



# Imarticus Learning

## Project on: Credit Risk Analysis

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## Credit Risk Analysis

What is Credit Risk Analysis?

Why Credit Risk Analysis is used?



Where Credit Risk Analysis is used?

## Data Analysis

- Data Analysis refers to extracting knowledge and insight from large amount of Data.

### Process of extracting Gold :



Searching Gold Mine



Removing Soil and Rock  
Or  
Extracting Gold



Required Gold

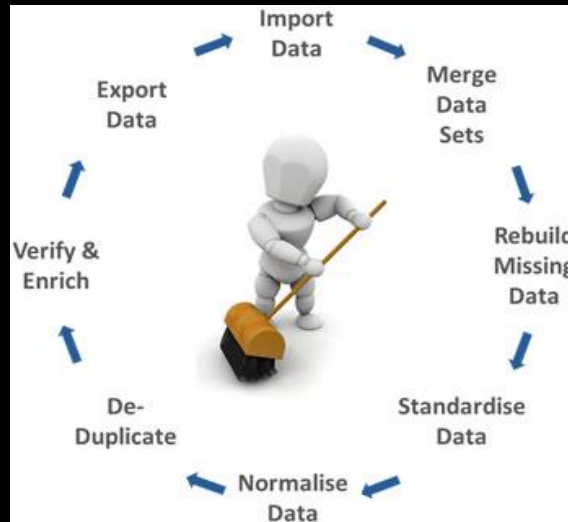
# Credit Risk Analysis

## Our Data Analysis

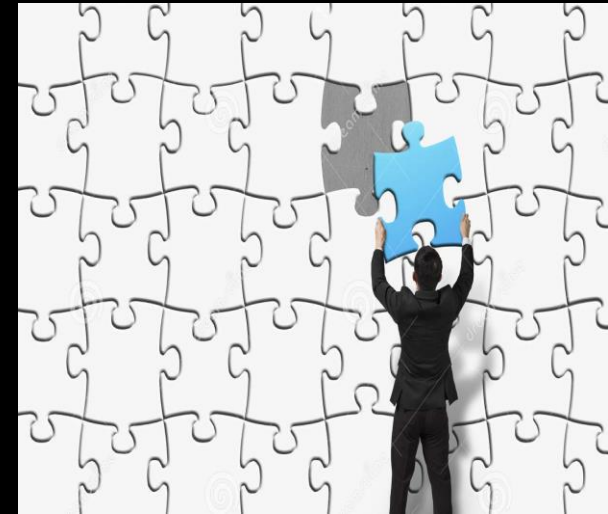
### Process of Extracting Data for Credit Risk Analyse:



Data Taken from  
XYZ Co-operation



Data Cleaning and  
Feature Selection



Required Data for  
Analysis

# Credit Risk Analysis

## OBJECTIVE

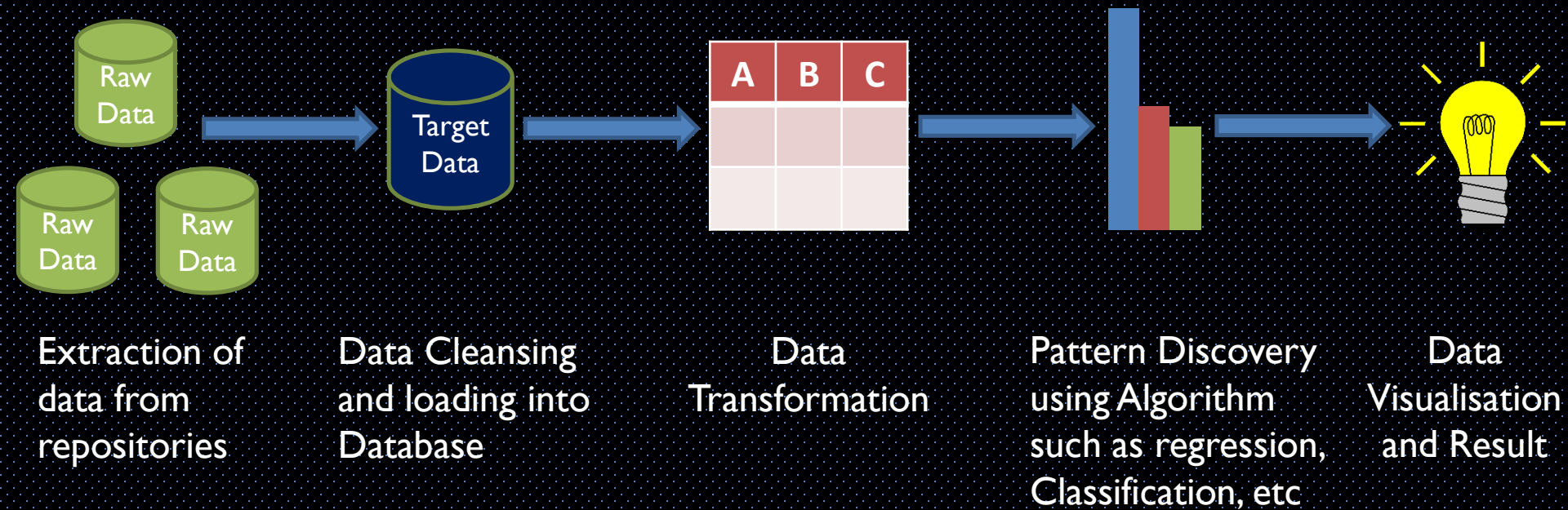


Loan Issuer

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# Credit Risk Analysis

## Steps used for Credit Risk Analysis



# Extraction of Data from XYZ Co-operation

## Data Contains

1. All the details of the people who applied for loan between June 2007 and December 2015
2. The data contains the indicator of default, payment information, Credit history, etc.
3. Columns – 73  
Rows - 855696

