

# RICE DISEASE CLASSIFIER

Computer Vision

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- Rice is one of the staple foods of most countries in Asia especially in South East region.
- Rice hold a significant role. Any disruption due to crop diseases in rice production can significantly affect both local economies and food availability.
- High demand on rice means rice plant sector has to come up with strategic plans to maintain food security and sufficiency.



# Background

# Proposed Solution

## Rice Plant Diseases Classifier App

- Accepts image of rice plant leaf as an input.
- Classifies image into 4 distinct classes.
- Help detect early state of the disease.
- Gives insights regarding the detected disease along side with treatment recommendations.
- Contribute to SDG 2 Zero Hunger by minimizing crop loss and ensuring food security.
- Web based application.



# Data Collection

Kaggle 1 (3000+ images):

- Healthy
- Hispa
- Brownspot
- Bacterial Leaf Blight
- Leaf Blast

Data Mendeley (4000+ images):

- Brownspot
- Bacterial Leaf Blight
- Leaf Blast

Kaggle 2 (2000+ images):

- Healthy
- Narrow Brown Spot
- Brownspot
- Bacterial Leaf Blight
- Leaf Blast
- Leaf Scald

UC Irvine (120 images):

- Brownspot
- Bacterial Leaf Blight
- Leaf Smut

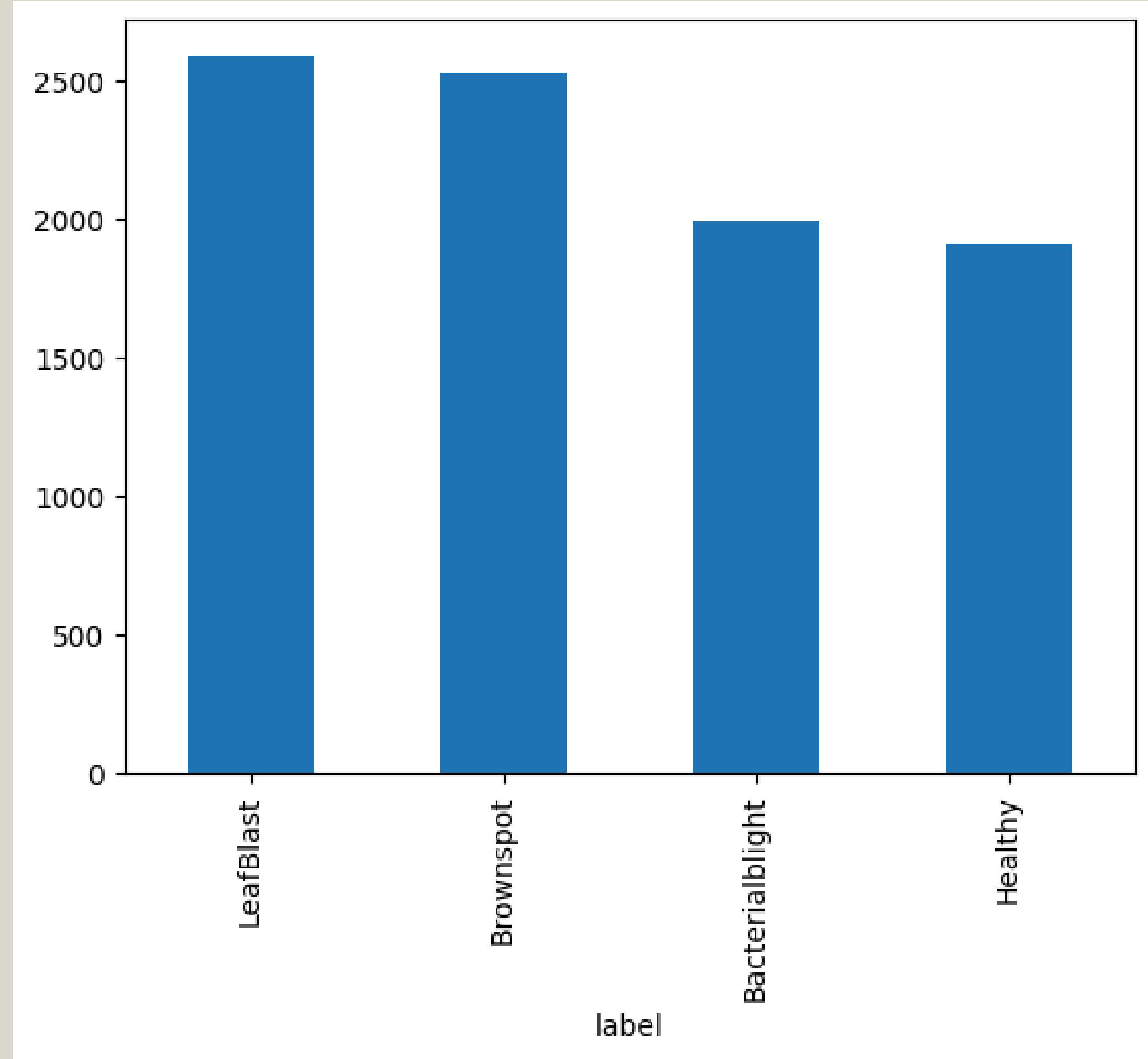
**Classes Picked:**

Healthy, Brownspot, Bacterial Leaf Blight & Leaf Blast

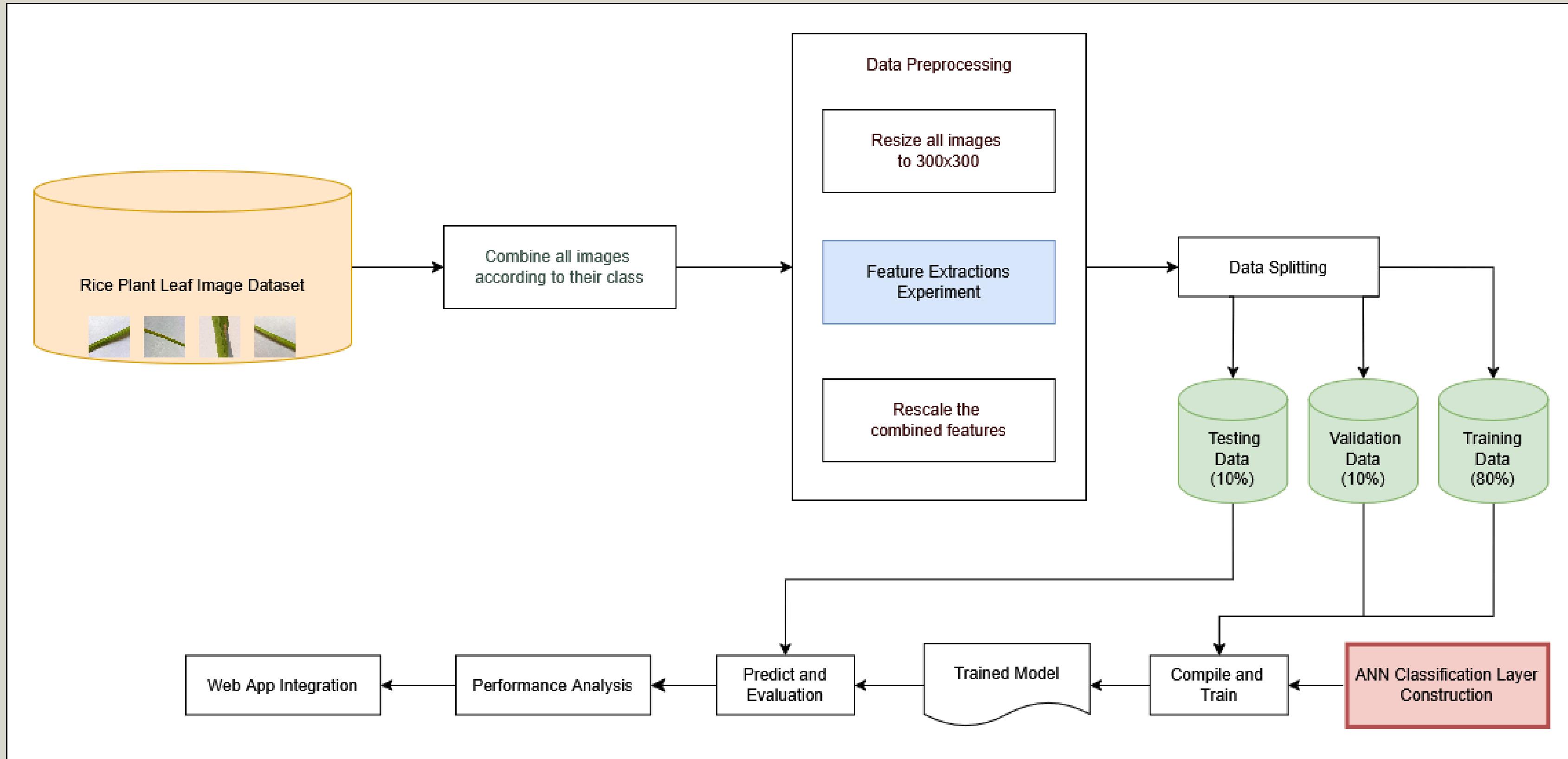
# Data Collection

Image Dataset of Rice Leaf					
	Kaggle 1	Kaggle 2	Mendeley Data	UC Irvine	
Healthy					<b>None</b>
Leaf Blast					<b>None</b>
Bacterial Leaf Blight					
Brown Spot					

