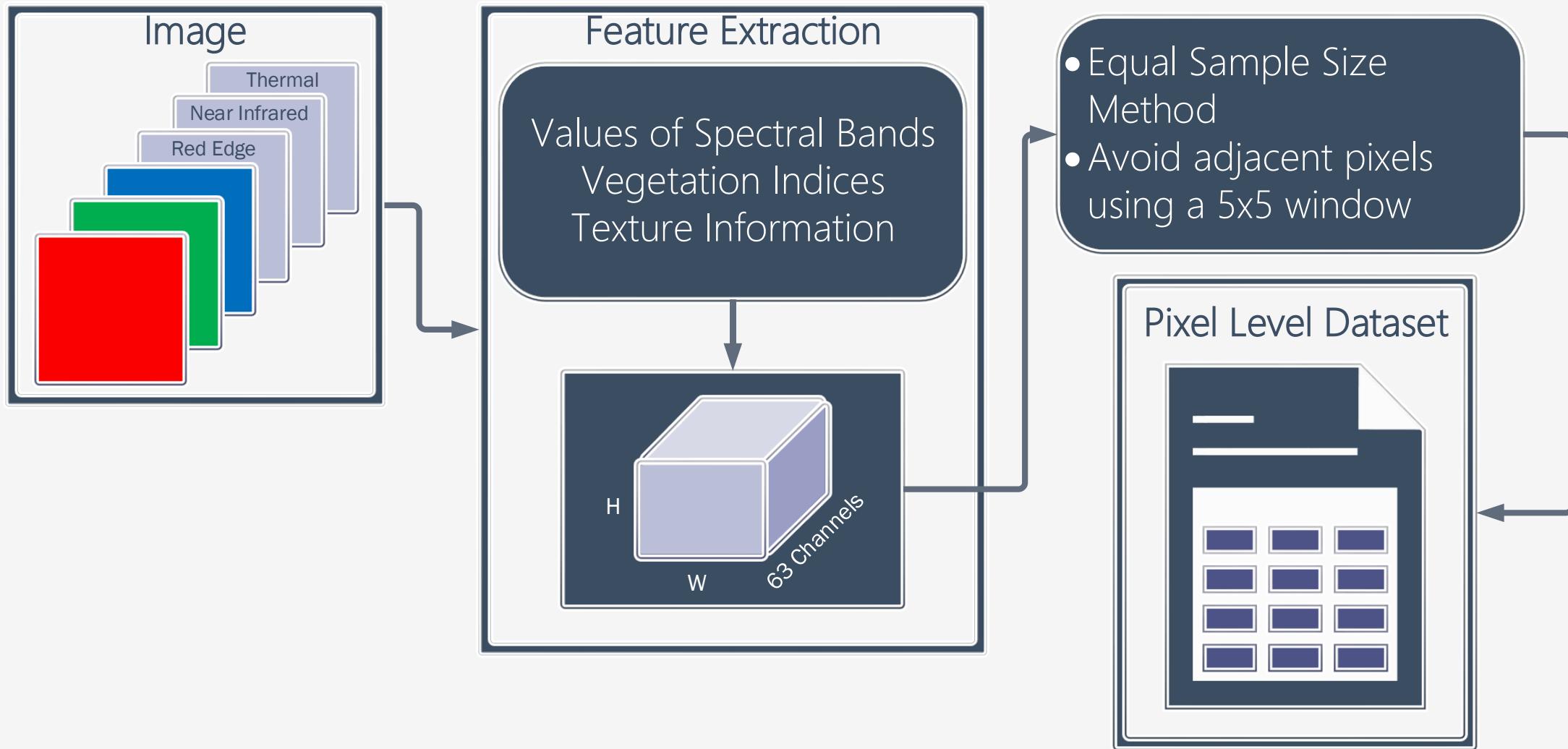
# Feature Extraction

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## Texture Information



# Changing Form of Dataset using Feature Extraction & Sampling



# Pixel Classification Dataset

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	R	G	B	RE	NIR	TH	RENDVI	NDVI	GNDVI
0	37.0	99.0	39.0	85.0	96.0	254.0	0.060773	0.443609	1.000000
1	112.0	75.0	121.0	72.0	60.0	254.0	1.000000	1.000000	1.000000
2	56.0	84.0	64.0	150.0	136.0	254.0	1.000000	0.416667	0.236364
3	51.0	46.0	29.0	46.0	52.0	254.0	0.061224	0.009709	0.061224
4	61.0	121.0	40.0	117.0	124.0	254.0	0.029046	0.340541	0.012245
	BNDVI	...		T_std	T_contrast	T_homogeneity	T_ASM	T_energy	\
0	0.422222	...		173.627426	0.0		25.0	625.0	25.0
1	1.000000	...		173.627426	0.0		25.0	625.0	25.0
2	0.360000	...		173.627426	0.0		25.0	625.0	25.0
3	0.283951	...		173.627426	0.0		25.0	625.0	25.0
4	0.512195	...		173.627426	0.0		25.0	625.0	25.0
	T_max	T_entropy	label		imgName		mode		
0	25.0	8.070781	0		img_1000000036		development		
1	25.0	8.070781	1		img_2000000127		development		
2	25.0	8.070781	1		img_3000000088		deploy		
3	25.0	8.070781	0		img_3000000046		deploy		
4	25.0	8.070781	0		img_1000000005		development		

