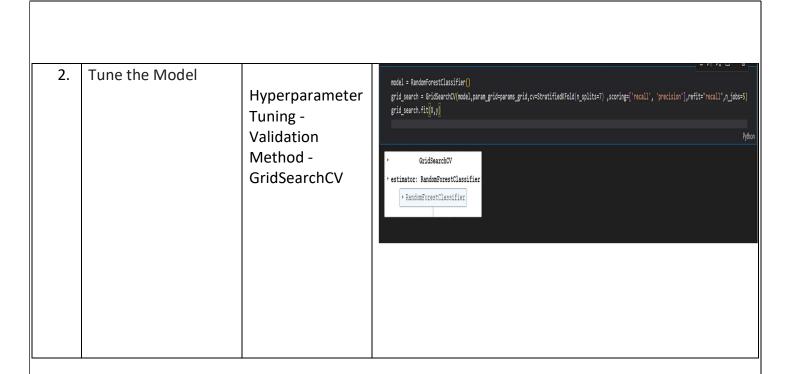
# **Project Development Phase Model Performance Test**

Date	14 November 2023		
Team ID	Team-592631		
Project Name	Detection of Smoke using Machine		
	Learning		
Maximum Marks	10 Marks		

## **Model Performance Testing:**

S.N o.	Parameter	Values	Screenshot			
1.	Model Summary	Classification Model: Accuray Score,F1- score,precision,r ecall & Classification Report -	<pre>: summary = {     "model":[],     "precision":[],     "recall":[],     "f1_score":[],     "accuracy":[] } for i in eval.keys():     summary["model"].append(i)     summary["precision"].append(np.mean(eval[i]["recall"]))     summary["recall"].append(np.mean(eval[i]["accuracy"]))     summary["accuracy"].append(np.mean(eval[i]["f1"]))     summary["f1_score"].append(np.mean(eval[i]["f1"])) pd.DataFrame(summary).style.background_gradient(cmap="Blues")  :     model precision    recall f1_score    accuracy      log    0.907578    0.951494    0.929015    0.896088  1    random    0.999888    0.999978    0.999933    0.999904  2    svm    0.975141    0.989432    0.982234    0.974421  3    KNN    0.998794    0.999263    0.999028    0.998611  4    ada    0.999509    0.999777    0.999643    0.999489  5    gradient    0.999732    0.999866    0.999799    0.999713</pre>			



#### **Model Summary:**

```
summary = {
    "model":[],
    "precision":[],
    "recall":[],
    "f1_score":[],
    "accuracy":[]
}
for i in eval.keys():
    summary["model"].append(i)
    summary["precision"].append(np.mean(eval[i]["precision"]))
    summary["recall"].append(np.mean(eval[i]["recall"]))
    summary["accuracy"].append(np.mean(eval[i]["accuracy"]))
    summary["f1_score"].append(np.mean(eval[i]["f1"]))
pd.DataFrame(summary).style.background_gradient(cmap="Blues")
```

model precision recall f1\_score accuracy 0 0.907578 0.951494 0.929015 0.896088 log 0.999888 0.999978 0.999933 0.999904 1 random 2 0.975141 0.989432 0.982234 0.974421 svm 0.998794 0.999263 0.999028 0.998611 3 KNN 0.999509 0.999777 0.999643 0.999489 0.999732 0.999866 0.999799 0.999713 5 gradient

### Tune the Model: Hyper Paramater Tuning - GridSearchCV

```
model = RandomForestClassifier()
grid_search = GridSearchCV(model,param_grid=params_grid,cv=StratifiedKFold(n_splits=7) ,scoring=['recall', 'precision'],refit="recall",n_jobs=5)
grid_search.fit(X,y)

Python

GridSearchCV

estimator: RandomForestClassifier

RandomForestClassifier
```

#### Using Stratified K-Fold CV we brought the accuracy of the model to 1.00

```
new_skf = StratifiedKFold(n_splits=7,shuffle=True,random_state=42)
new_index = new_skf.split(X,y)
j = 1
for train,test in new_index:
    grid_search.best_estimator_.fit(X[train],y[train])
    print("Fold",j)
    print(classification_report(y[test],grid_search.best_estimator_.predict(X[test])))
    j += 1
Python
```

	precision	песатт	TI-Score	support	
9	1.00	1.00	1.00	2554	
1	1.00	1.00	1.00	6394	
accuracy			1.00	8948	
macro avg	1.00	1.00	1.00	8948	
weighted avg	1.00	1.00	1.00	8948	
3 3					
Fold 2					
	precision	recall	f1-score	support	
0	1.00	1.00	1.00	2554	
1	1.00	1.00	1.00	6393	
accuracy			1.00	8947	
macro avg	1.00	1.00	1.00	8947	
weighted avg	1.00	1.00	1.00	8947	
Fold 3					
	precision	recall	f1-score	support	
9	1.00	1.00	1.00	2553	
1	1.00	1.00	1.00	6394	
accuracy			1.00	8947	
macro avg	1.00	1.00	1.00	8947	
weighted avg	1.00	1.00	1.00	8947	