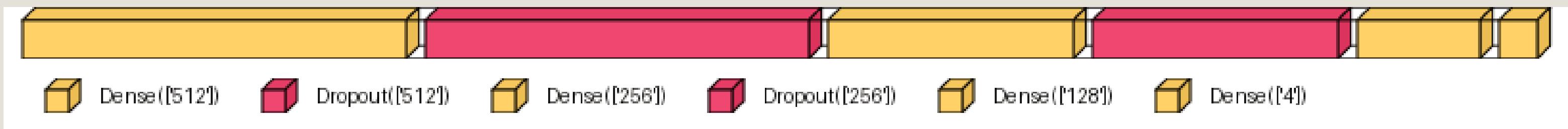
# Data Collection



# Project Pipeline



# ANN Architecture (Classifier)



```
model = Sequential()
input_dim = rescaled_features.shape[1]

model.add(Dense(512, activation='relu', input_shape=(input_dim,)))
model.add(Dropout(0.3))
model.add(Dense(256, activation='relu'))
model.add(Dropout(0.3))
model.add(Dense(128, activation='relu'))
model.add(Dense(num_classes, activation='softmax'))
```

- Optimizer = Adam → learning rate = 0.001
- Loss = Categorical cross entropy

# Experiment - 1



## COLOR (HSV HISTOGRAM)

Processes the image by analyzing its color distribution in terms of Hue, Saturation, and Brightness rather than simple Red Green Blue values.



## SHAPE (HU MOMENTS)

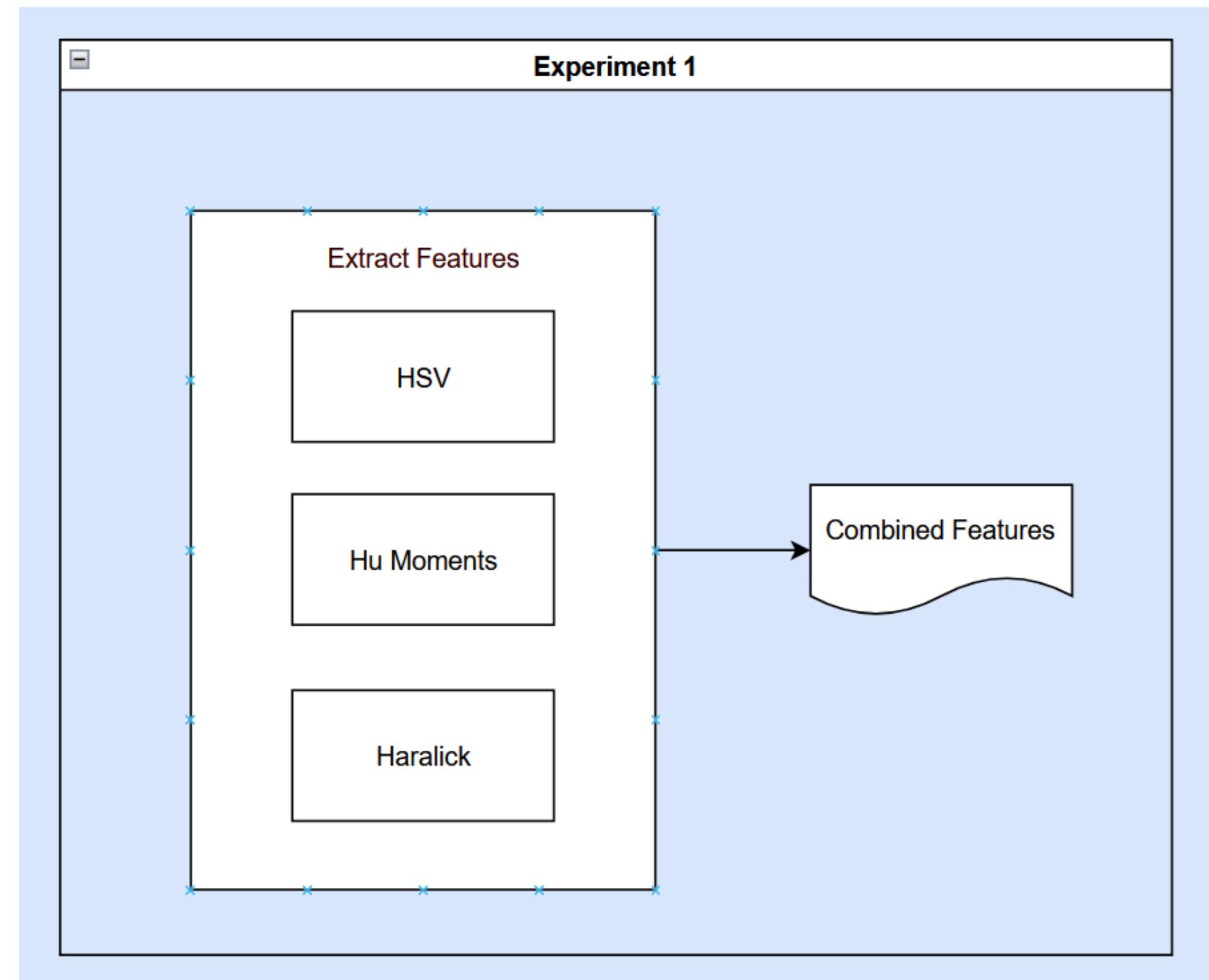
Extract the fundamental geometry of an object by calculating statistical weighted averages of its silhouette or contour