# Comparison of Different Classifiers

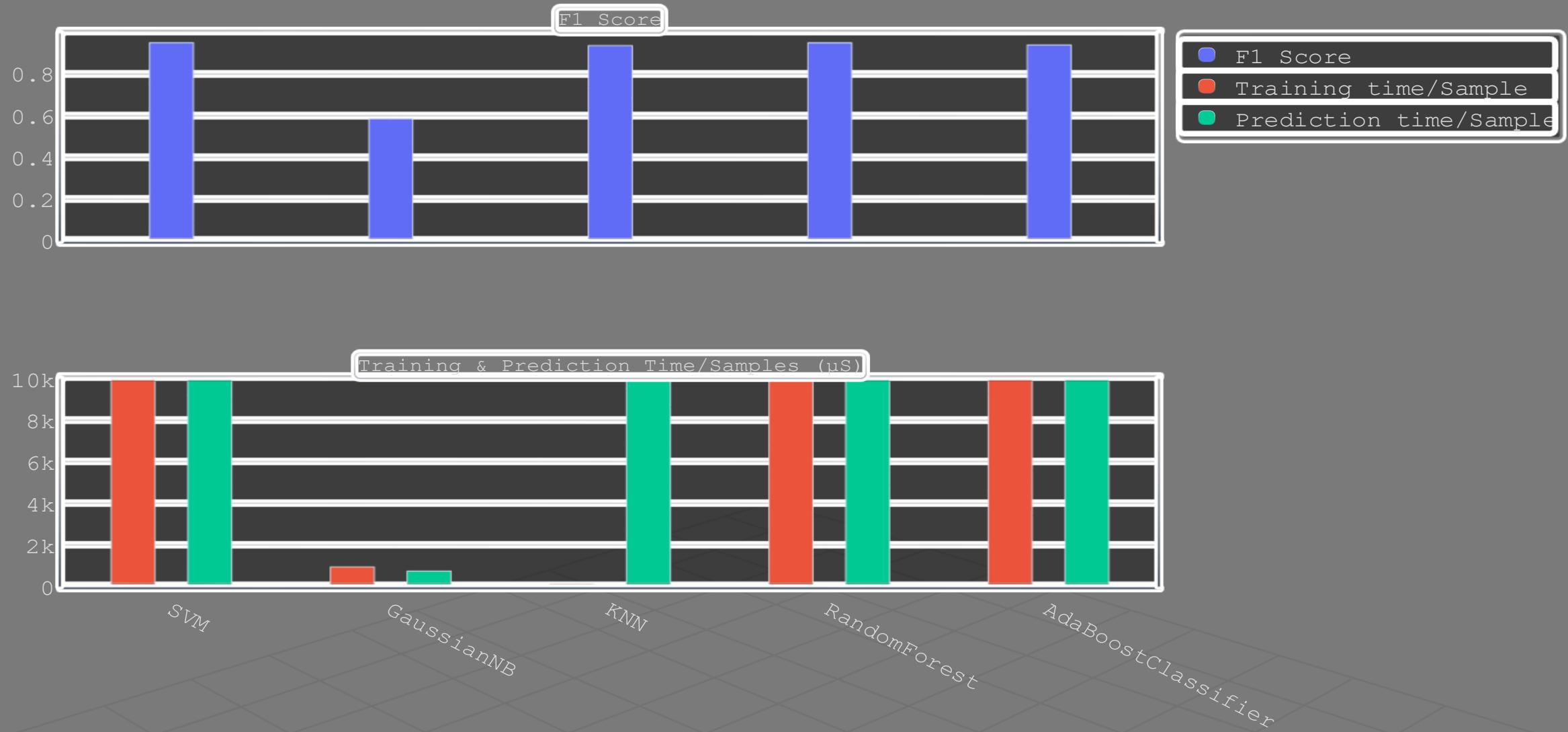
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Comparison of baseline models in terms of F1 Score, training and prediction time.

Machine Learning Models: SVM-RBF, Gaussian Naïve Bayes, K-Nearest Neighbor, Random Forest, AdaBoost

- *Comparison Using All Features*
  - Implementation of Feature Selection based on importance weights during the training process of Random Forest and AdaBoost.

## Model Comparison: All Features



# Comparison of Different Classifiers

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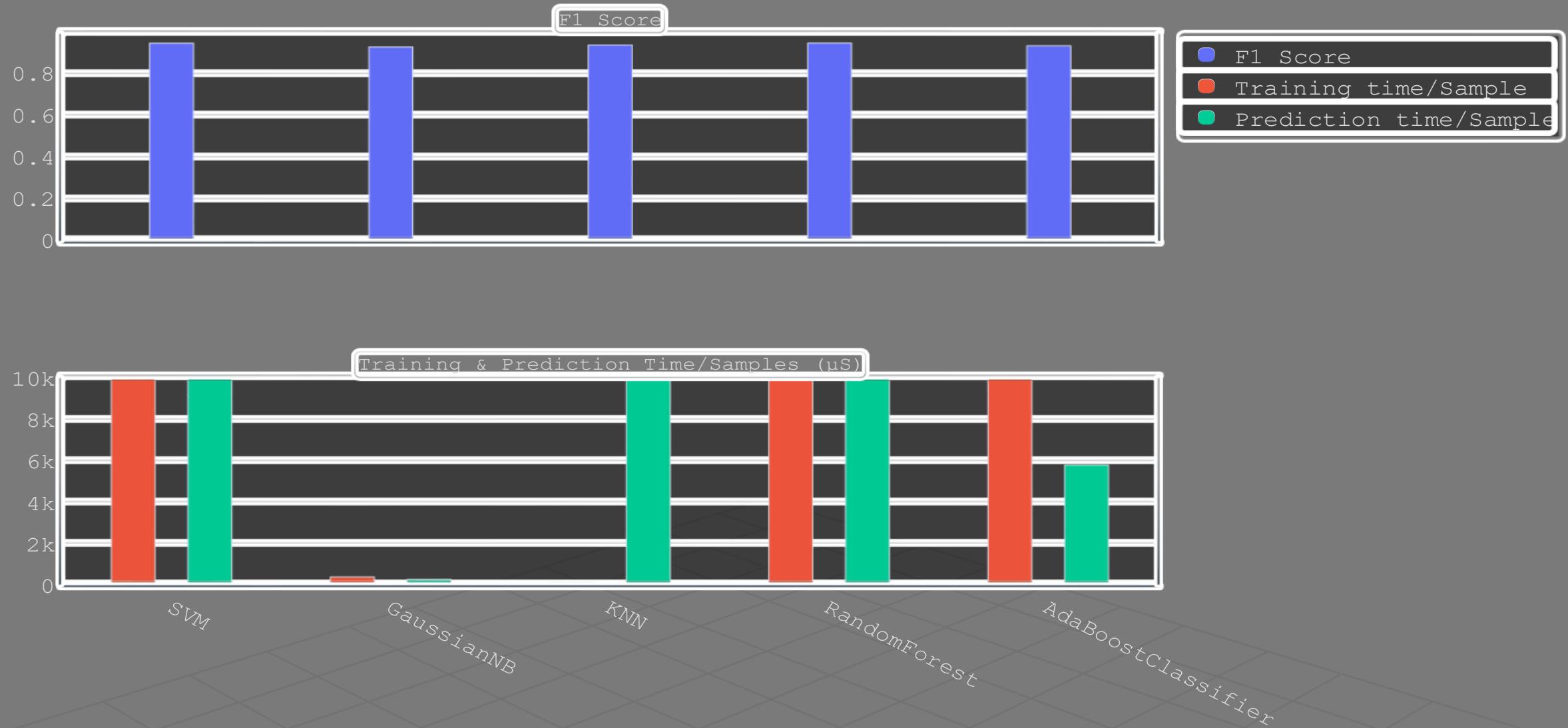
Comparison of baseline models in terms of F1 Score, training and prediction time.