## Pre - processing of Data

### Data Cleaning

#### Pre Processing the Data

Find Missing Value and Calculating Missing Value Percentage.

```
In [11]: Count_Null=Credit_DS.isnull().sum()  
print(Count_Null)
```

...

```
In [13]: Per_Null =round(Credit_DS.isnull().sum()/len(Credit_DS) * 100,2)  
print(Per_Null)
```

...

```
In [16]: Missing_Val.to_csv(r'D:\Desktop\MissingValues.csv')
```



# Credit Risk Analysis

## Pre - processing of Data

Column Name	count	% Count	Column Name	count	% Count
annual_inc_joint	855527	99.95	inq_last_12m	842681	98.45
dti_joint	855529	99.95	desc	734157	85.77
verification_status_joint	855527	99.95	mths_since_last_record	724785	84.67
il_util	844360	98.64	mths_since_last_major_derog	642830	75.1
mths_since_rcnt_il	843035	98.49	mths_since_last_delinq	439812	51.38
open_acc_6m	842681	98.45	next_pymnt_d	252971	29.55
open_il_6m	842681	98.45	tot_coll_amt	67313	7.86
open_il_12m	842681	98.45	tot_cur_bal	67313	7.86
open_il_24m	842681	98.45	total_rev_hi_lim	67313	7.86
total_bal_il	842681	98.45	emp_title	49443	5.78
open_rv_12m	842681	98.45	emp_length	43061	5.03
open_rv_24m	842681	98.45	last_pymnt_d	8862	1.04
max_bal_bc	842681	98.45	revol_util	446	0.05
all_util	842681	98.45	last_credit_pull_d	50	0.01
inq_fi	842681	98.45	collections_12_mths_ex_med	56	0.01
total_cu_tl	842681	98.45			

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## Data Transformation

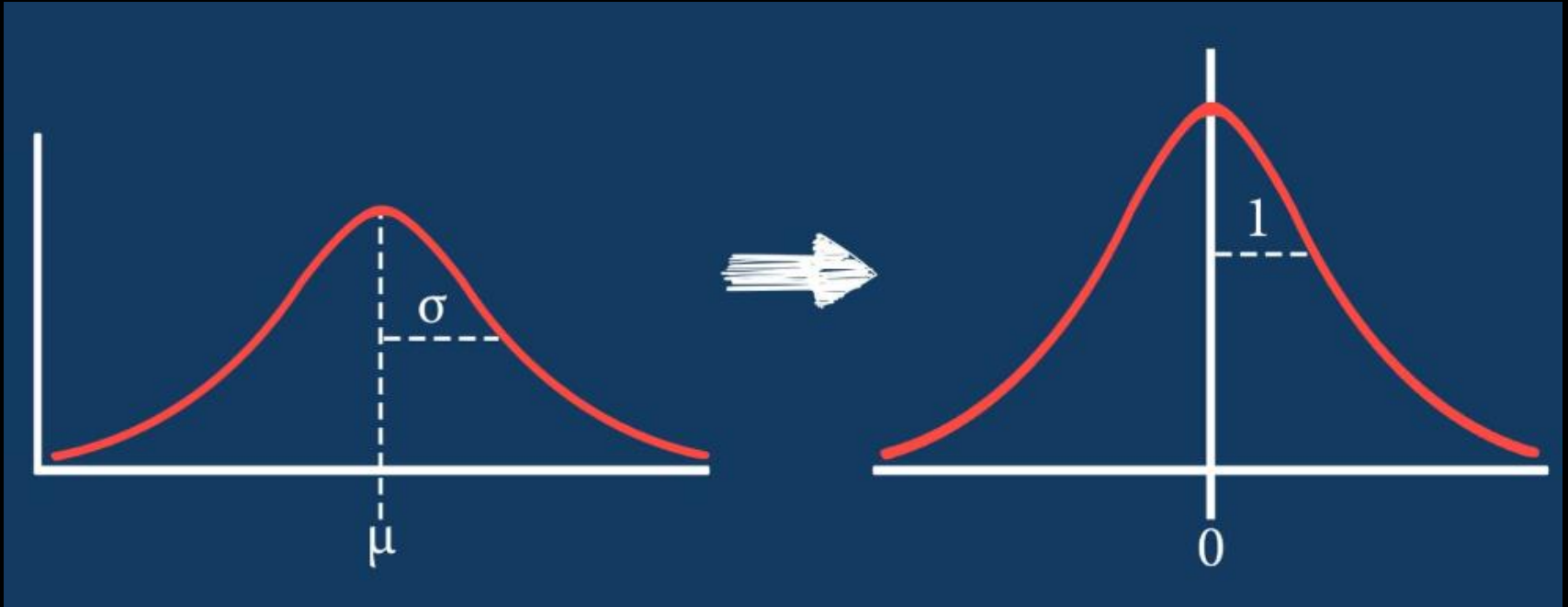
### Converting Categorical data to Numeric

Column Name
Term
Grade
Emp_Length
Home_ownership
Verification_status
Purpose
Initial_List_Status
Application_type
Open_account

**For Example:**

```
In [30]: Credit_DS.grade.value_counts()
grade_final={'A':6,'B':5,'C':4,'D':3,'E':2,'F':1,'G':0}
Credit_DS.grade=[grade_final[item] for item in Credit_DS.grade]
print(Credit_DS.grade)
```

## Data Standardisation



# Credit Risk Analysis

## Data Splitting

**Credit\_DS**

Rows : 855696

Columns : 38

June 2007 to May 2015

June 2015 to Dec 2015

**Training**

Rows : 598978

Columns : 38

**Testing**

Rows : 256991

Columns : 38

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## Models used for Analysing

