

Exploratory Data Analysis

HealthCare-DataScience Project
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LISUM19
30.05.2023

Problem Statement



ABC is a pharmaceutical company and desires to understand the persistency of the drug per physician description. To solve this issue, ABC company reached an analytics company to automate this process of identification.

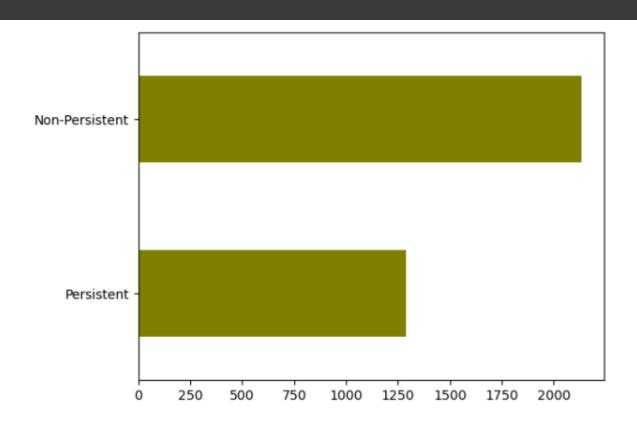
Data Understanding



	Column	Non-Null Count	
	Ptid	3424 non-null	
	Persistency_Flag	3424 non-null	
	Gender	3424 non-null	
	Race	3424 non-null	
	Ethnicity	3424 non-null	
	Region Age Bucket	3424 non-null 3424 non-null	
	Ntm Speciality	3424 non-null	
	Ntm Specialist Flag	3424 non-null	object
	Ntm_Speciality_Bucket	3424 non-null	object
	Gluco_Record_Prior_Ntm	3424 non-null 3424 non-null 3424 non-null	object
11	Gluco_Record_During_Rx	3424 non-null	object
	Dexa_Freq_During_Rx	3424 non-null	int64
	Dexa_During_Rx Frag_Frac_Prior_Ntm		object
	Frag_Frac_During_Rx	3424 non-null	object
	Risk Segment Prior Ntm	3424 non-null	object
17	Tscore_Bucket_Prior_Ntm	3424 non-null	object
	Risk_Segment_During_Rx	3424 non-null	
	Tscore_Bucket_During_Rx	3424 non-null	
	Change_T_Score Change_Risk_Segment	3424 non-null 3424 non-null	
	Adherent Flag	3424 non-null	
	Idn Indicator	3424 non-null	
	Injectable_Experience_During_Rx	3424 non-null	
25	Comorb Encounter For Screening For Malignant Neoplasms	3424 non-null	
26	Comorb_Encounter_For_Immunization		object
	Comorb_Encntr_For_General_Exam_W_O_Complaint,_Susp_Or_Reprtd_Dx		object
	Comorb_Vitamin_D_Deficiency Comorb_Other_Joint_Disorder_Not_Elsewhere_Classified		object
38	Comorb_Enchtr_For_Oth_Sp_Exam_W_O_Complaint_Suspected_Or_Reprtd_Dx	3424 pop-pull	object
31	Comorb Long Term Current Drug Therapy	3424 non-null	object
32	Comorb_Dorsalgia	3424 non-null	object
33	Comorb_Personal_History_Of_Other_Diseases_And_Conditions	3424 non-null	
	Comorb_Other_Disorders_Of_Bone_Density_And_Structure	3424 non-null	
	Comorb_Disorders_of_lipoprotein_metabolism_and_other_lipidemias Comorb Osteoporosis without current pathological fracture	3424 non-null 3424 non-null	
	Comorb_Personal_history_of_malignant_neoplasm	3424 non-null	
	Comorb Gastro esophageal reflux disease	3424 non-null	
	Concom_Cholesterol_And_Triglyceride_Regulating_Preparations	3424 non-null	object
	Concom_Narcotics	3424 non-null	
	Concom_Systemic_Corticosteroids_Plain	3424 non-null	
	Concom_Anti_Depressants_And_Mood_Stabilisers Concom_Fluoroguinolones	3424 non-null	object
	Concom_Cephalosporins	3424 non-null 3424 non-null 3424 non-null	object
	Concom Macrolides And Similar Types	3424 non-null	object
46	Concom Broad Spectrum Penicillins	3424 non-null	object
	Concom_Anaesthetics_General		object
	Concom_Viral_Vaccines		object
	Risk_Type_1_Insulin_Dependent_Diabetes	3424 non-null	
	Risk_Osteogenesis_Imperfecta Risk_Rheumatoid_Arthritis	3424 non-null 3424 non-null	
	Risk_Untreated_Chronic_Hyperthyroidism	3424 non-null	
	Risk_Untreated_Chronic_Hypogonadism	3424 non-null	
	Risk_Untreated_Early_Menopause	3424 non-null	
	Risk_Patient_Parent_Fractured_Their_Hip	3424 non-null	
	Risk_Smoking_Tobacco	3424 non-null	
	Risk_Chronic_Malnutrition_Or_Malabsorption Risk_Chronic_Liver_Disease	3424 non-null 3424 non-null	object
59	Risk_Family_History_Of_Osteoporosis		object
68	Risk_Low_Calcium_Intake	3424 non-null	
61	Risk_Vitamin_D_Insufficiency		object
62	Risk_Poor_Health_Frailty		object
	Risk_Excessive_Thinness		object
	Risk_Hysterectomy_Oophorectomy		object
66	Risk_Estrogen_Deficiency Risk_Immobilization	3424 non-null 3424 non-null	object
67	Risk Recurring Falls	3424 non-null	object