Machine Learning Models: SVM-RBF, Gaussian Naïve Bayes, K-Nearest Neighbor, Random Forest, AdaBoost

- *Comparison Using All Features*
- *Comparison Using Features Selected by Random Forest*

```
{"features": [  
    "RE",  
    "NIR",  
    "G_mean",  
    "RE_mean",  
    "RE_homogeneity",  
    "RE_ASM",  
    "RE_energy",  
    "RE_entropy",  
    "NIR_mean",  
    "NIR_contrast",  
    "NIR_homogeneity",  
    "NIR_ASM",  
    "NIR_energy",  
    "NIR_entropy"  
]}
```

## Model Comparison: Features selected by Random Forest



# Comparison of Different Classifiers

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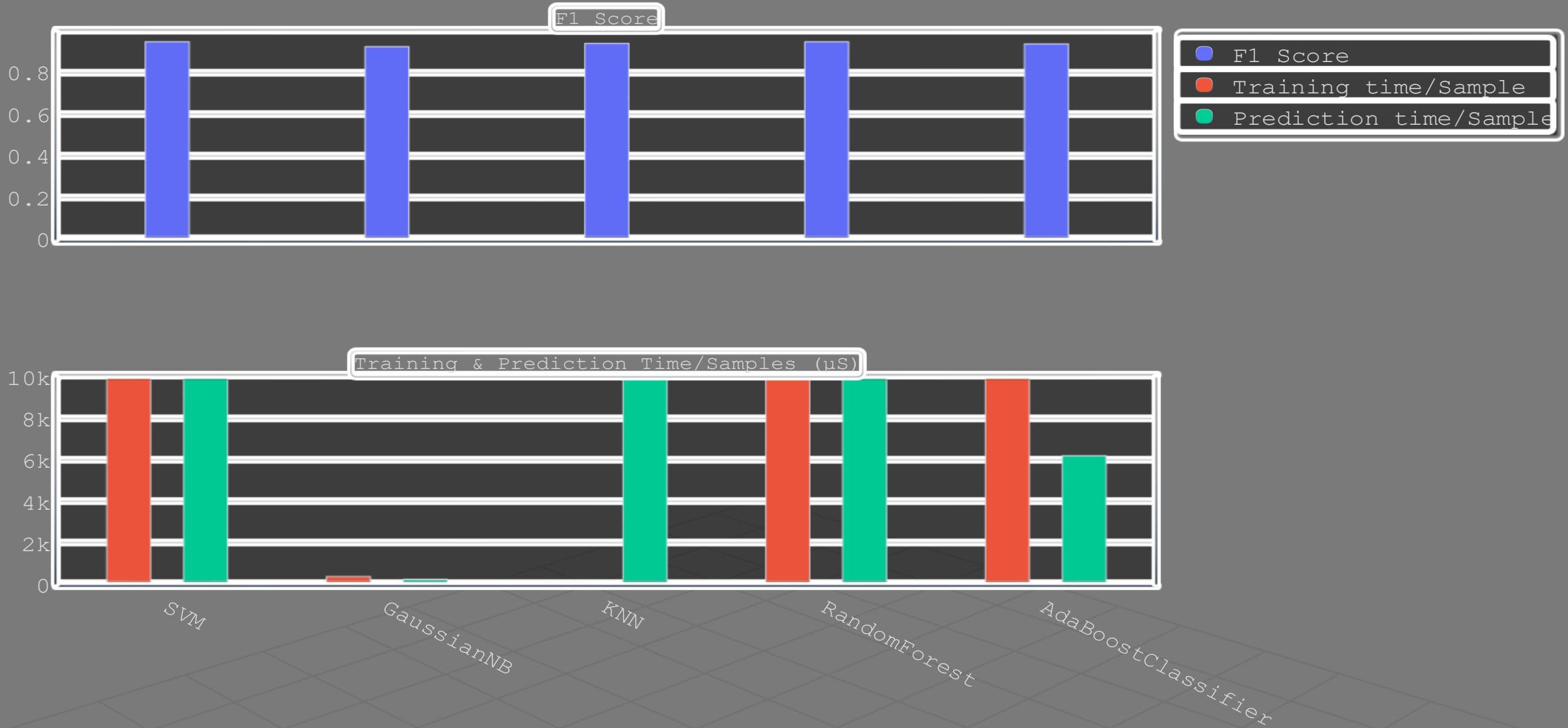
Comparison of baseline models in terms of F1 Score, training and prediction time.