**Logistic  
Regression**

**Decision  
Tree**

**Adaptive Boost  
Classifier**

**Extra Trees  
Classifier**

# Understanding Model Implementation

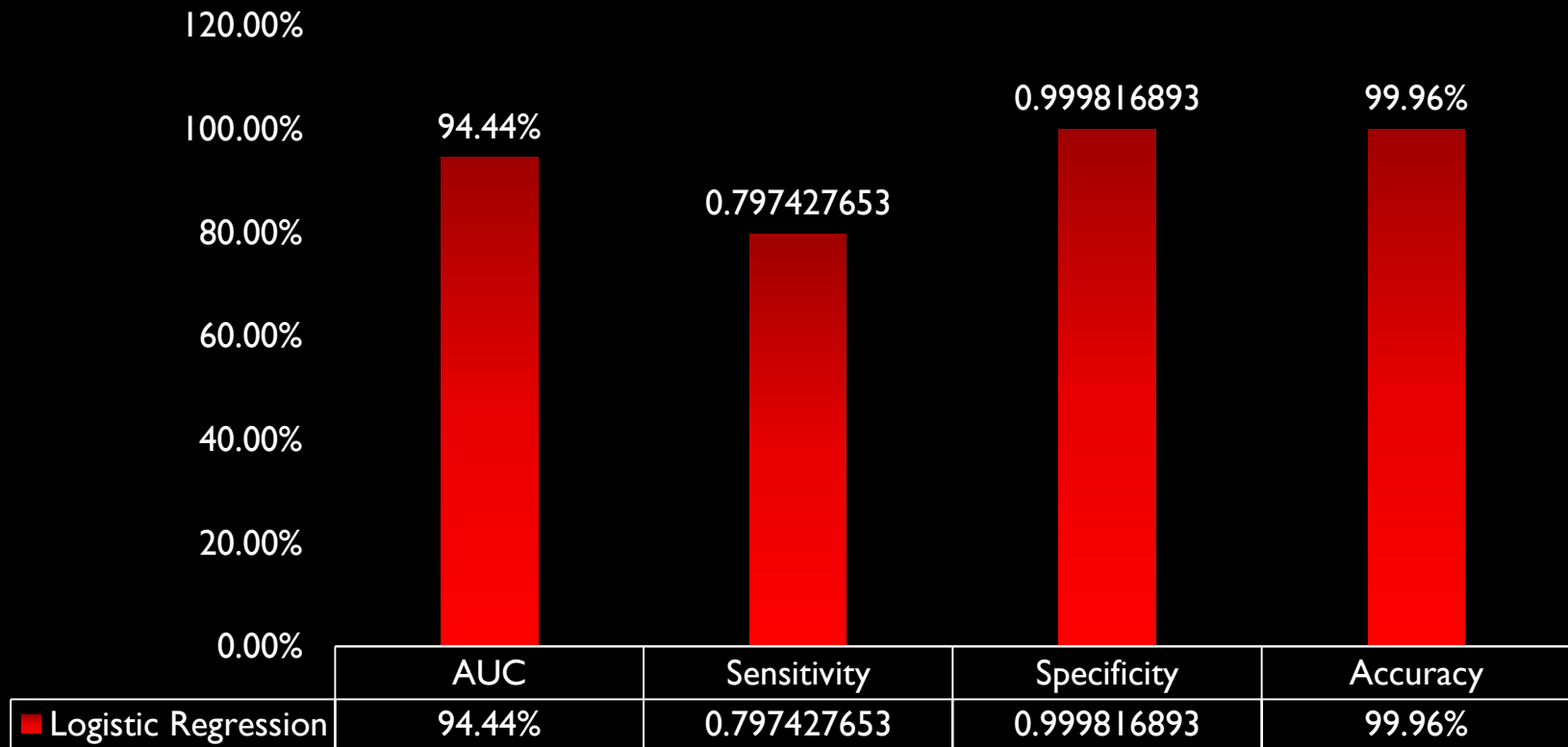
**Logistic  
Regression**



# Credit Risk Analysis

## Model – 1 – Logistic Regression

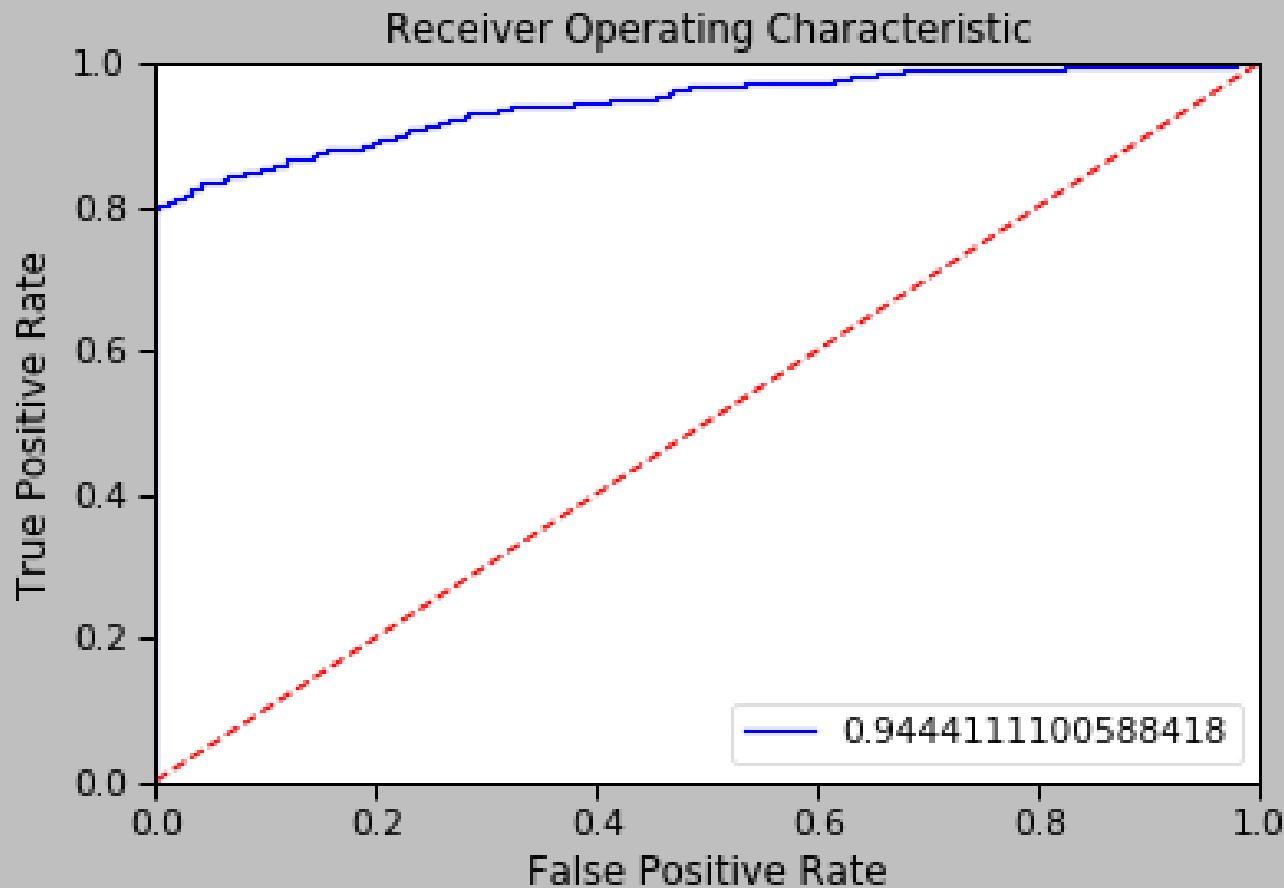
### Logistic Regression Model



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# Credit Risk Analysis

## Receiver Operating Characteristic



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# Understanding Model Implementation