- Dataset has 68 columns.
- Our dataset has only one numerical column which is dexa frequency during rx.
- -66 Columns are categorical values most of them being simply «Y» or «N»





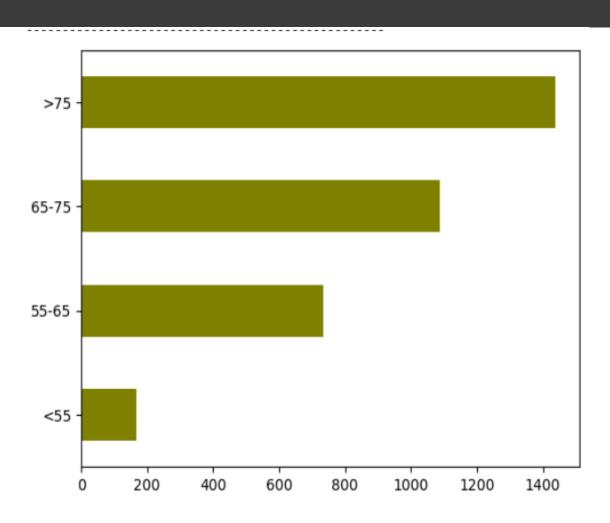
Target Feature, Persistent Flag

Uneven Data;

Non Persistent: 2135

Persistent: 1289





Age Bucket;