Machine Learning Models: SVM-RBF, Gaussian Naïve Bayes, K-Nearest Neighbor, Random Forest, AdaBoost

- *Comparison Using All Features*
- *Comparison Using Features Selected by Random Forest*
- *Comparison Using Features Selected by AdaBoost*

```
{"features": [  
    "G",  
    "RE",  
    "NIR",  
    "GNDVI",  
    "RRVI",  
    "R_std",  
    "R_energy",  
    "R_entropy",  
    "G_mean",  
    "G_std",  
    "G_contrast",  
    "G_homogeneity",  
    "B_mean",  
    "B_std",  
    "RE_mean",  
    "RE_contrast",  
    "RE_energy",  
    "RE_entropy",  
    "NIR_mean",  
    "NIR_std",  
    "NIR_contrast",  
    "NIR_homogeneity"]}
```

## Model Comparison: Features selected by AdaBoost



# Comparison of Different Classifiers

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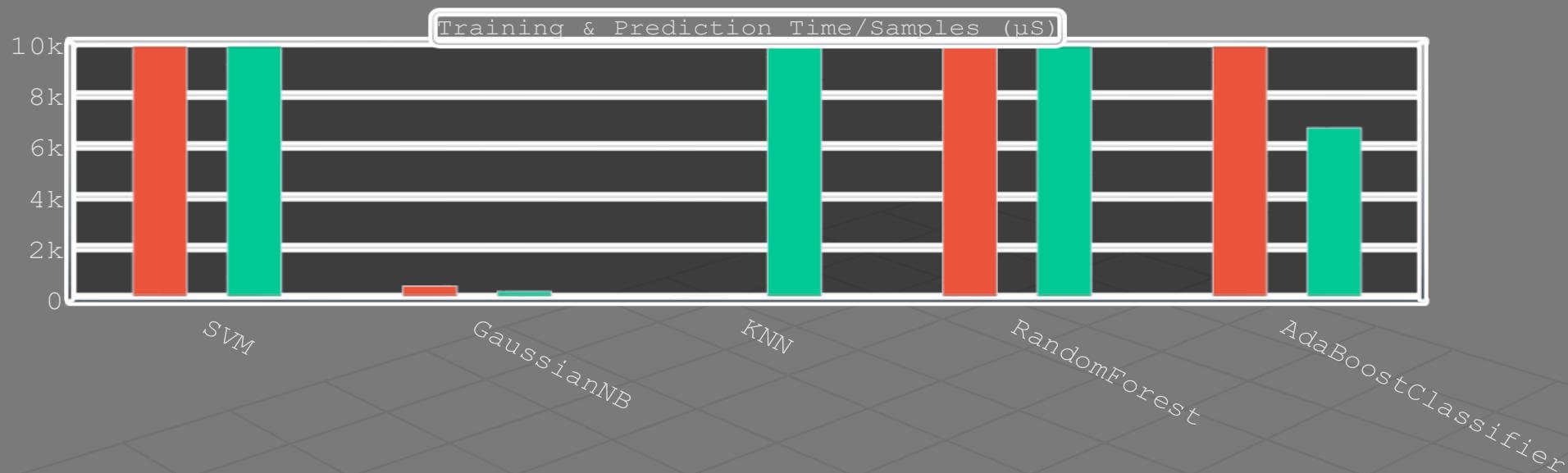
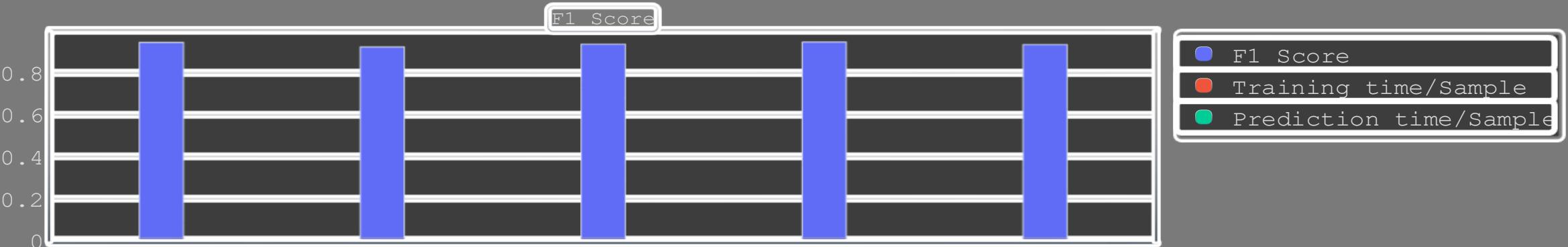
Comparison of baseline models in terms of F1 Score, training and prediction time.

Machine Learning Models: SVM-RBF, Gaussian Naïve Bayes, K-Nearest Neighbor, Random Forest, AdaBoost

- Comparison Using All Features
- Comparison Using Features Selected by Random Forest
- Comparison Using Features Selected by AdaBoost
- Comparison Using Features Selected by ANOVA statistical Test

```
{"features": [  
    "G",  
    "RE",  
    "NIR",  
    "G_mean",  
    "G_contrast",  
    "G_homogeneity",  
    "G_ASM",  
    "G_energy",  
    "G_max",  
    "G_entropy",  
    "B_std",  
    "RE_mean",  
    "RE_contrast",  
    "RE_homogeneity",  
    "RE_ASM",  
    "RE_energy",  
    "RE_max",  
    "RE_entropy",  
    "NIR_mean",  
    "NIR_contrast",  
    "NIR_homogeneity",  
    "NIR_ASM",  
    "NIR_energy",  
    "NIR_max",  
    "NIR_entropy"]}]
```

## Model Comparison: Features selected by ANOVA Method



# Comparison of Different Classifiers

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Comparison of baseline models in terms of F1 Score, training and prediction time.

Machine Learning Models: SVM-RBF, Gaussian Naïve Bayes, K-Nearest Neighbor, Random Forest, AdaBoost

- Comparison Using All Features
- Comparison Using Features Selected by Random Forest
- Comparison Using Features Selected by AdaBoost
- Comparison Using Features Selected by ANOVA statistical Test

For this task, training dataset is used by performing a random split.

Data Preprocessing: Standardization  $\rightarrow x' = \frac{x - \mu}{\sigma}$

# Training of Selected Classifier

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Selection: Gaussian Naïve Bayes combined with Features Selected By Random Forest.

How well a classifier generalizes ?

Which is the range of expected errors of the classifier ?