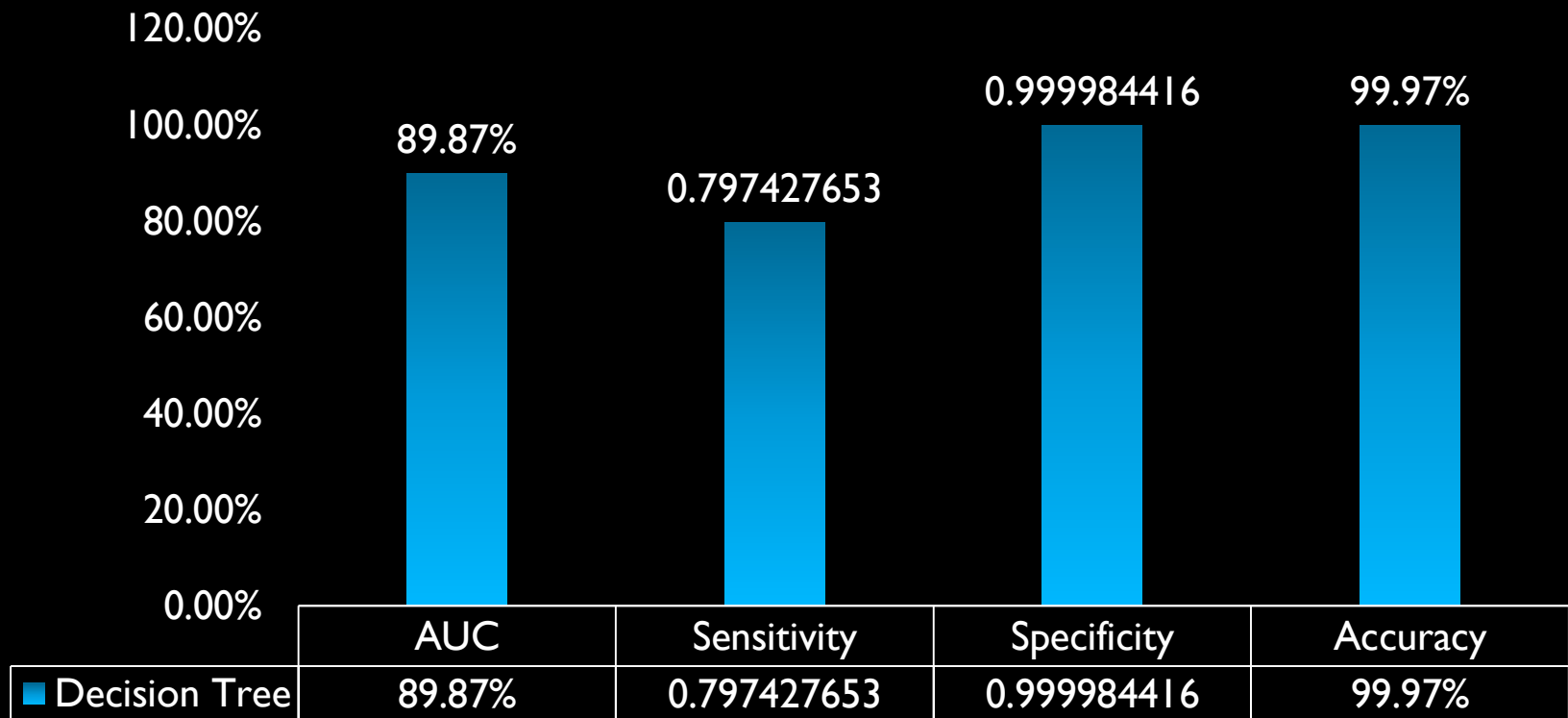
**Logistic  
Regression**

**Decision  
Tree**

# Credit Risk Analysis

## Model – 2 Decision Tree

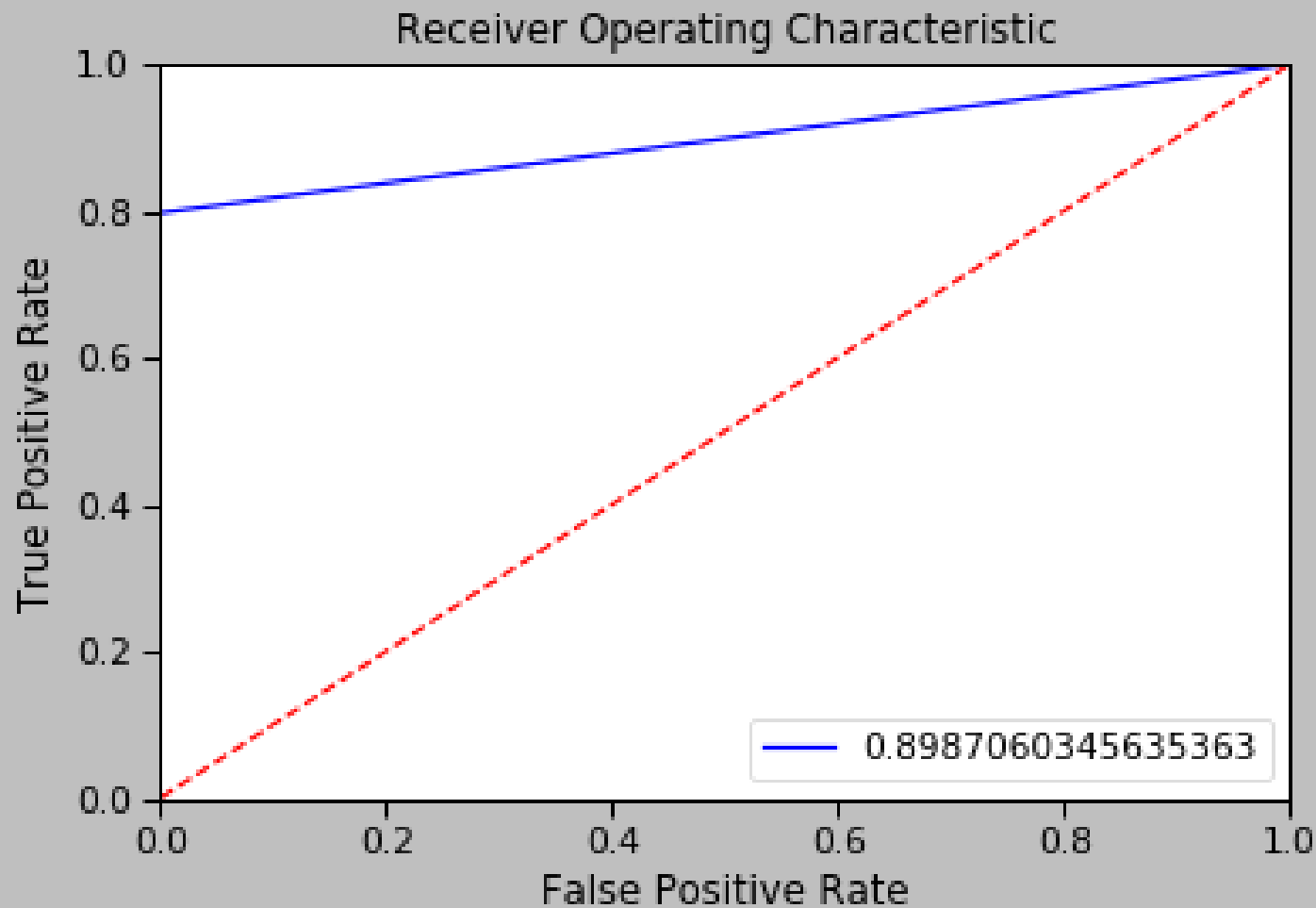
### Decision Tree



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# Credit Risk Analysis

## Receiver Operating Characteristic



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# Understanding Model Implementation