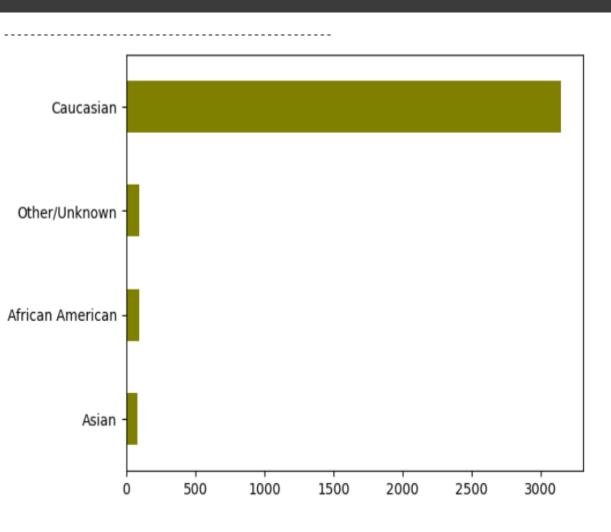
>75:1439

65-75: 1086

55-65: 733

<55: 166





Race

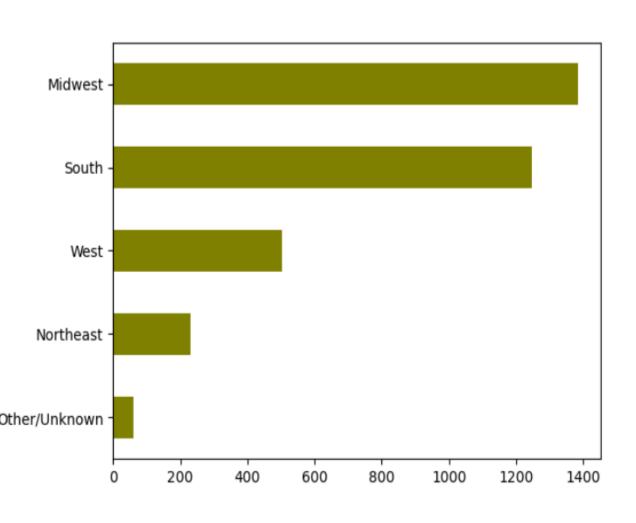
Caucasian: 3148,

Other/Unknown: 97,

African American: 95,

Asian: 84,





Region

Midwest: 1383

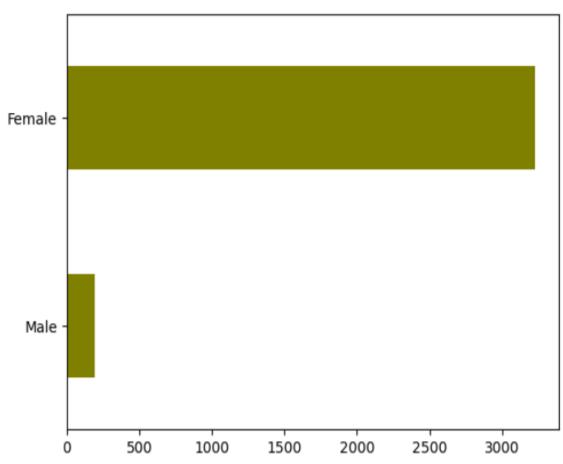
South: 1247

West: 502

Other/Unkown:60







Gender

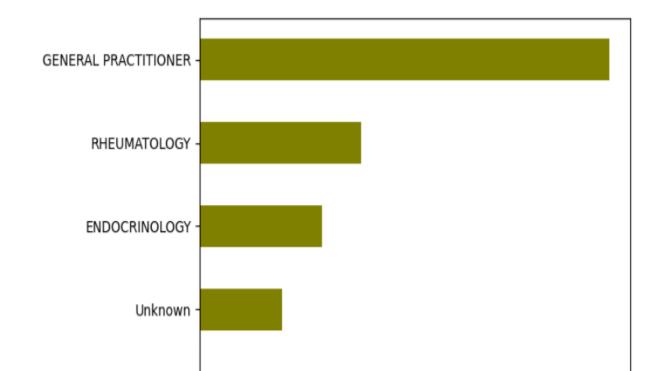
Uneven Dataset

Female Dominating Dataset

Female: 3424

Male: 194





400

600

200

800

1000

1200

1400

1600

ONCOLOGY -

Ntm Speciality

36 Unique Values

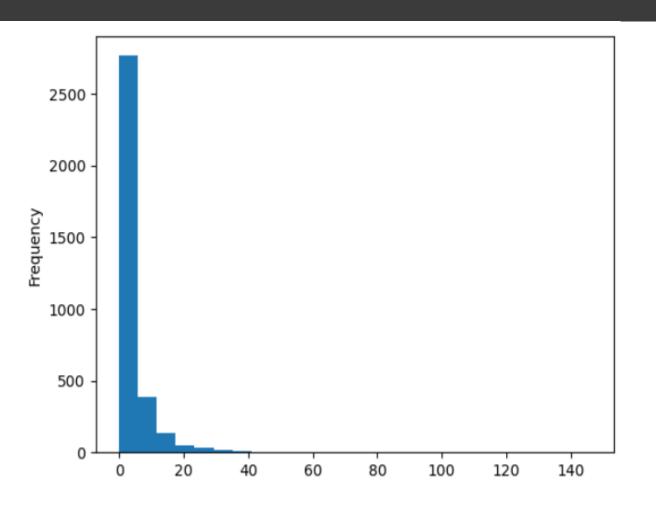
With Leading;

