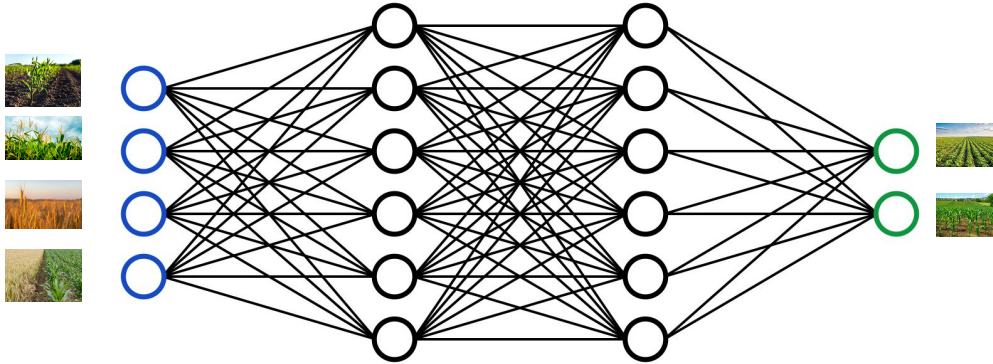
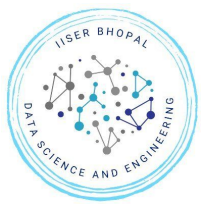


Project DSE316

Crop Recommendation using Machine Learning



By:
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Gaurav Kumar Jha
Rachit Bakshi
Mayank Srivastava



Dataset

	N	P	K	temperature	humidity	ph	rainfall	label
0	90	42	43	20.879744	82.002744	6.502985	202.935536	rice
1	85	58	41	21.770462	80.319644	7.038096	226.655537	rice
2	60	55	44	23.004459	82.320763	7.840207	263.964248	rice
3	74	35	40	26.491096	80.158363	6.980401	242.864034	rice
4	78	42	42	20.130175	81.604873	7.628473	262.717340	rice
...
2195	107	34	32	26.774637	66.413269	6.780064	177.774507	coffee
2196	99	15	27	27.417112	56.636362	6.086922	127.924610	coffee
2197	118	33	30	24.131797	67.225123	6.362608	173.322839	coffee
2198	117	32	34	26.272418	52.127394	6.758793	127.175293	coffee
2199	104	18	30	23.603016	60.396475	6.779833	140.937041	coffee

2200 rows × 8 columns

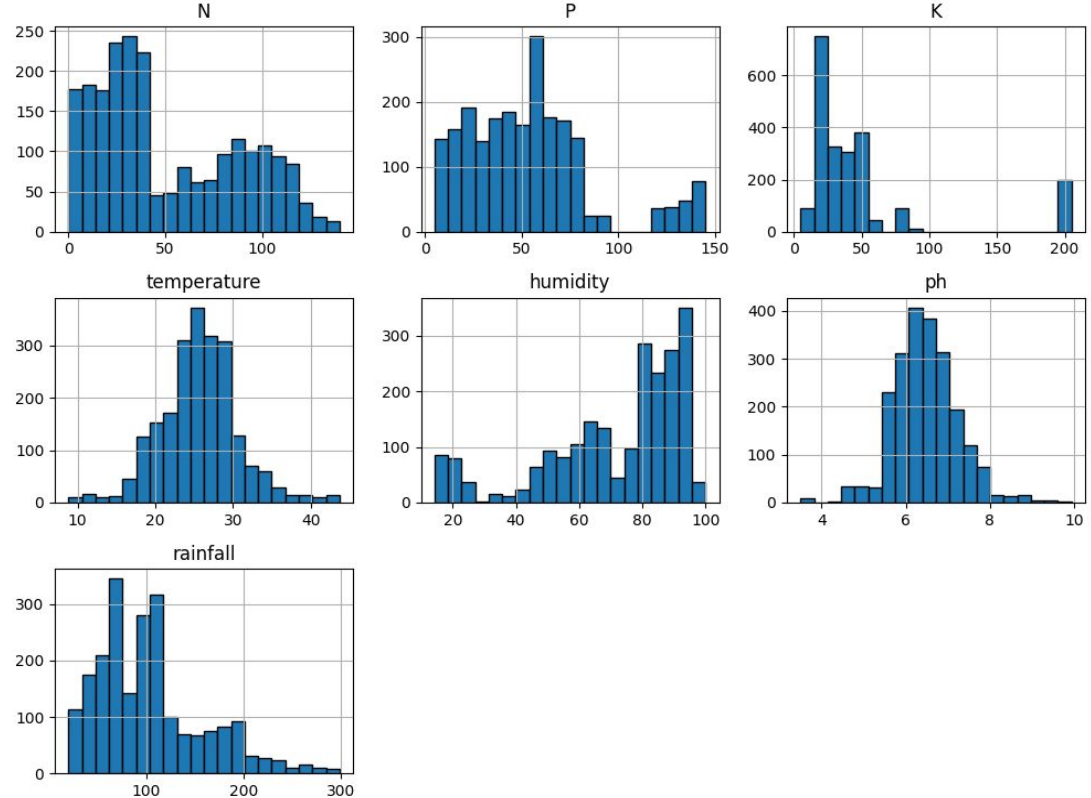
Data Pre-Processing

No Missing Values

No Duplicate rows

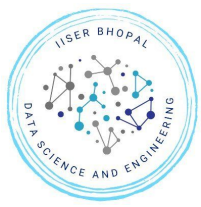
Number of unique labels: 22

Feature Distributions



Data Pre-Processing





Features



Input Features:

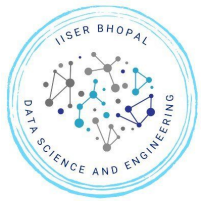
N, P, K,
Temperature,
Humidity, Ph,
Rainfall

N	P	K	temperature	humidity	ph	rainfall
90	42	43	20.879744	82.002744	6.502985	202.935536
85	58	41	21.770462	80.319644	7.038096	226.655537
60	55	44	23.004459	82.320763	7.840207	263.964248
74	35	40	26.491096	80.158363	6.980401	242.864034
78	42	42	20.130175	81.604873	7.628473	262.717340
...
107	34	32	26.774637	66.413269	6.780064	177.774507
99	15	27	27.417112	56.636362	6.086922	127.924610
118	33	30	24.131797	67.225123	6.362608	173.322839
117	32	34	26.272418	52.127394	6.758793	127.175293
104	18	30	23.603016	60.396475	6.779833	140.937041

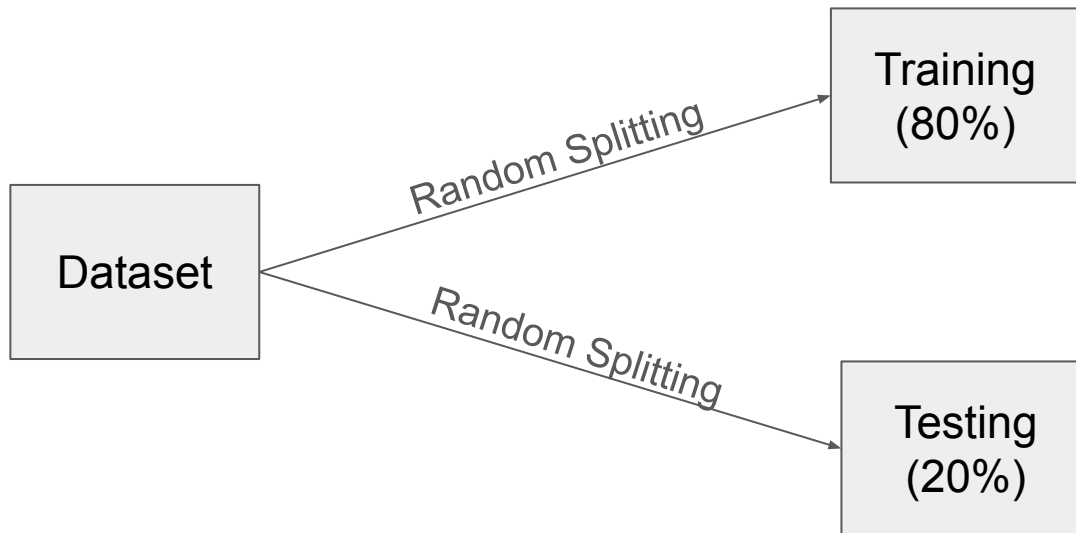
Target Feature: label

```
label
rice      100
maize     100
chickpea  100
kidneybeans 100
pigeonpeas 100
mothbeans 100
mungbean  100
blackgram 100
lentil     100
pomegranate 100
banana     100
mango      100
grapes     100
watermelon 100
muskmelon 100
apple      100
orange     100
papaya     100
coconut    100
cotton     100
jute       100
coffee    100
Name: count, dtype: int64
```