**Logistic  
Regression**

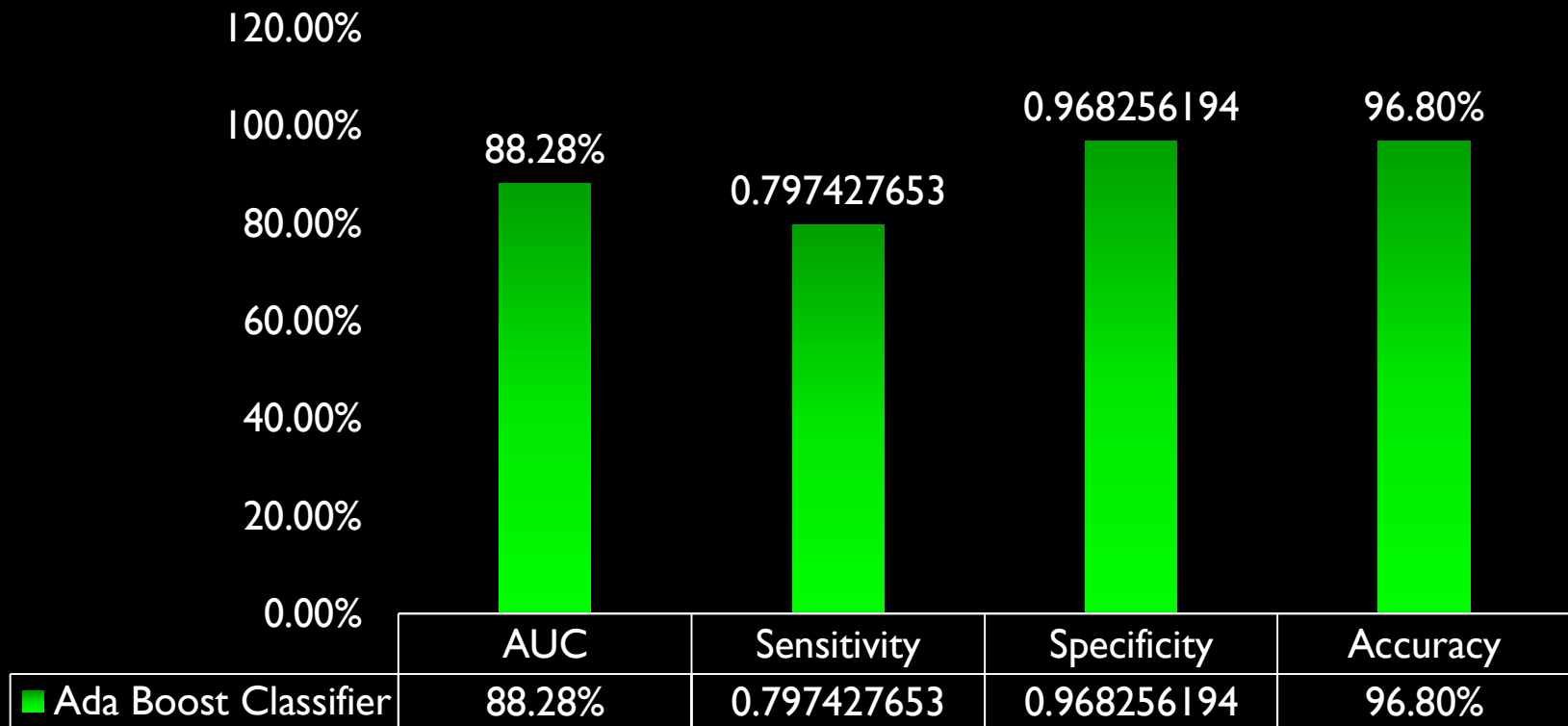
**Decision  
Tree**

**Adaptive Boost  
Classifier**

# Credit Risk Analysis

## Model – 3 Adaptive Boosting Classifier

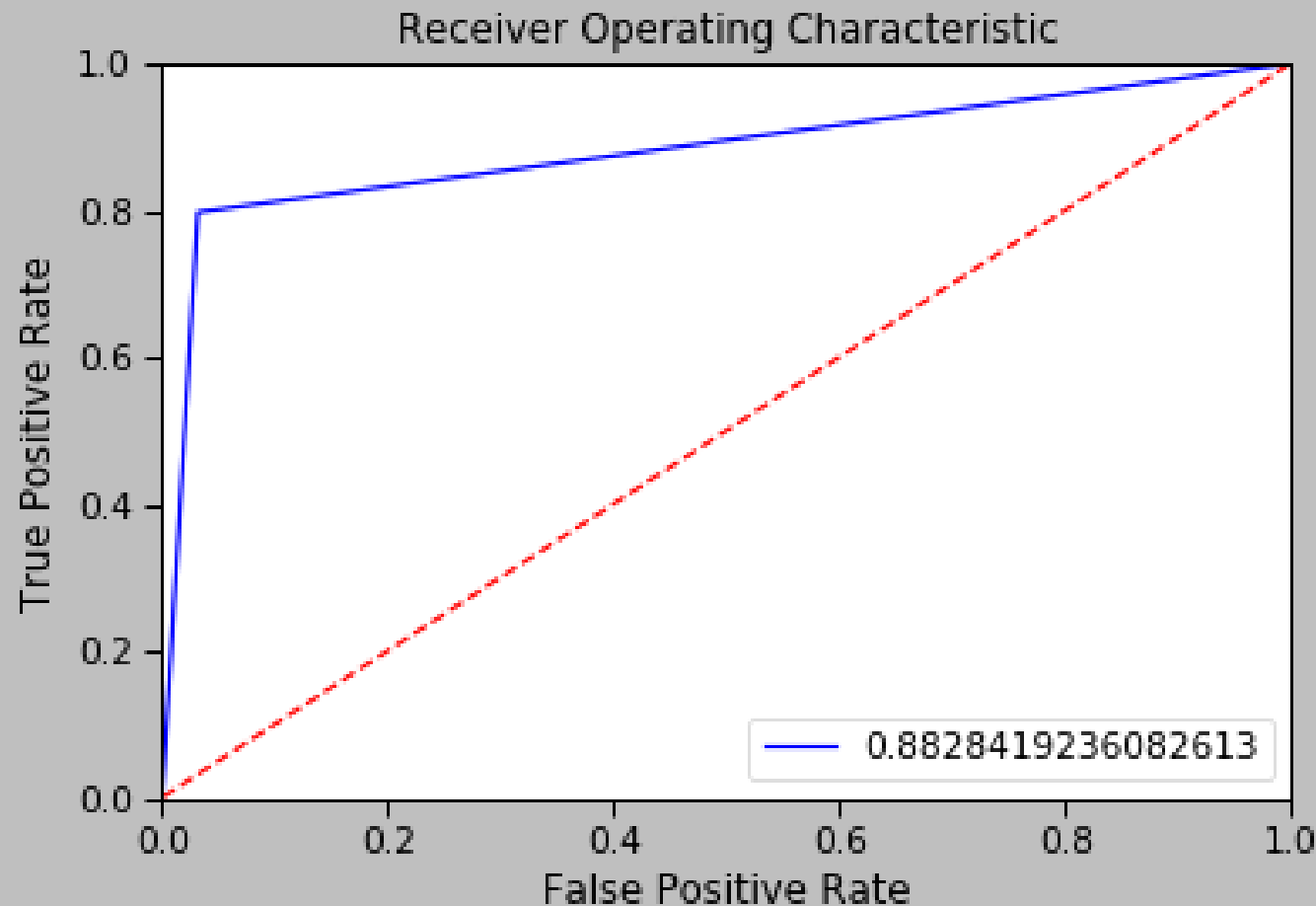
Ada Boost Classifier



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# Credit Risk Analysis

## Receiver Operating Characteristic



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# Understanding Model Implementation