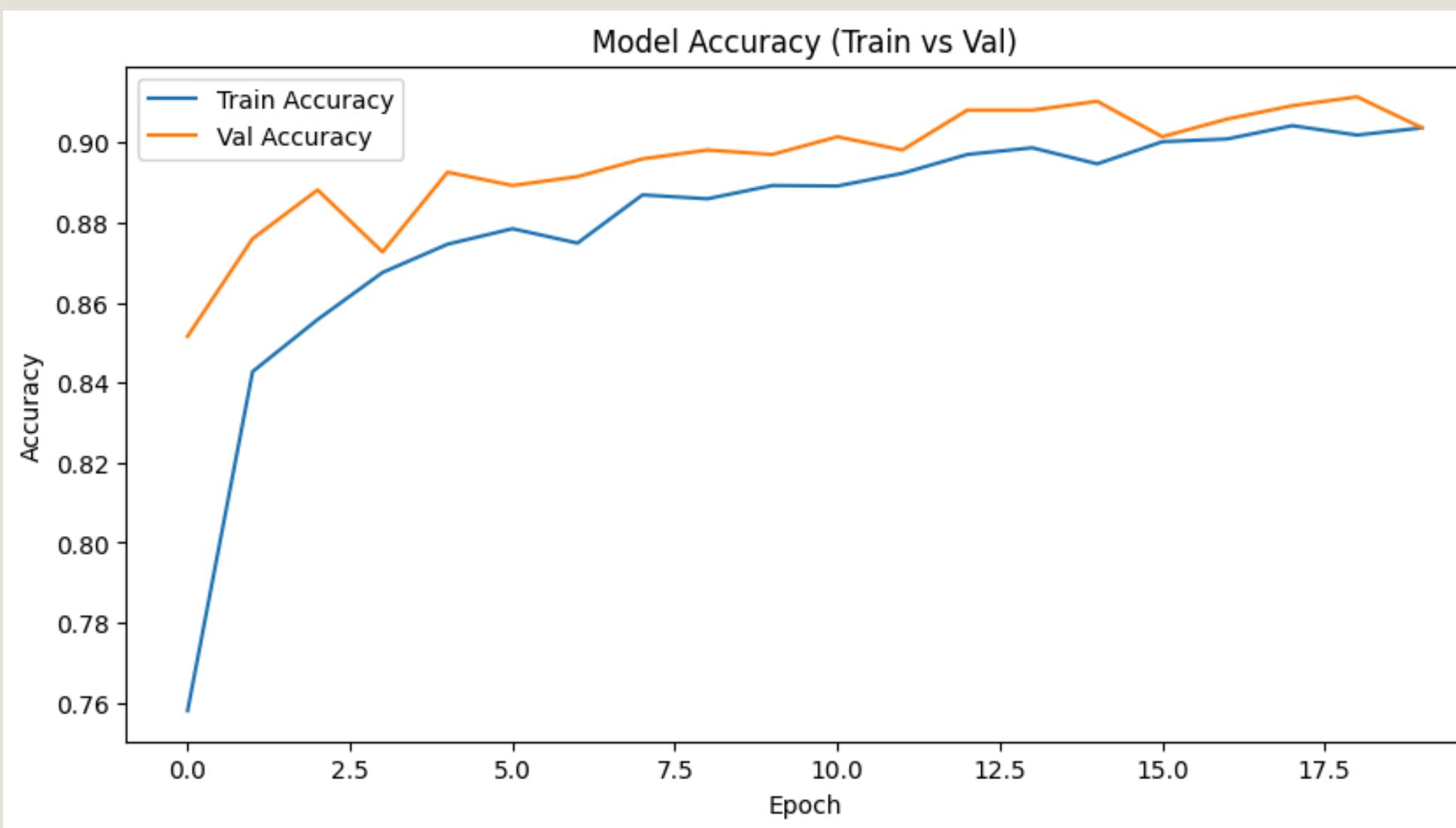
## TEXTURE (HARALICK FEATURES)

A set of statistical numbers that calculated contrast, homogeneity, energy, and correlation

# Experiment - 1



# Experiment - 1



	precision	recall	f1-score	support
Bacterialblight	0.98	1.00	0.99	200
Brownspot	0.93	0.93	0.93	253
Healthy	0.76	0.90	0.82	191
LeafBlast	0.92	0.79	0.85	259
accuracy			0.90	903
macro avg	0.90	0.90	0.90	903
weighted avg	0.90	0.90	0.90	903

# Experiment - 2

## → COLOR (HSV HISTOGRAM)

Processes the image by analyzing its color distribution in terms of Hue, Saturation, and Brightness rather than simple Red Green Blue values.