General Practitioner: 1535

Rheumatology: 604

Endocrinology: 458



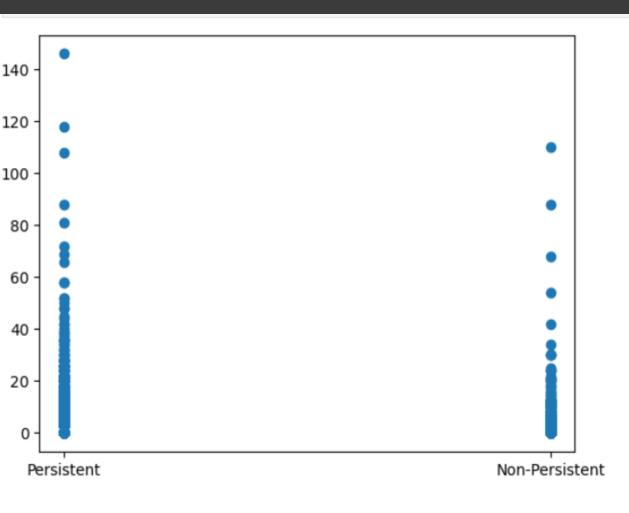


Dexa Frequancy During Rx Only Numerical Column In The Dataset

0	Dexa_Freq_During_Rx		
	count	3424.000000	
	mean	3.016063	
	std	8.136545	
	min	0.000000	
	25%	0.000000	
	50%	0.000000	
	75%	3.000000	
	max	146.000000	

Finding Outliers.





Distribution of Dexa Frequancy During Rx with target column persistancy flag. Clearly outliers are visible. This outliers are dropped.

Unknown Values In the Dataset.



Race	2.832944
Ethnicity	2.657710
Region	1.752336
Ntm_Speciality	9.053738
Risk_Segment_During_Rx	43.720794
Tscore_Bucket_During_Rx	43.720794
Change_T_Score	43.720794
Change Risk Segment	65.099299

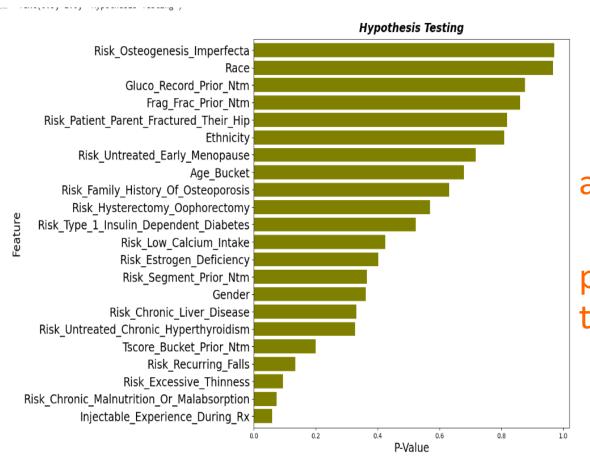
Percentage of Unknown Values In the columns.

If the percentage is higher than 40, column is dropped.

Else, the unknown data is filled with the most frequent value

Chi2 Analysis





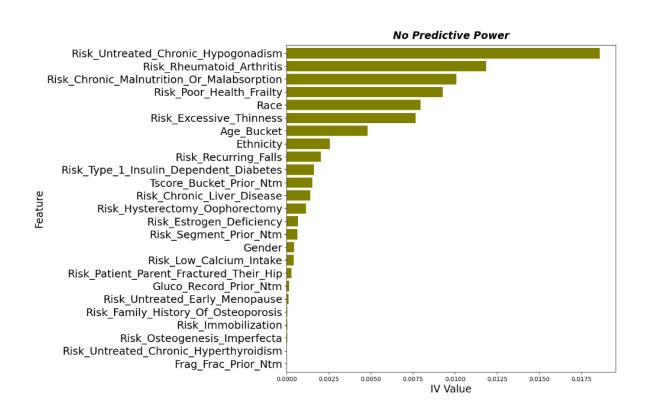
Reduction of Columns

At the left columns with low assosiation are shown

If P value is higher it has a higher probability of being not assosiated with the target variable

Information Value





IV technique for feature selection Columns with no predictive power.

Suggested ML Models:



Logistic Regression: WOE/IV

DecisionTree/RandomForest

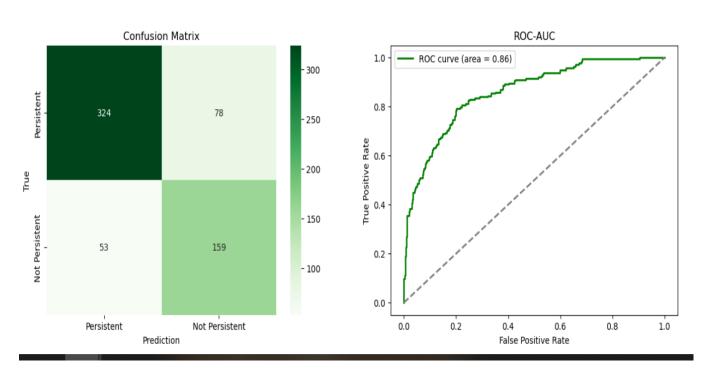
Gradient Boosting Machines (GBM) /

XGBoost / LightGBM

Support Vector Machines (SVM):