Mean  $F1$ : 0.918 | Sigma  $F1$ : 0.003

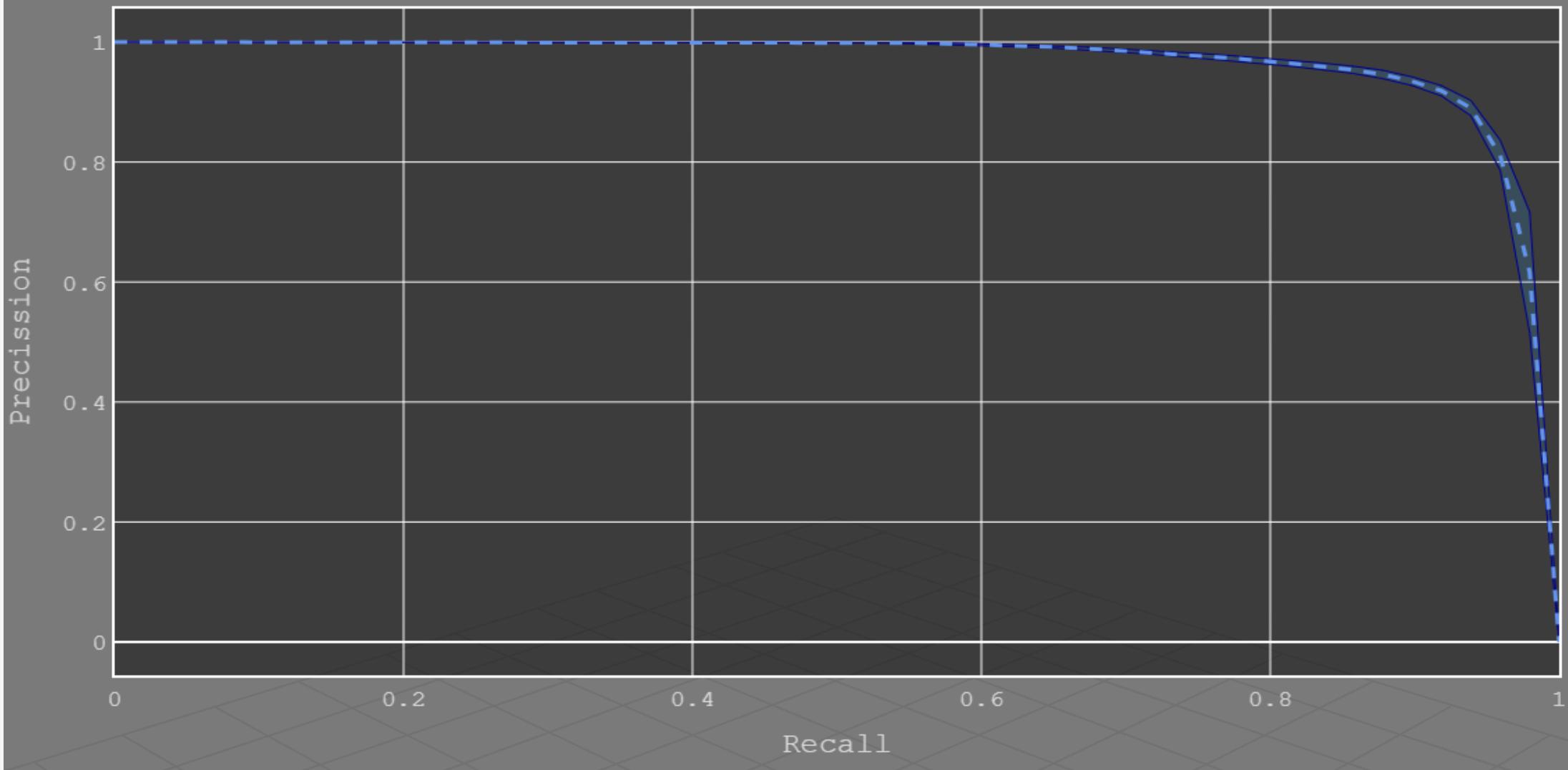
Selection of decision threshold →  $\max(\mathbf{TPR} - \mathbf{FPR})$

Finally, the entire dataset used to train the model for the last time.