## → SHAPE (HU MOMENTS)

Extract the fundamental geometry of an object by calculating statistical weighted averages of its silhouette or contour

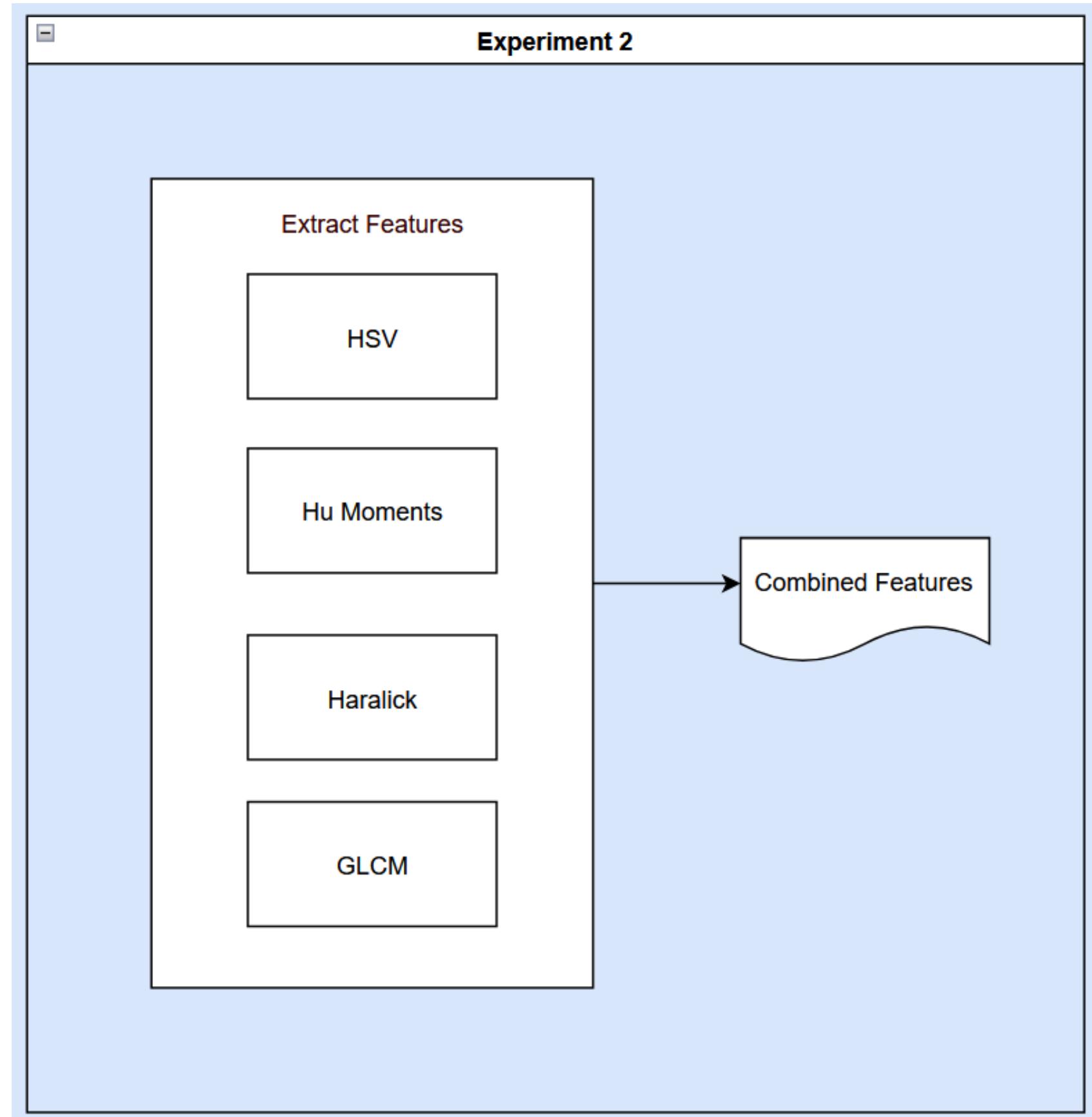
## → TEXTURE (HARALICK FEATURES)

A set of statistical numbers that calculated contrast, homogeneity, energy, and correlation