Neural Networks

Model 1 Logistic Regression (Base Model)



Accuracy: 0.7866449511400652

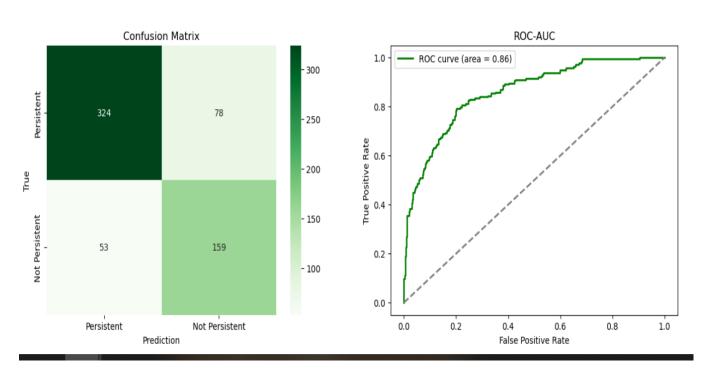
Data Glacier

Precision: 0.6708860759493671

Recall: 0.75

F1 Score: 0.7082405345211581

Model 1 Logistic Regression (Base Model)



Accuracy: 0.7866449511400652

Data Glacier

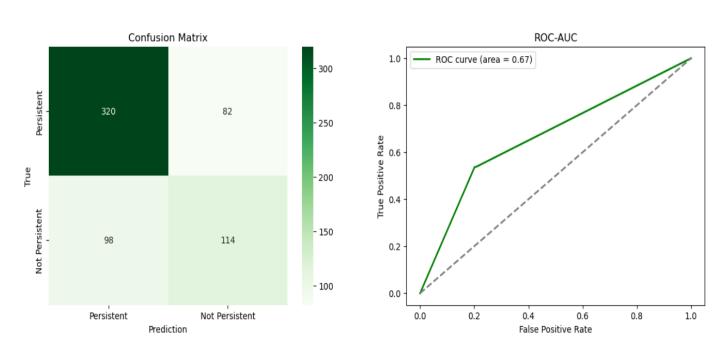
Precision: 0.6708860759493671

Recall: 0.75

F1 Score: 0.7082405345211581

Model 2 Decision Tree.