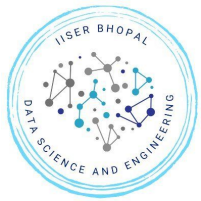


Random Forest



Labels are encoded by Numerical Encoding

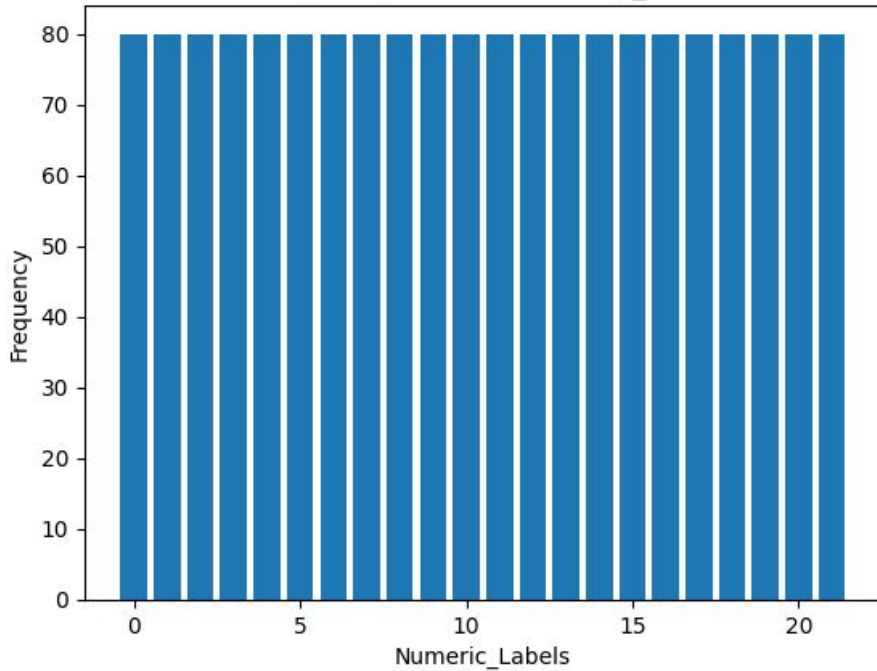
Data	Shape
X_train	(1760, 7)
y_train	(1760,)
X_test	(440, 7)
y_test	(440)



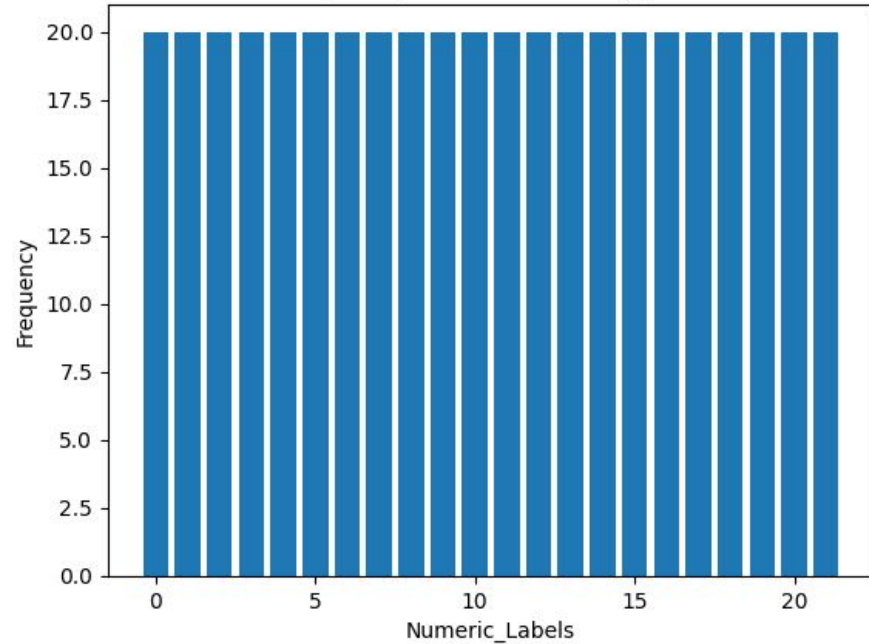
Random Forest

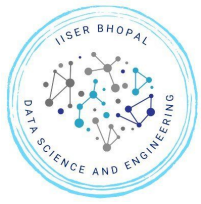


Frequency of Each label in y_train



Frequency of Each label in y_test





Random Forest



N_estimators = 10

Train the model

Testing

Evaluation Metrics

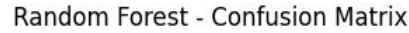
Accuracy: 0.99318181818182

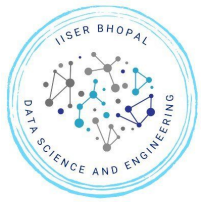
Precision (macro): 0.9937032664305392

Recall (macro): 0.993181818181818

F1 (macro): 0.9931690047222781

F1 (micro): 0.9931818181818182





Tune Random Forest



```
param_grid = {  
    'rf_n_estimators': [10, 20, 30, 40, 50, 70, 80, 100],  
    'rf_max_depth': [None, 20, 30],  
    'rf_min_samples_split': [2, 3, 4],  
    'rf_min_samples_leaf': [1, 2],  
    'rf_class_weight': [None, 'balanced']  
}
```



Stratified K Fold : number of splits = 5, Shuffle = True



GridsearchCV



Fitting 5 folds for each of 288 candidates, totalling 1440 fits

Tune Random Forest

Best hyperparameters:

```
{'rf_class_weight': None, 'rf_max_depth': None, 'rf_min_samples_leaf': 1, 'rf_min_samples_split': 4, 'rf_n_estimators': 30}
```