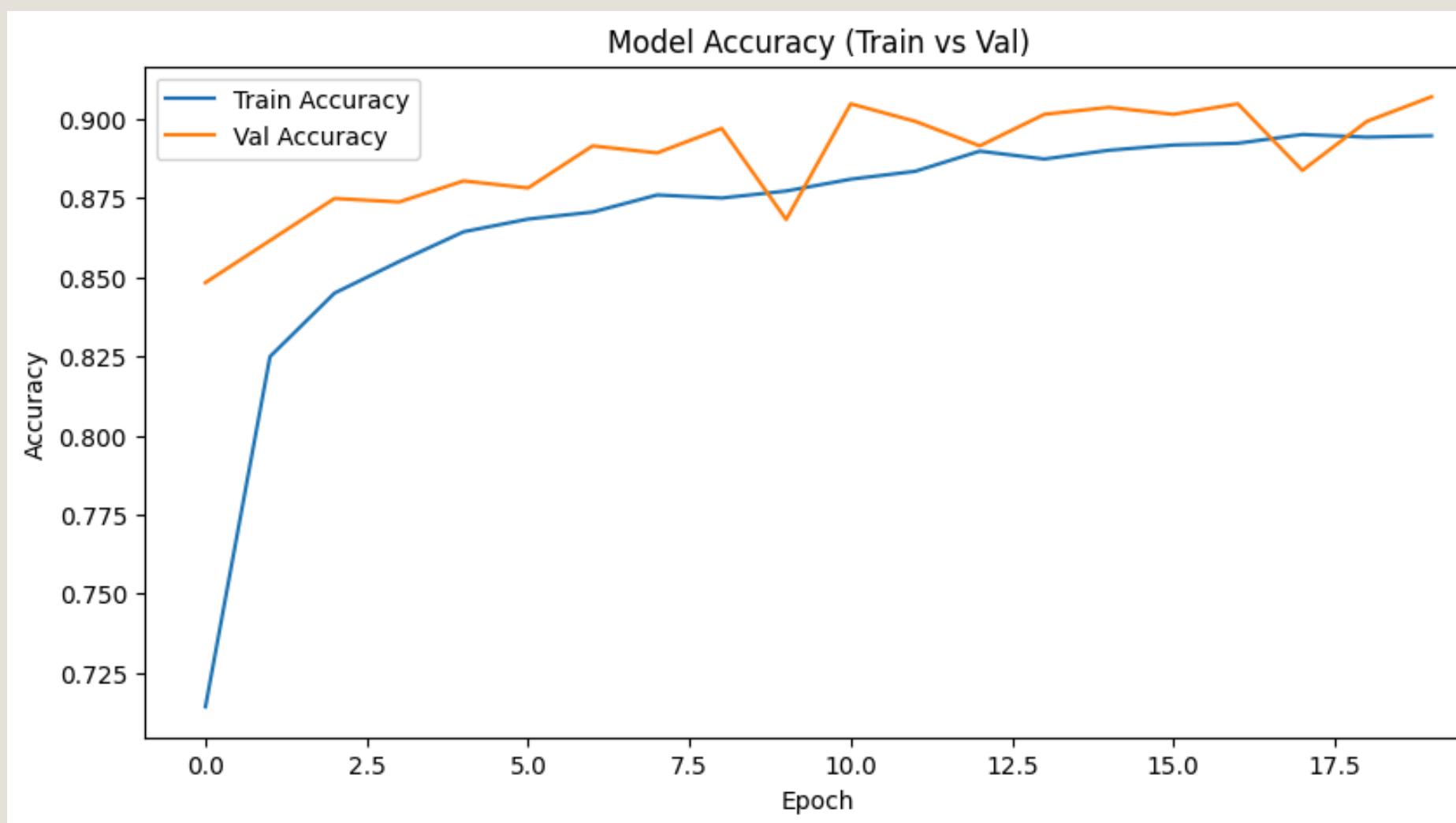
## → TEXTURE (GLCM)

A matrix that counts how often pixel patterns occur next to each other and to captures the spatial relation between pixels.

# Experiment - 2



# Experiment - 2



	precision	recall	f1-score	support
Bacterialblight	0.97	1.00	0.99	200
Brownspot	0.95	0.91	0.93	253
Healthy	0.74	0.91	0.81	191
LeafBlast	0.93	0.78	0.85	259
accuracy			0.89	903
macro avg	0.90	0.90	0.89	903
weighted avg	0.90	0.89	0.89	903