Accuracy: 0.7068403908794788

Precision: 0.5816326530612245

Recall: 0.5377358490566038

F1 Score: 0.5588235294117646

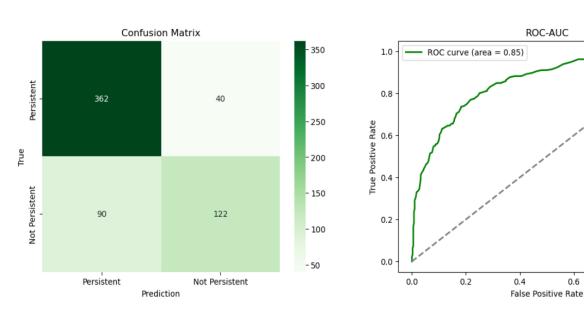
Model 3 Random Forest Classifier

0.6

0.8

1.0





Accuracy: 0.7882736156351792

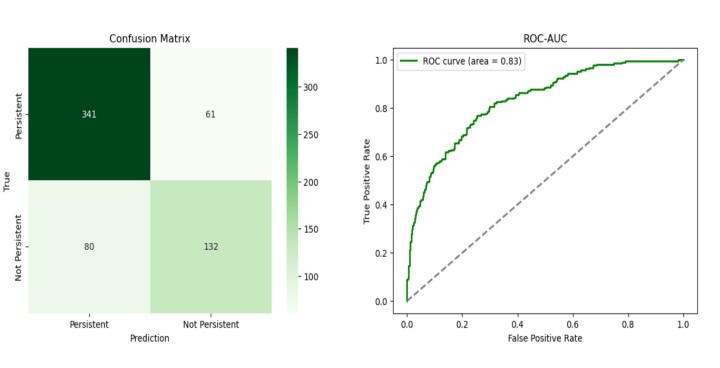
Precision: 0.7530864197530864

Recall: 0.5754716981132075

F1 Score: 0.6524064171122994

Model 4 Gradient Boosting





Accuracy: 0.7703583061889251

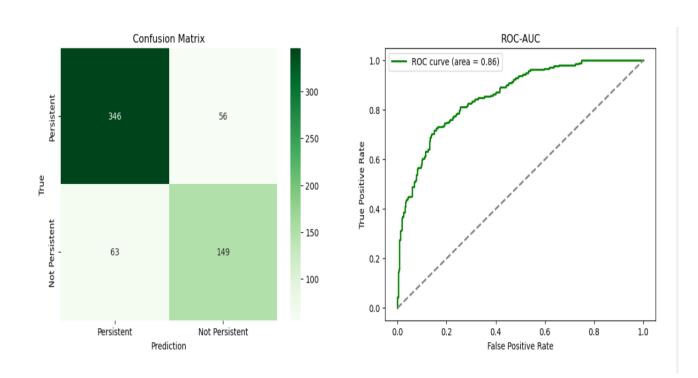
Precision: 0.6839378238341969

Recall: 0.6226415094339622

F1 Score: 0.6518518518519

Model 5 Support Vector Machines





Accuracy: 0.8061889250814332

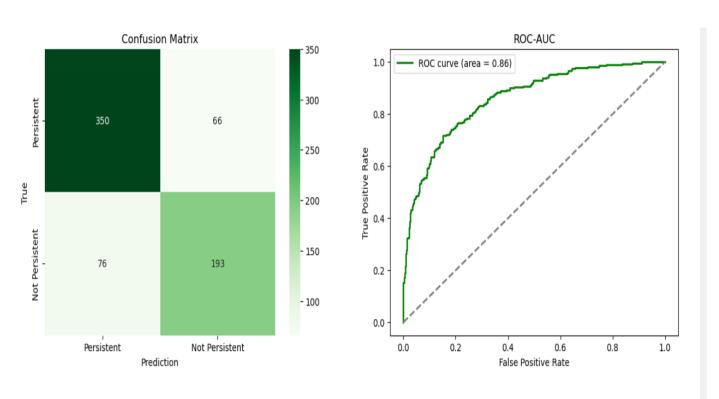
Precision: 0.7268292682926829