N_estimators = 30

Train the tuned
model

Testing

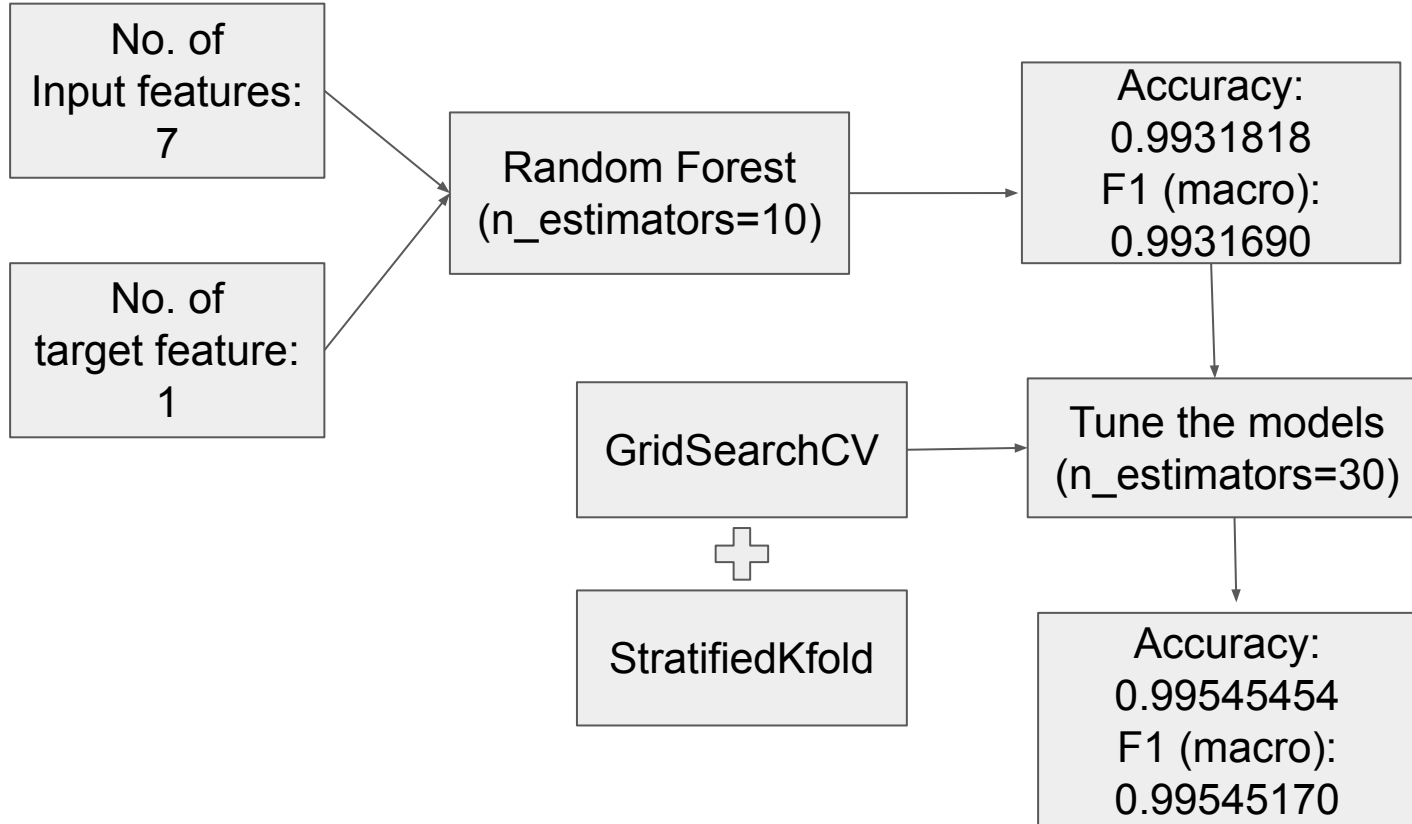
Accuracy: 0.99545454545455

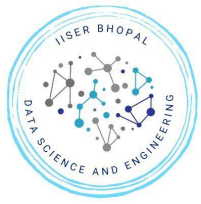
Precision (macro): 0.9956709956709957

Recall (macro): 0.9954545454545454

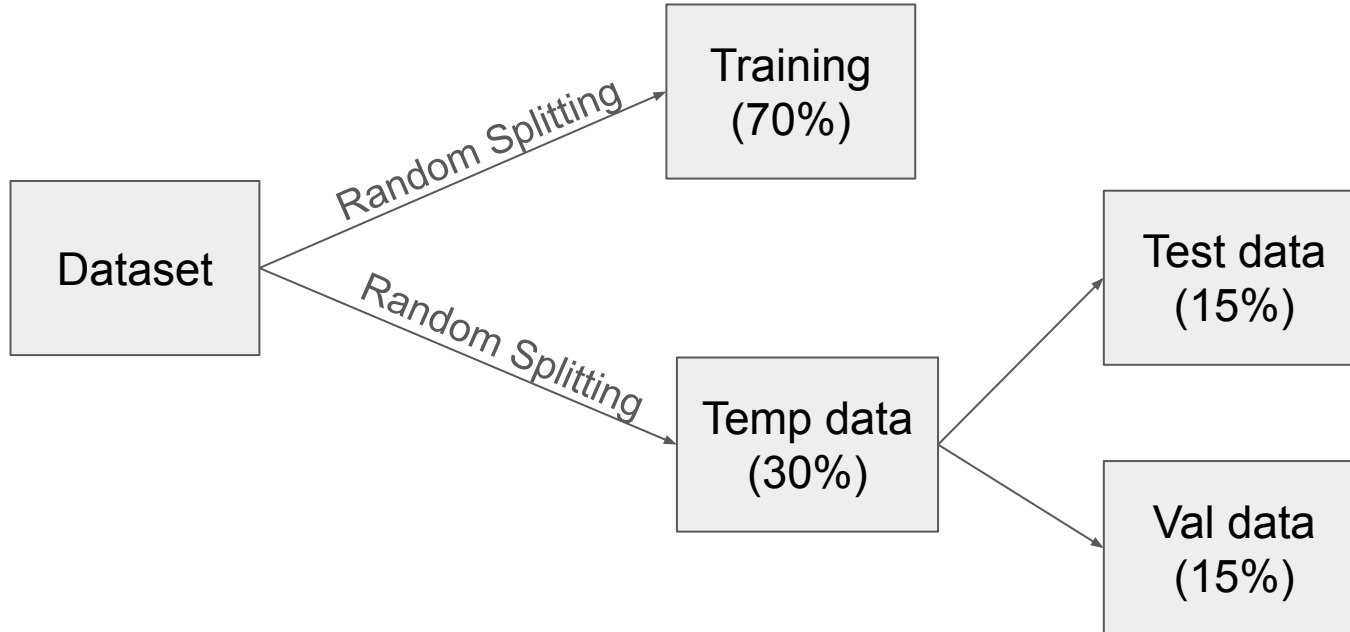
F1 (macro): 0.9954517027687758

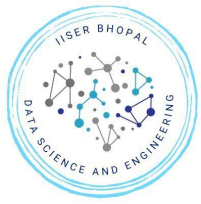
Pipeline of Random Forest





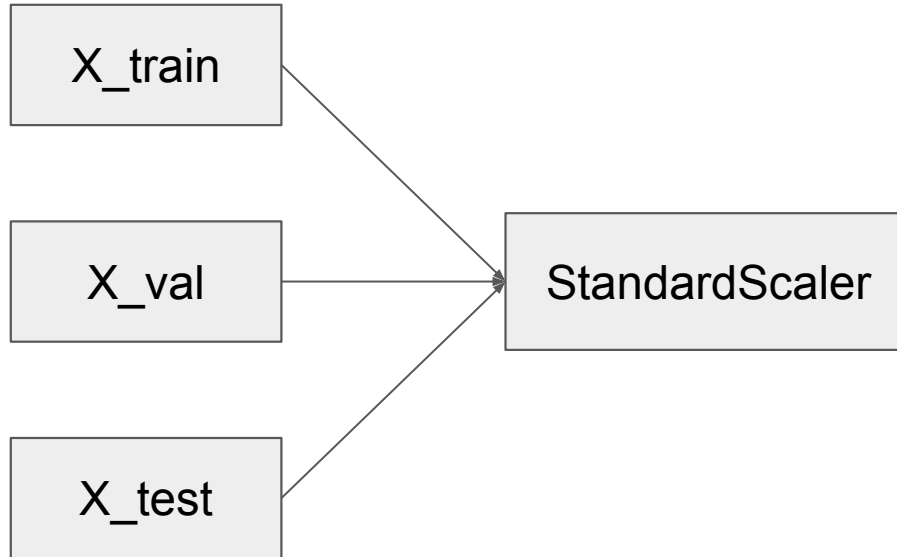
ANNs Data Pre-processing