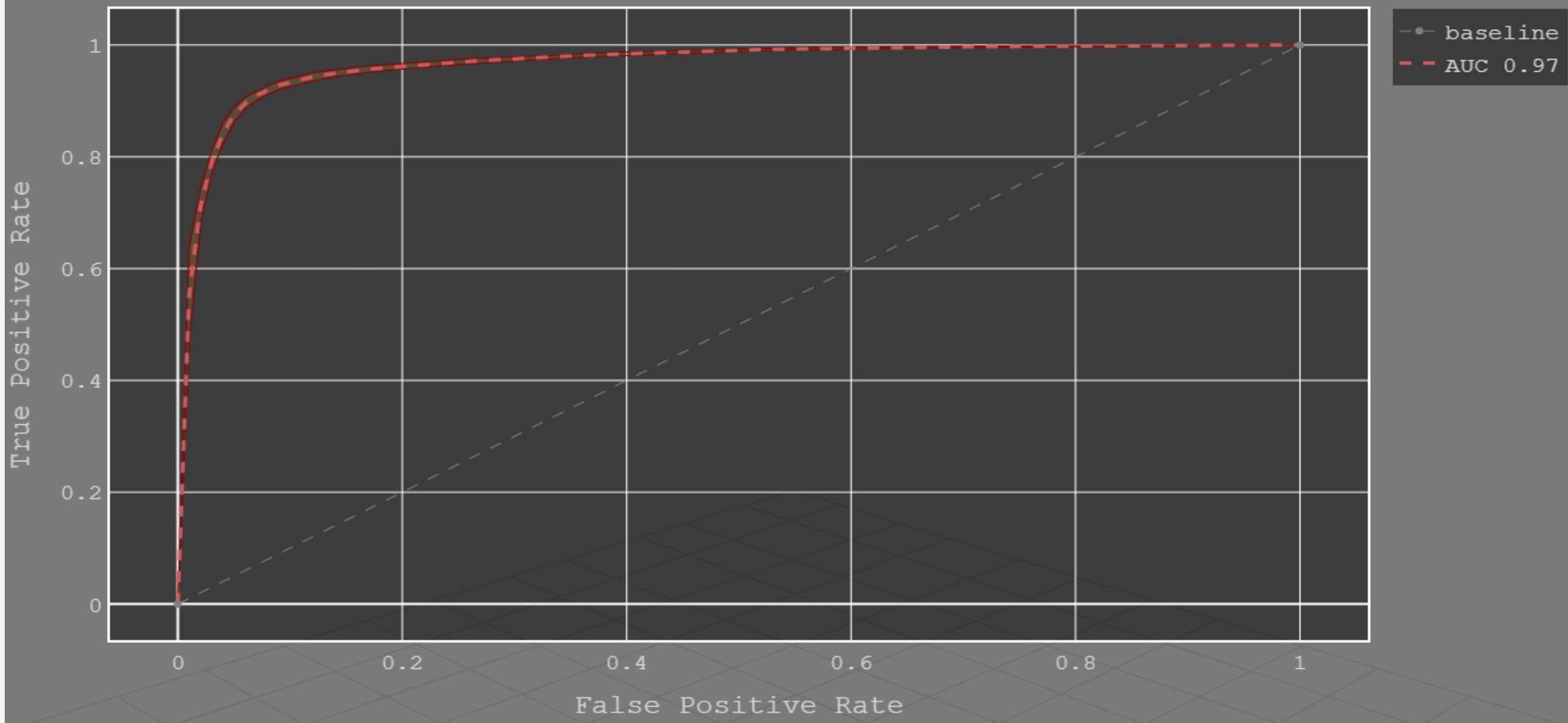


Cross Validation

## Cross Validation - Precision/Recall Curve

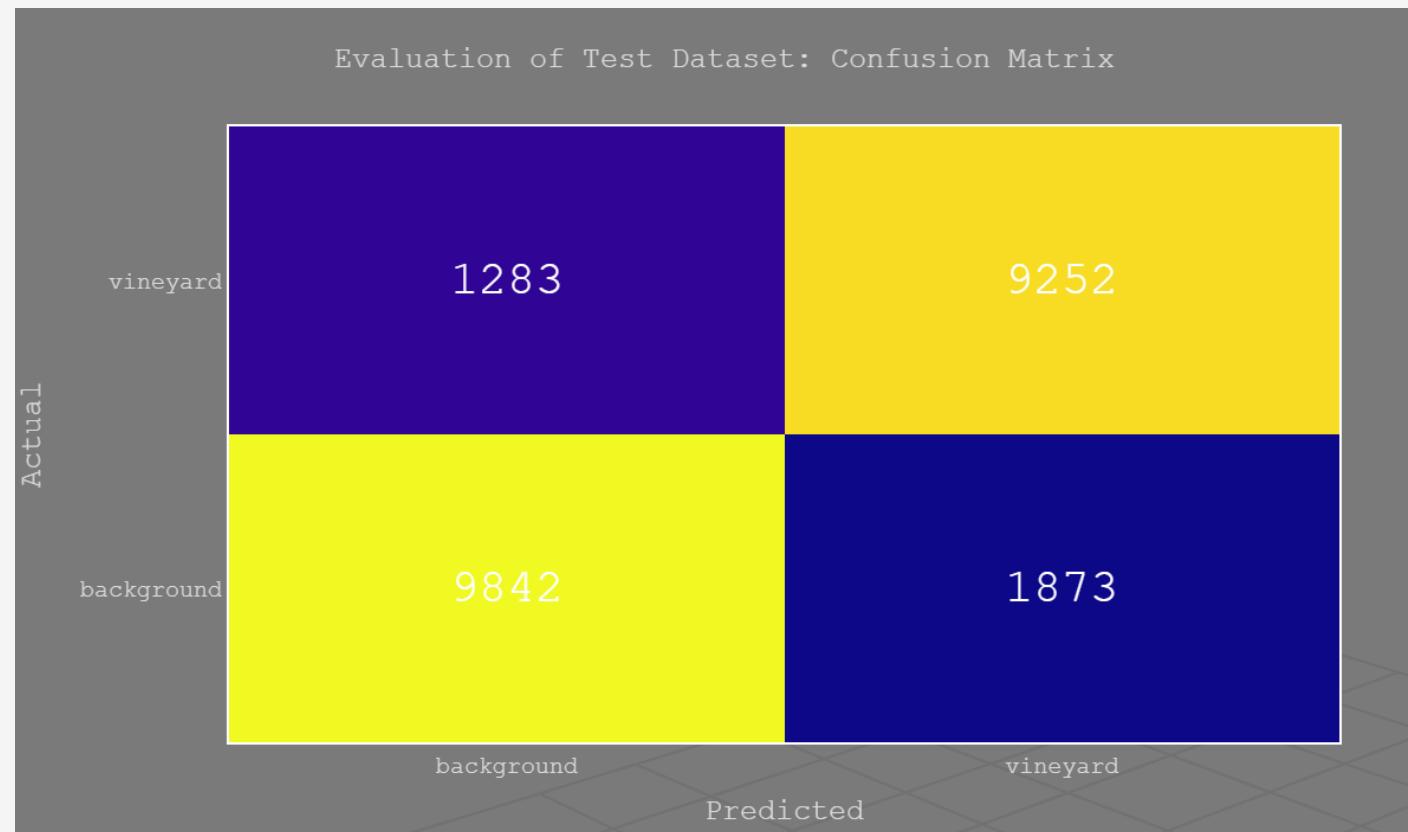


## Cross Validation - ROC Curve