**Logistic  
Regression**

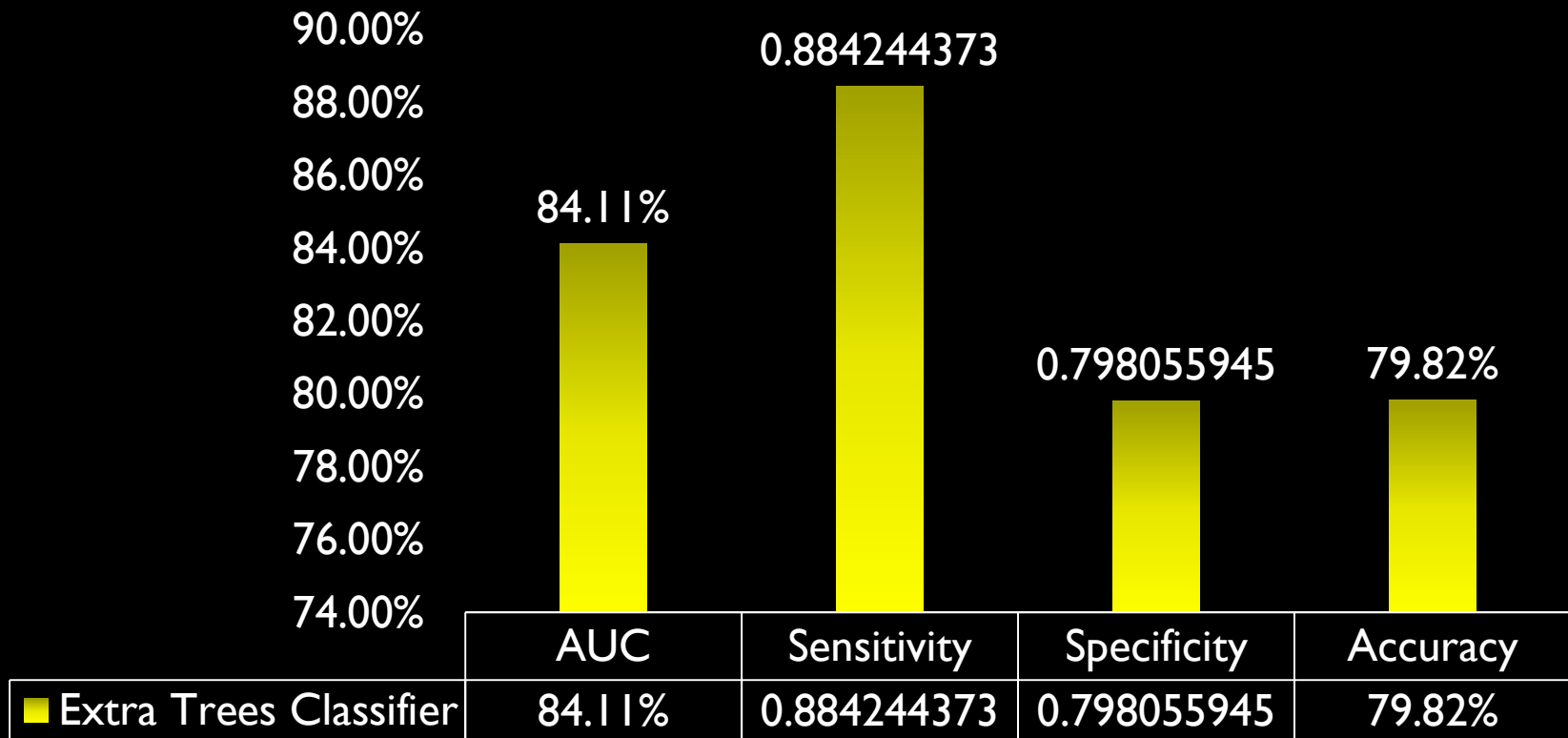
**Decision  
Tree**

**Adaptive Boost  
Classifier**

**Extra Trees  
Classifier**

## Model – 4 – Extra Trees Classifier

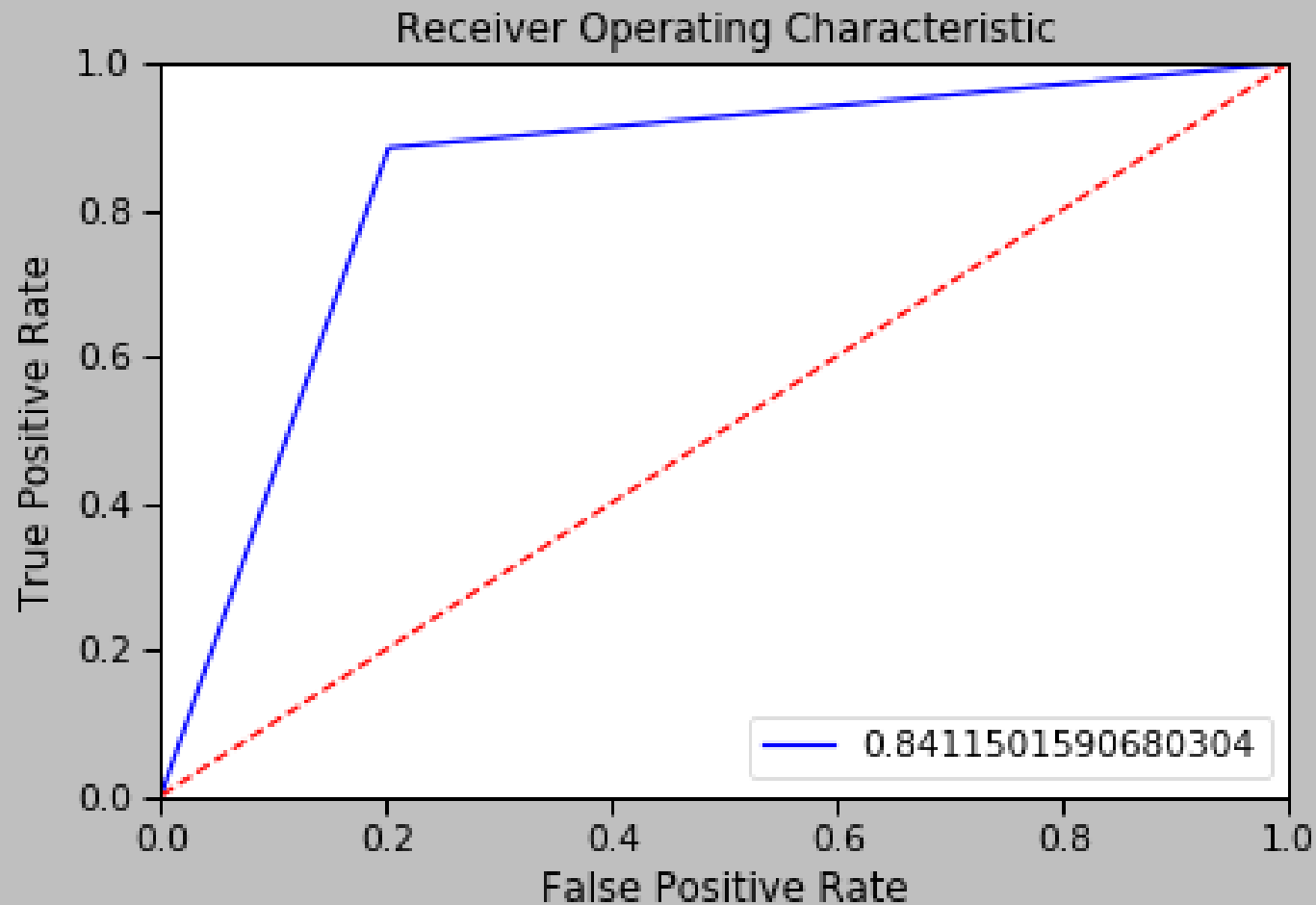
### Extra Trees Classifier





# Credit Risk Analysis

## Receiver Operating