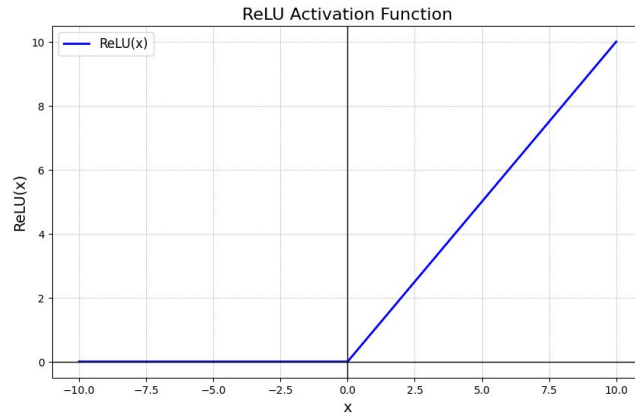
ANNs

Data Pre-processing

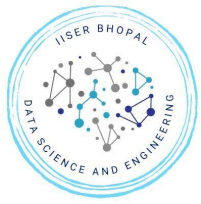


ANNs Model

```
model = keras.Sequential([  
    layers.Dense(8, activation='relu', input_shape=(X_train.shape[1],)),  
    layers.Dropout(0.1),  
    layers.Dense(8, activation='relu'),  
    layers.Dropout(0.1),  
    layers.Dense(8, activation='relu'),  
    layers.Dense(len(np.unique(y_enc)), activation='softmax')  
])
```