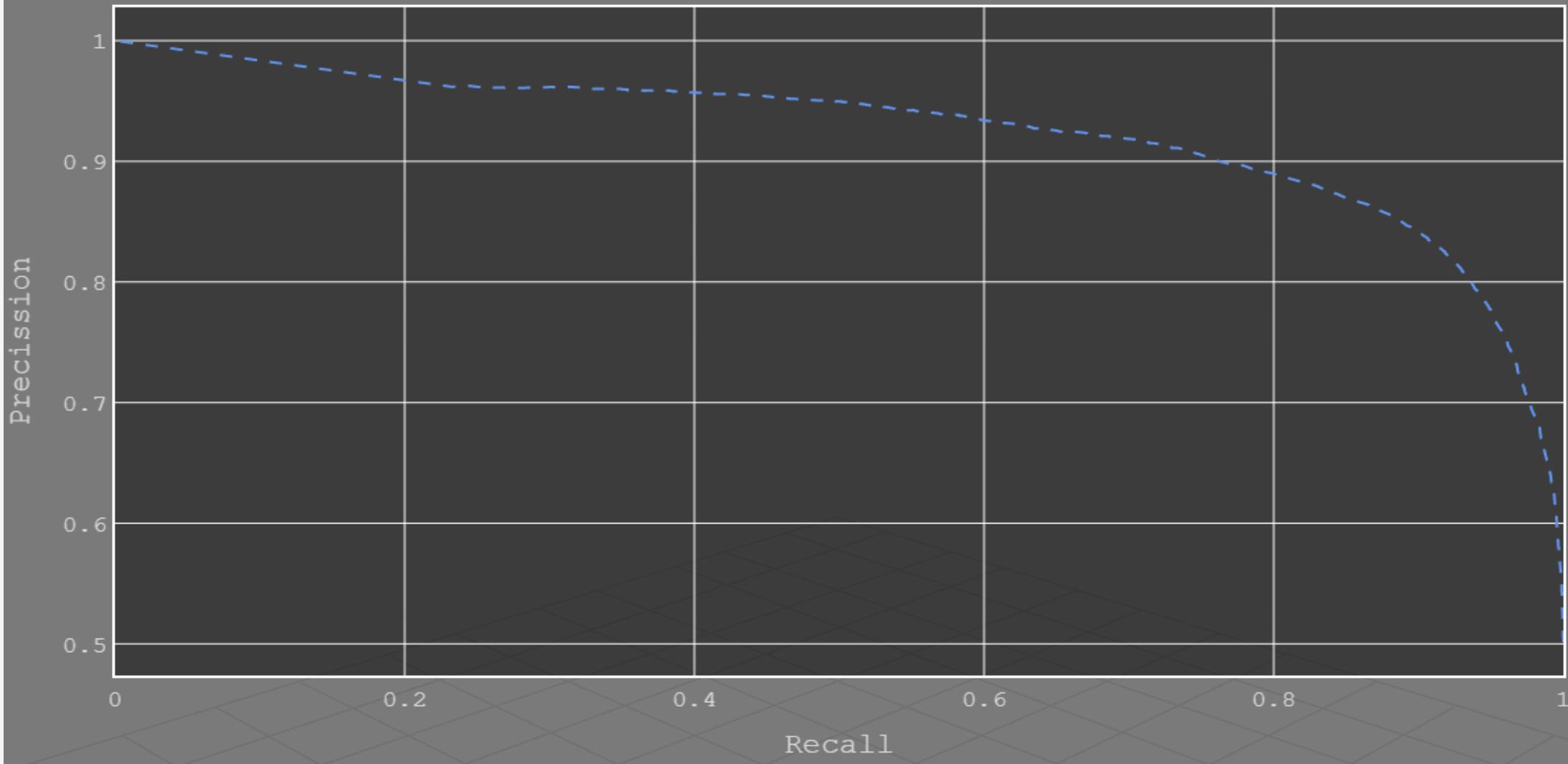
# Using Classifier to predict unknown data (Area 3)

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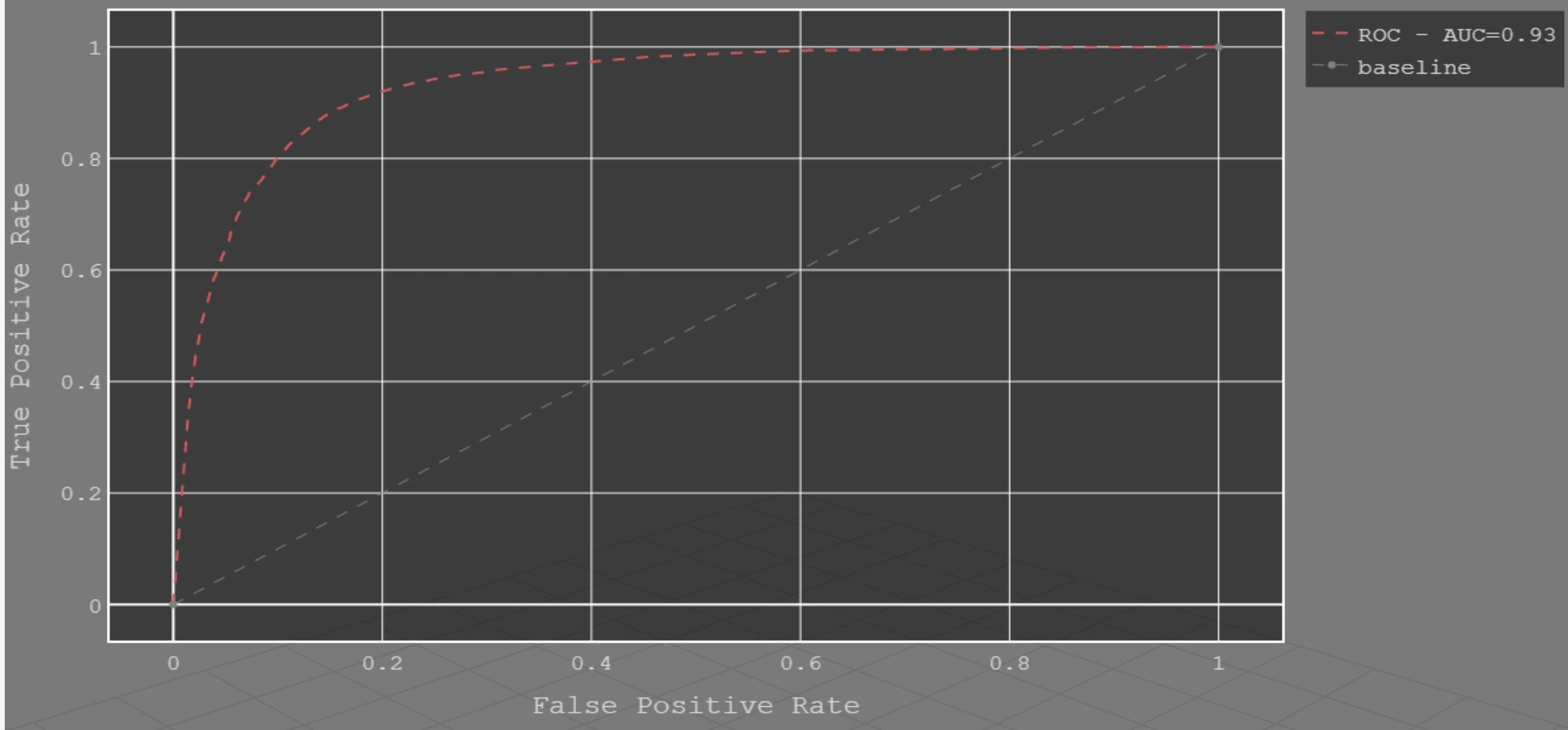