# Experiment - 3

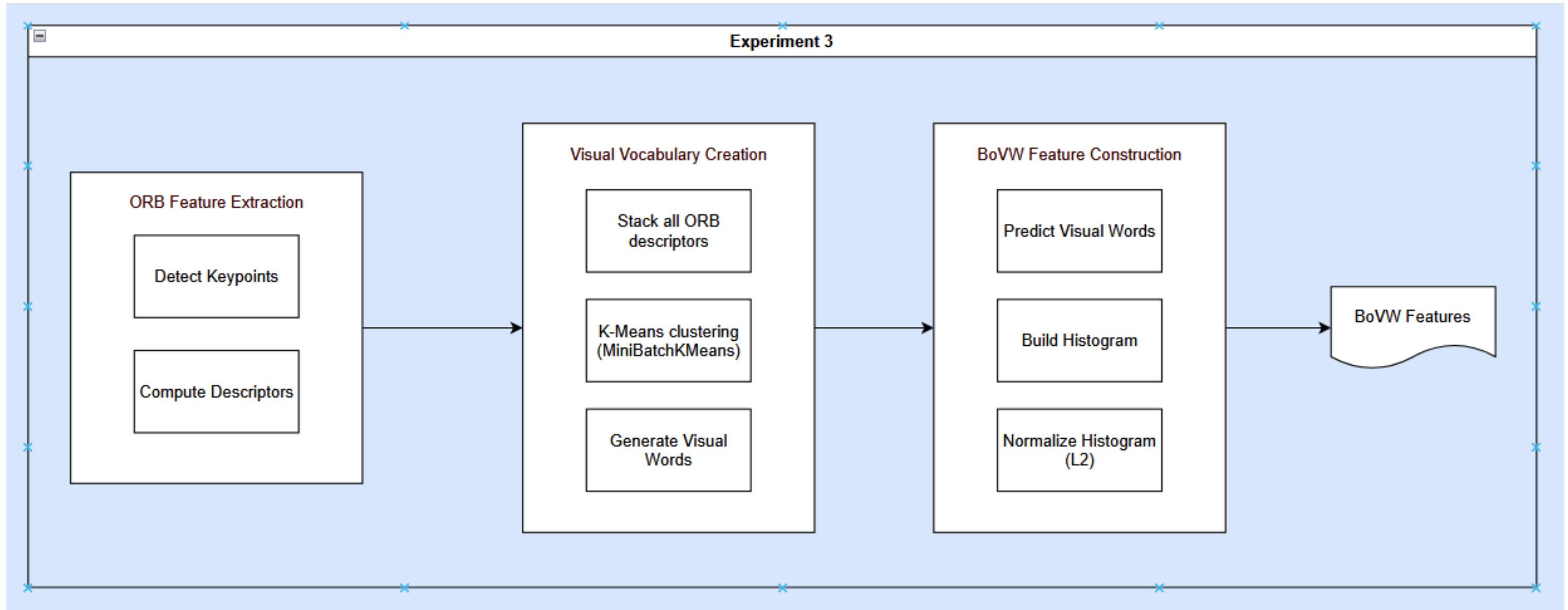
→ ORB

To scans the image to detect and describe specific keypoints, such as sharp corners, edges, or distinct points of interest

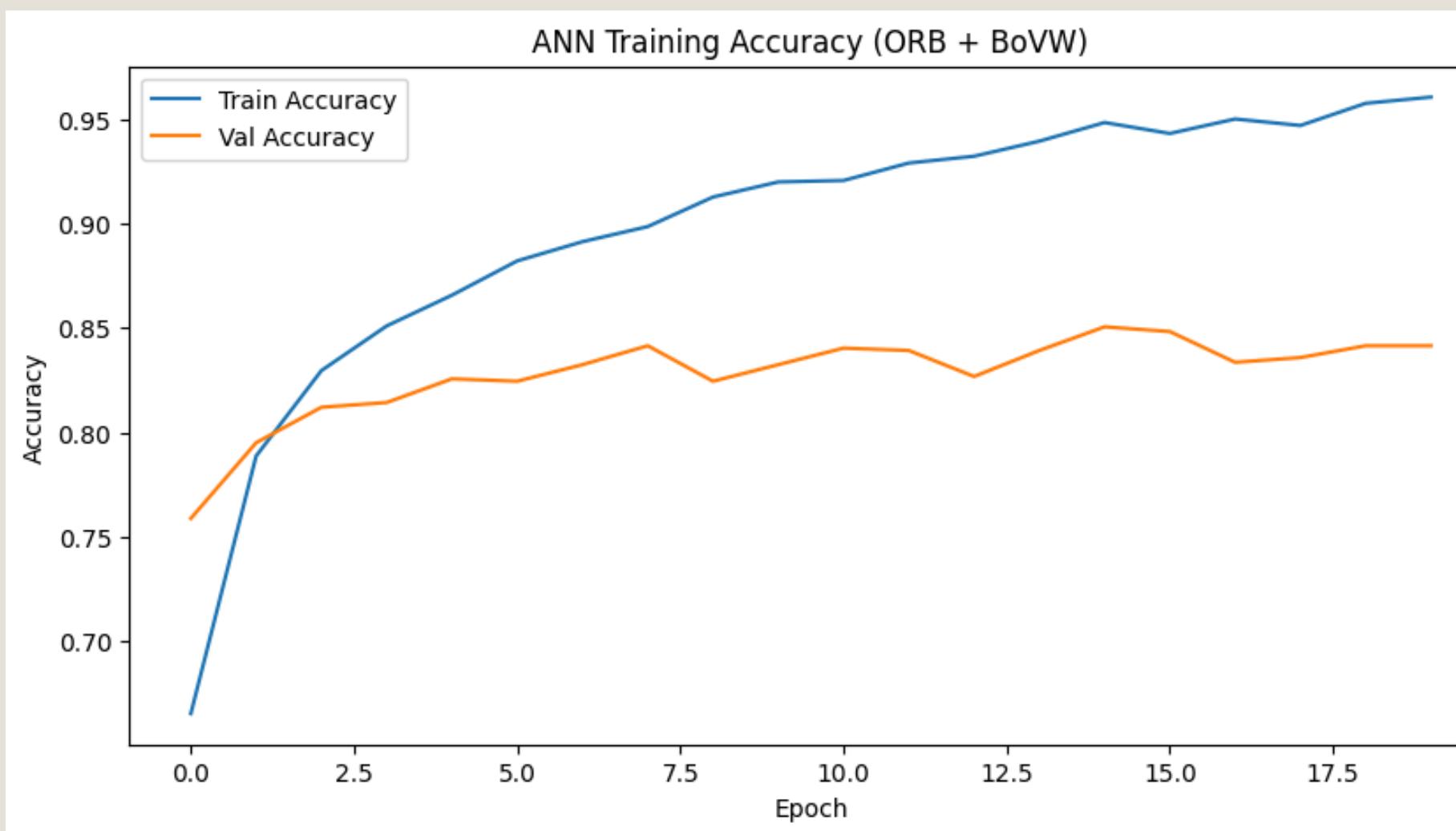
→ BOVW

To groups similar patterns into clusters and counts their frequency, transforming the scattered local details of an image into a fixed length histogram that the Neural Network can read