$$\sigma(\vec{z})_i = \frac{e^{z_i}}{\sum_{j=1}^K e^{z_j}}$$



ANNs Model

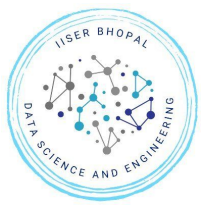
Optimizer: Adam

Loss: Sparse categorical crossentropy

Metrics: Accuracy

Early Stopping

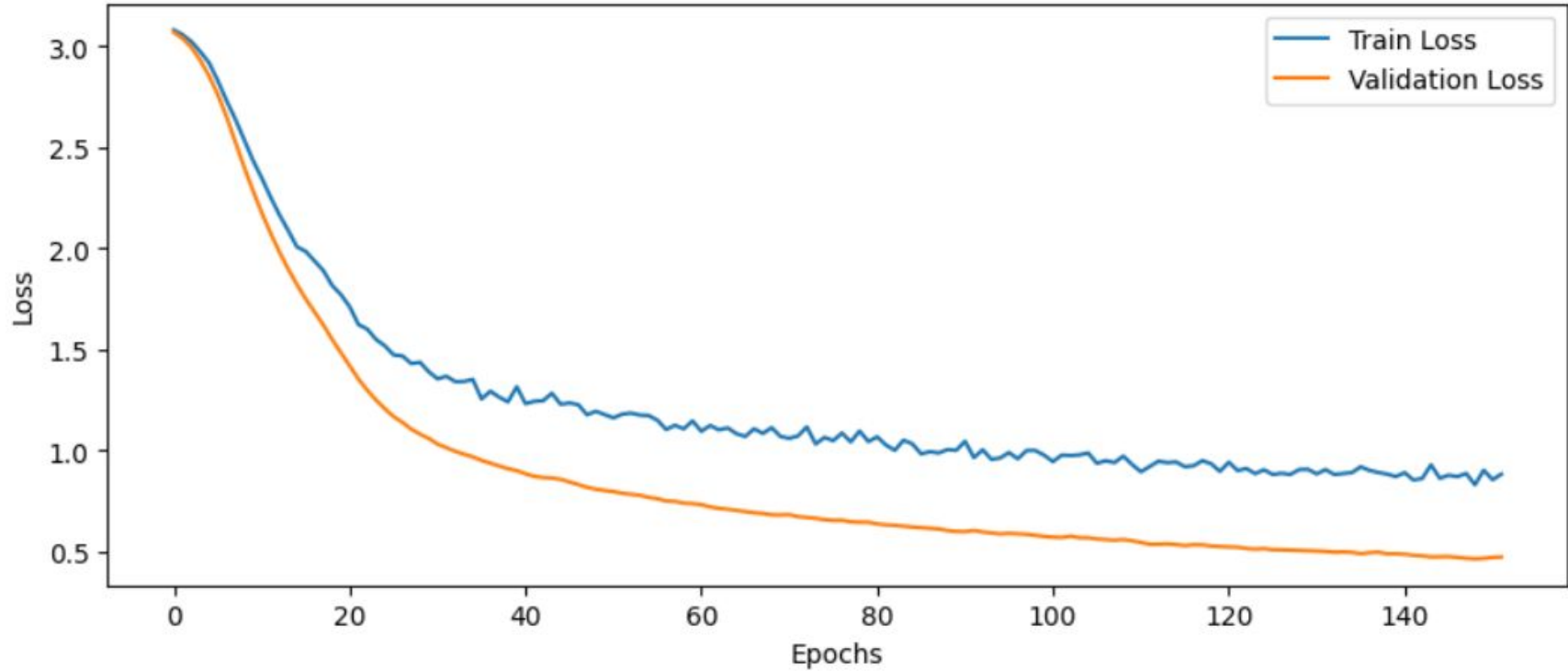
```
history = model.fit(  
    x_train, y_train,  
    epochs=200,  
    batch_size=64,  
    validation_data=(x_val, y_val),  
    callbacks=[early_stop],  
    verbose=1  
)
```