	precision	recall	f1-score
0	0.84	0.88	0.86
1	0.87	0.83	0.85
accuracy			0.85
macro avg	0.85	0.85	0.85
weighted avg	0.85	0.85	0.85

## Evaluation of Test Dataset: Precision/Recall Curve

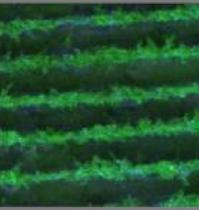


## Evaluation of Test Dataset: ROC Curve

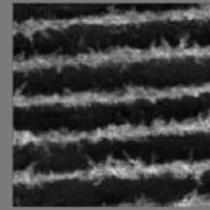


## Predictions from Test Dataset

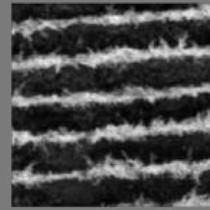
RGB Image



Red Edge Band



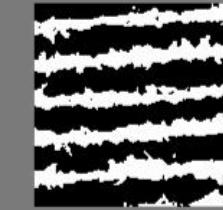
NIR Band



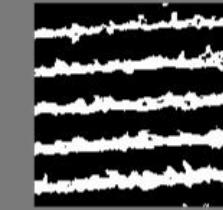
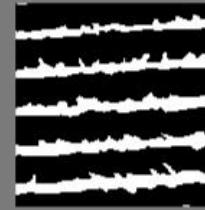
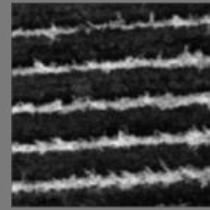
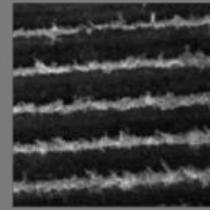
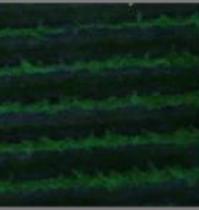
Ground Truth



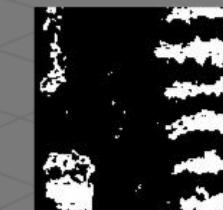
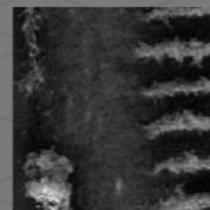
Precision



Dice Score = 0.86



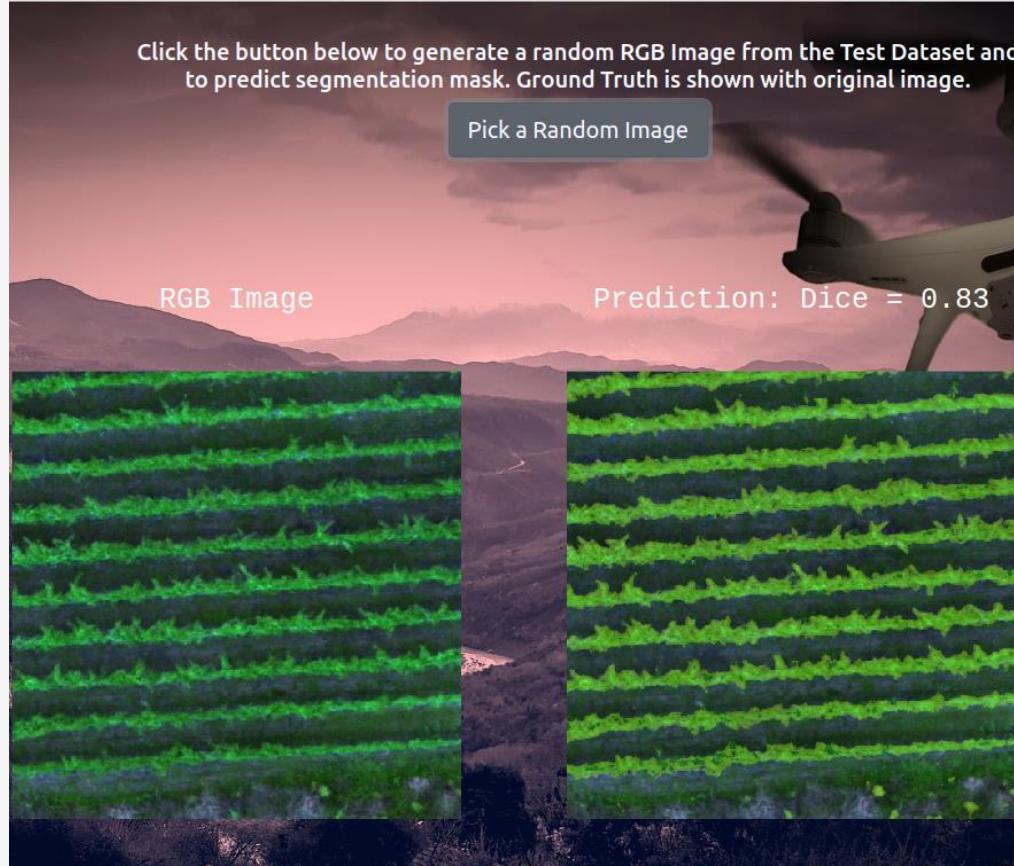
Dice Score = 0.79



Dice Score = 0.62

# Application

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# Application

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Thank you!