Recall: 0.7028301886792453

F1 Score: 0.7146282973621103

Model 6 Logistic Regression WoE





Accuracy: 0.7927007299270074

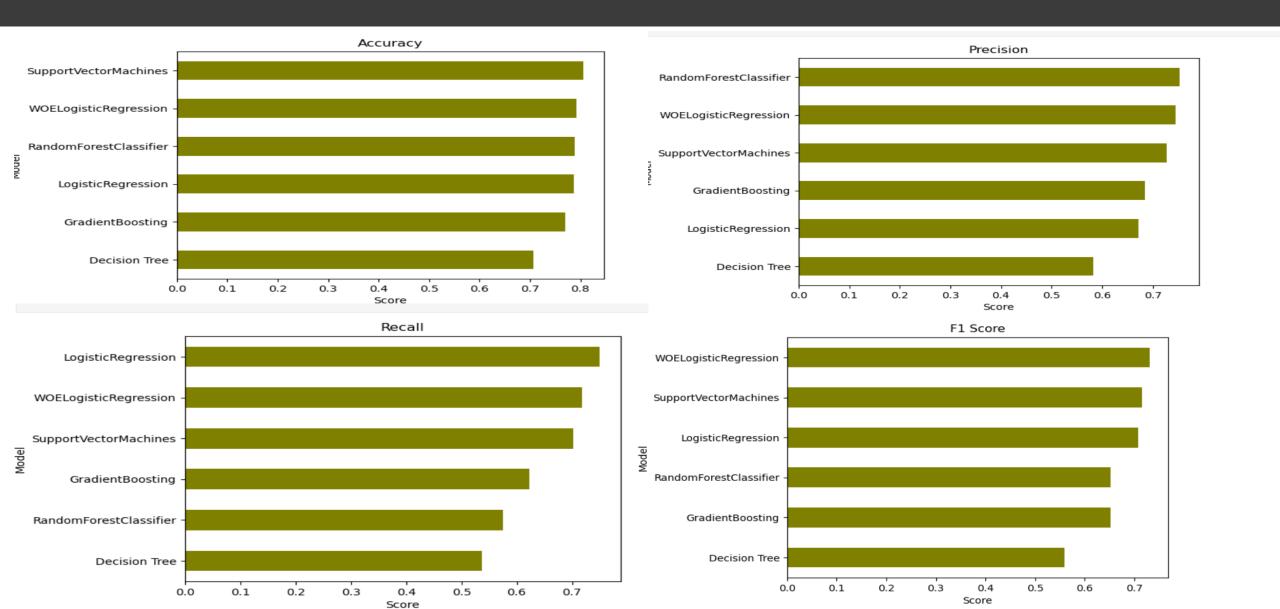
Precision: 0.7451737451737451

Recall: 0.7174721189591078

F1 Score: 0.731060606060606

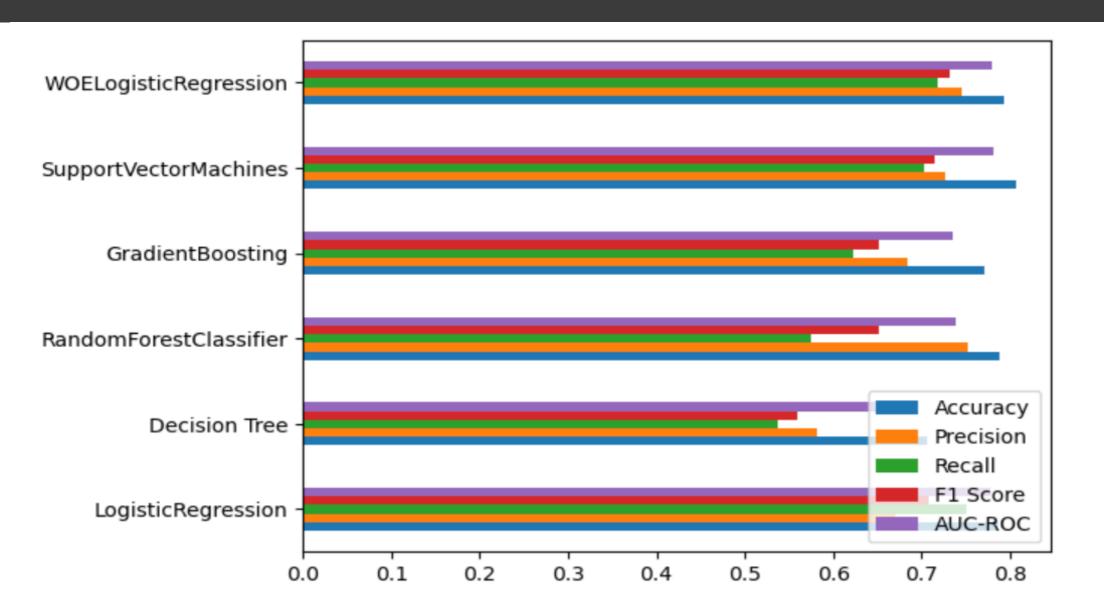
Final Metrics





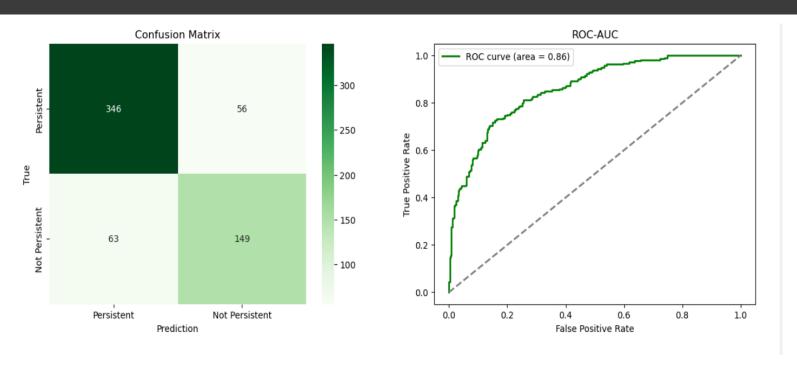
Final Metrics





Selected Model: SVM





Accuracy: 0.8061889250814332

Precision: 0.7268292682926829

Recall: 0.7028301886792453

F1 Score: 0.7146282973621103

^{*}Since It's high accuracy, high F1 score, and ease of implementation SVM is chosen as model to deploy.*

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Thank You