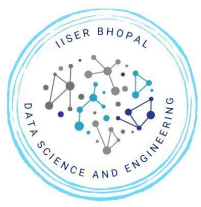


ANNs Model



Model Loss





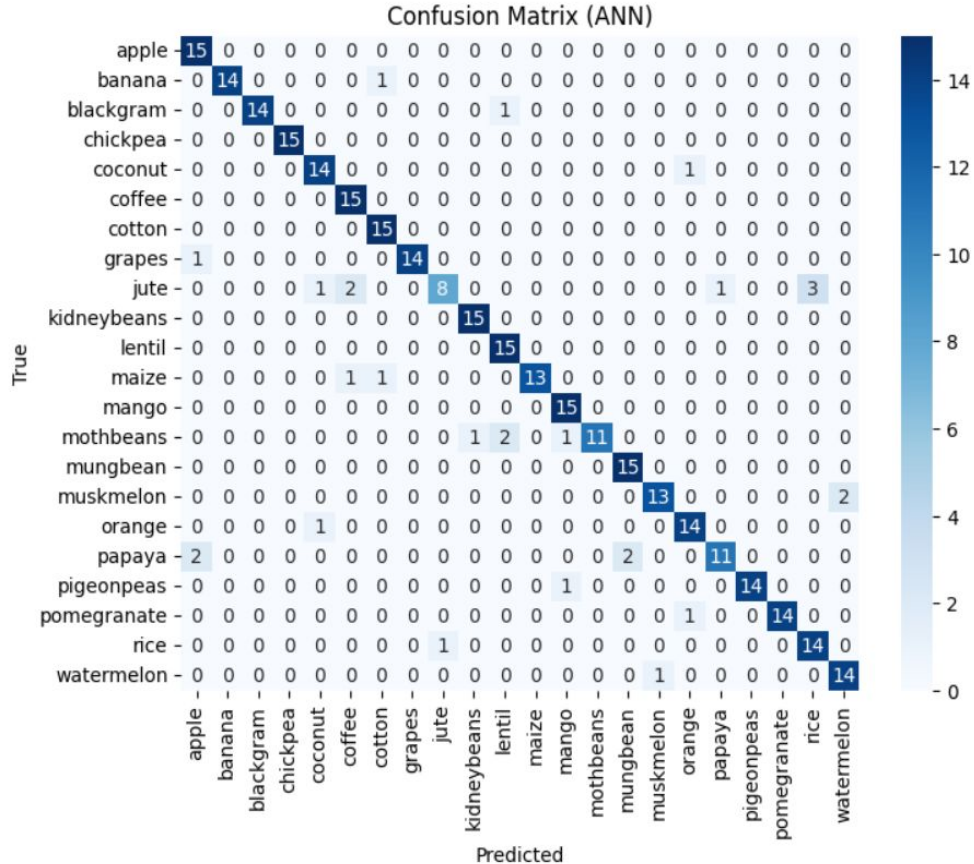
Accuracy: 0.9151515151515152

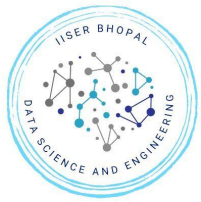
Precision (macro): 0.921237055428232

Recall (macro): 0.9151515151515153

F1 Score: 0.9123880349705759

ANNs Model



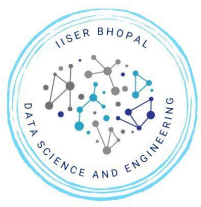


Tuned ANNs Model



```
param_space = {  
    'units': [16, 32, 64],  
    'dropout': [0.1, 0.2],  
    'lr': [1e-3, 1e-4],  
    'n_hidden': [ 2, 3, 4],  
    'batch_size': [32, 64],  
}
```

```
n_splits = 3  
skf = StratifiedKFold(n_splits=n_splits, shuffle=True, random_state=42)
```