# Experiment - 3



# Experiment - 3

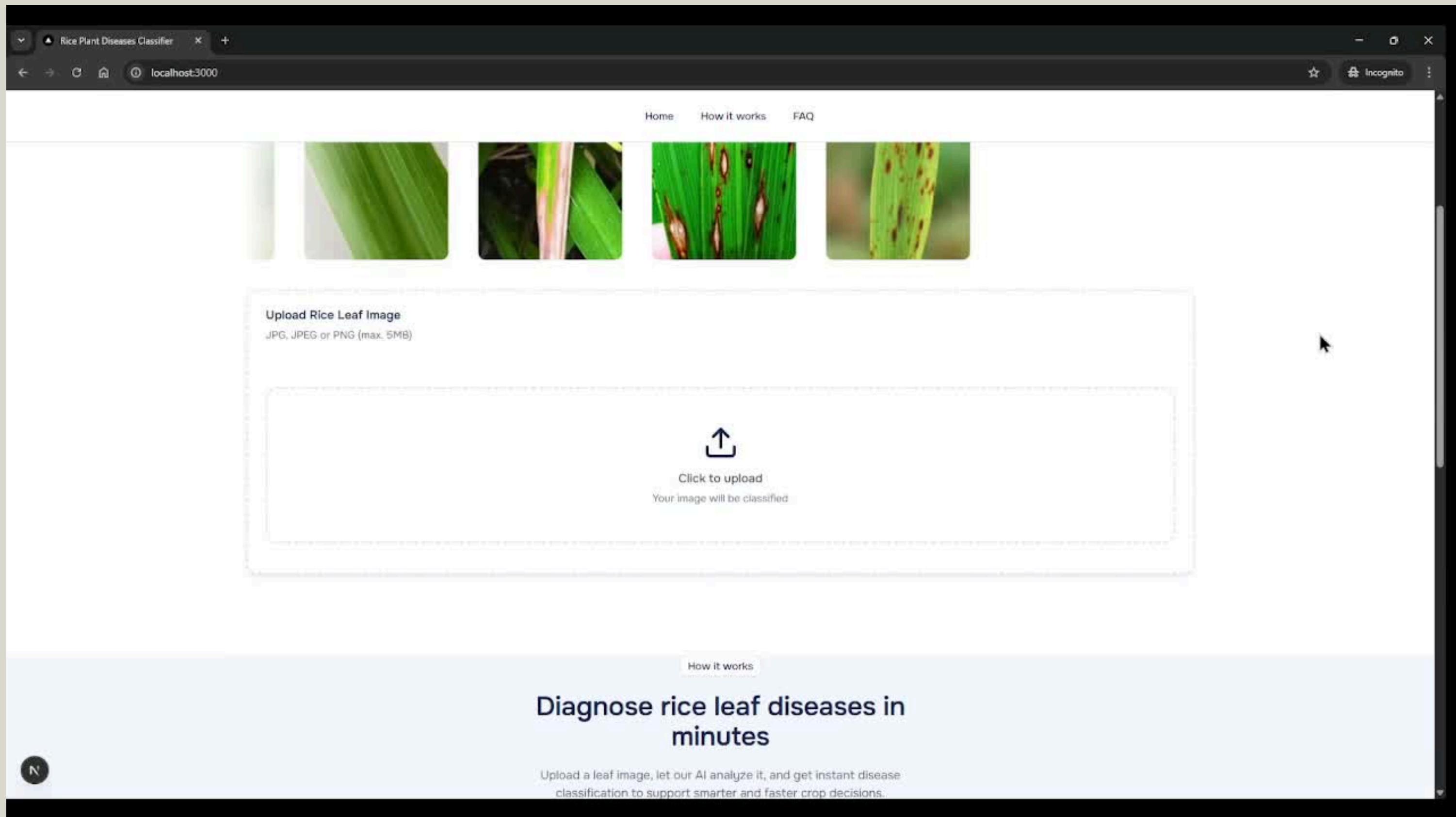


Classification Report:

	precision	recall	f1-score	support
Bacterialblight	0.95	0.94	0.95	200
Brownspot	0.90	0.84	0.87	244
Healthy	0.74	0.90	0.81	188
LeafBlast	0.87	0.79	0.83	252
accuracy			0.86	884
macro avg	0.87	0.87	0.86	884
weighted avg	0.87	0.86	0.86	884

# DEMO

[Link Gdrive](#)



# Contributors

## Pieter Allen

- Research and clean data
- Try different feature extraction techniques
- Train and Test
- Organize Github repository
- Build Frontend

## Fernando Gunawan

- Try different feature extraction techniques
- Construct ANN classifier architecture
- Train and Test
- Build Backend
- Demo

# Thank You.



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