Tuned ANNs Model

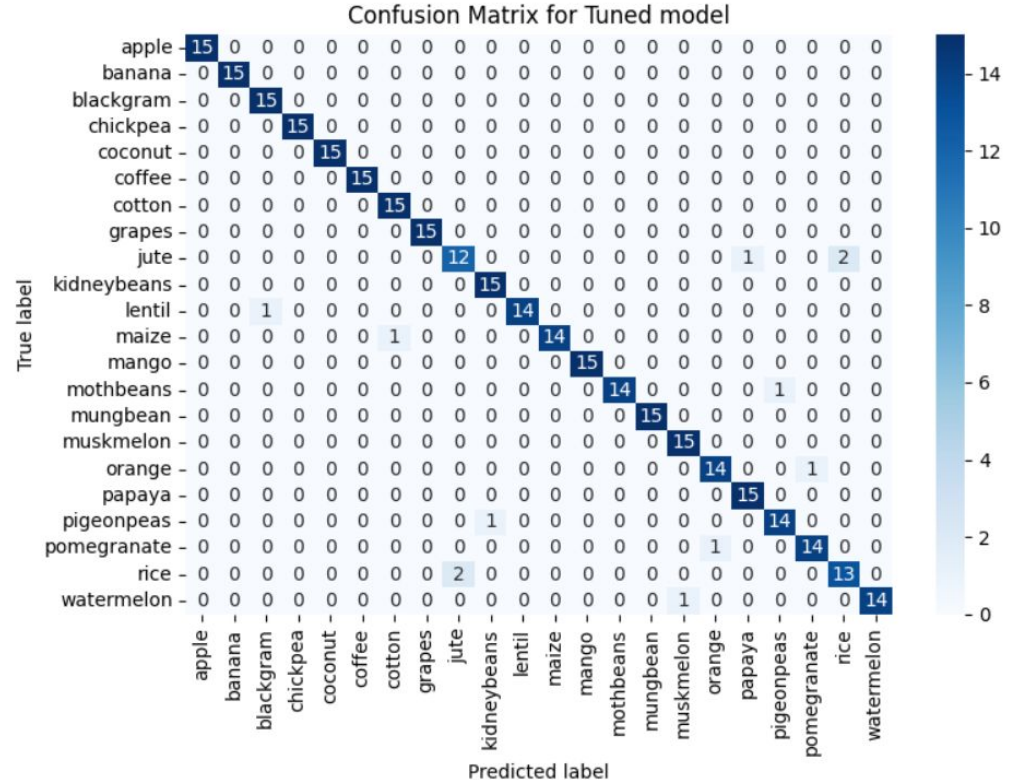


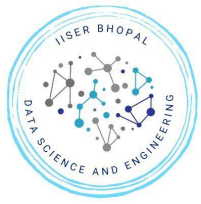
Test Accuracy: 0.9636363636363636

Precision (macro): 0.964150432900433

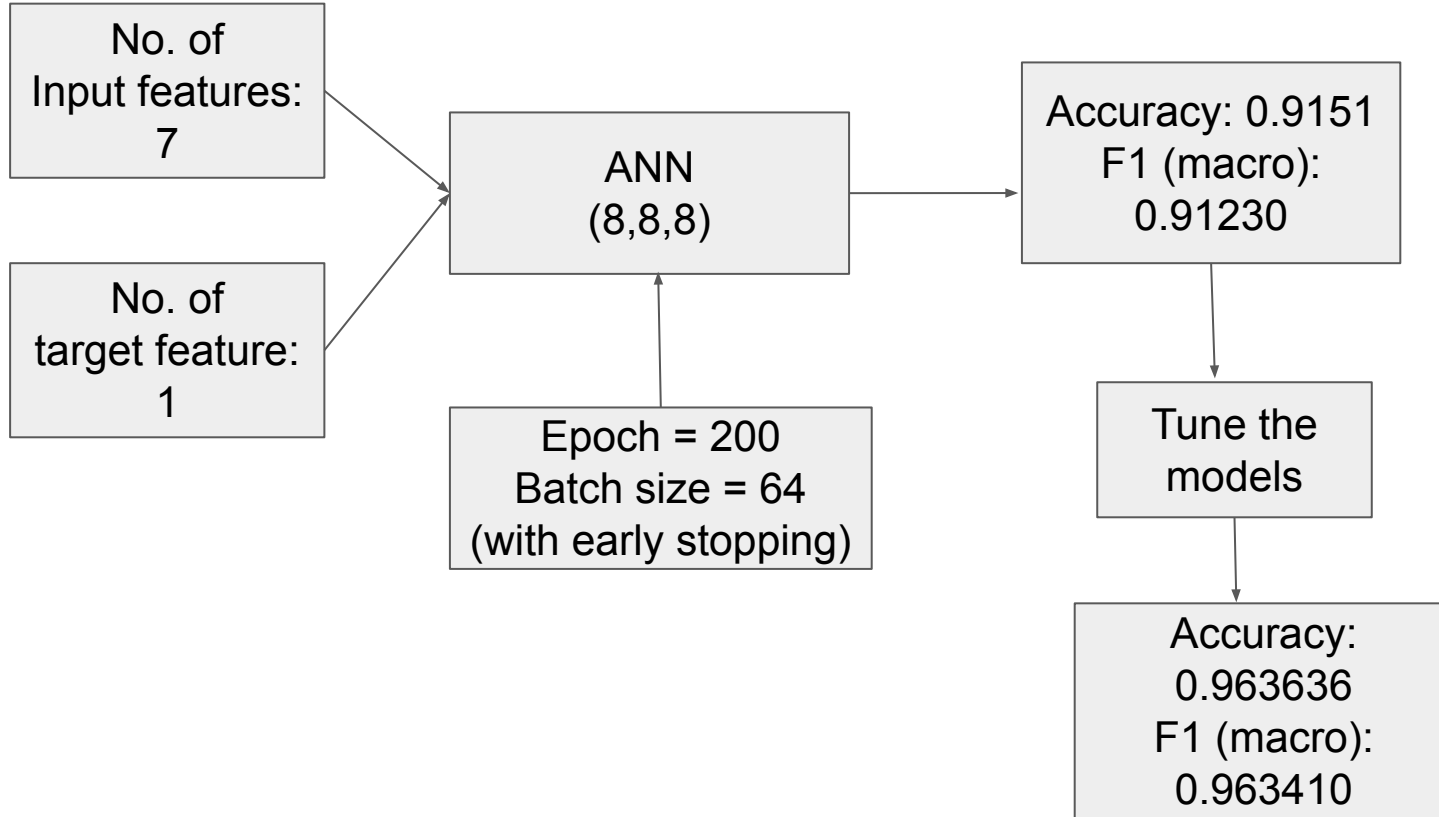
Recall (macro): 0.9636363636363636

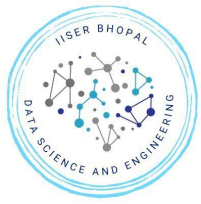
Test F1 (macro): 0.9634105234772644





Pipeline of ANNs





Comparison

Random Forest

Accuracy:

0.995454

F1 (macro):

0.995451

ANNs

Accuracy:

0.963636

F1 (macro):

0.963410

Best model performance: Random Forest

A soft, pink watercolor splash or ink blot is centered on a white background. The splash has irregular, feathered edges and contains subtle variations in pink tones, with some darker areas and lighter, almost white, speckles scattered throughout, giving it a delicate, artistic feel.

Thank
You