Characteristic

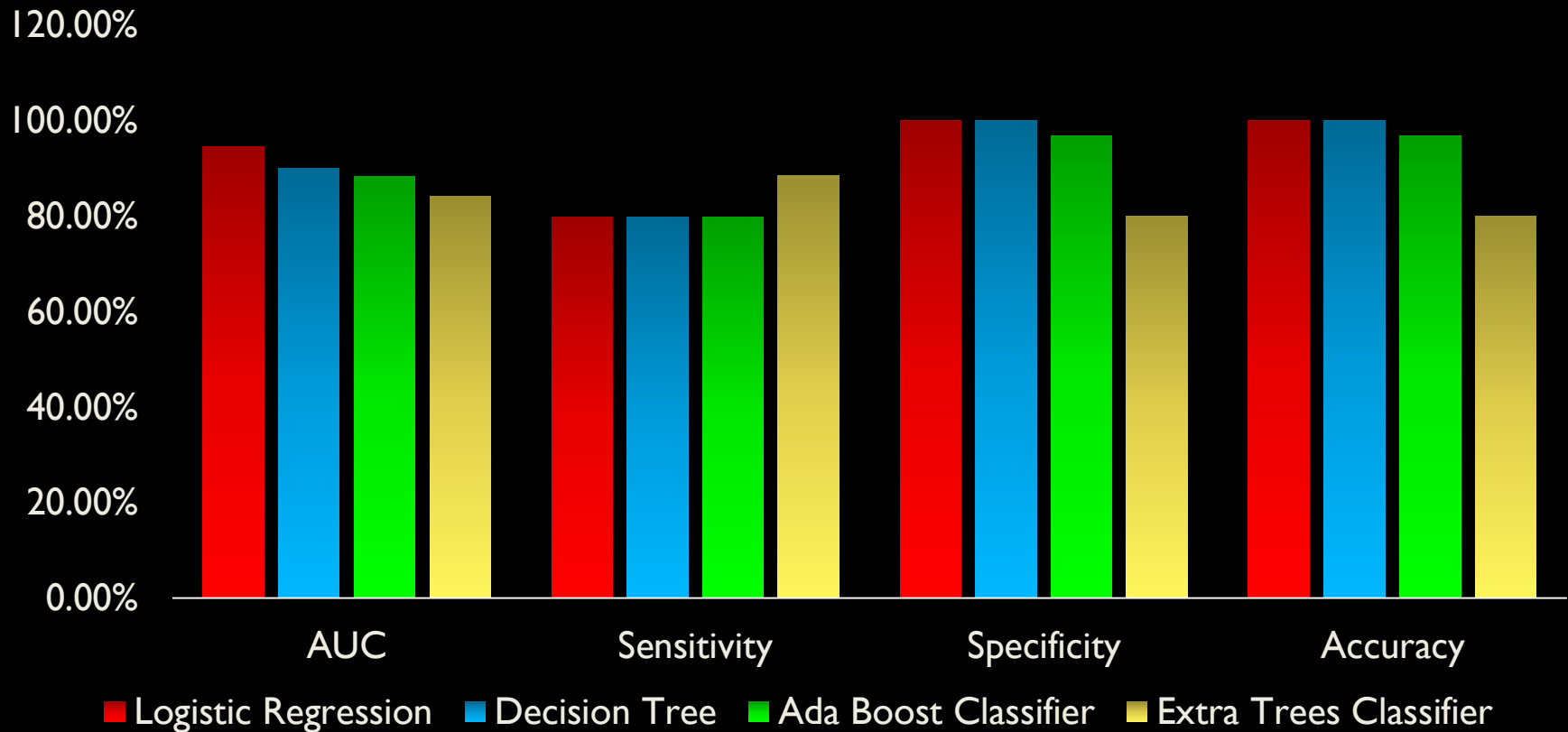


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# Credit Risk Analysis

## Which Model is Performing Better?

### Description of Models



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# Which Model is Performing Better?

**Logistic  
Regression**

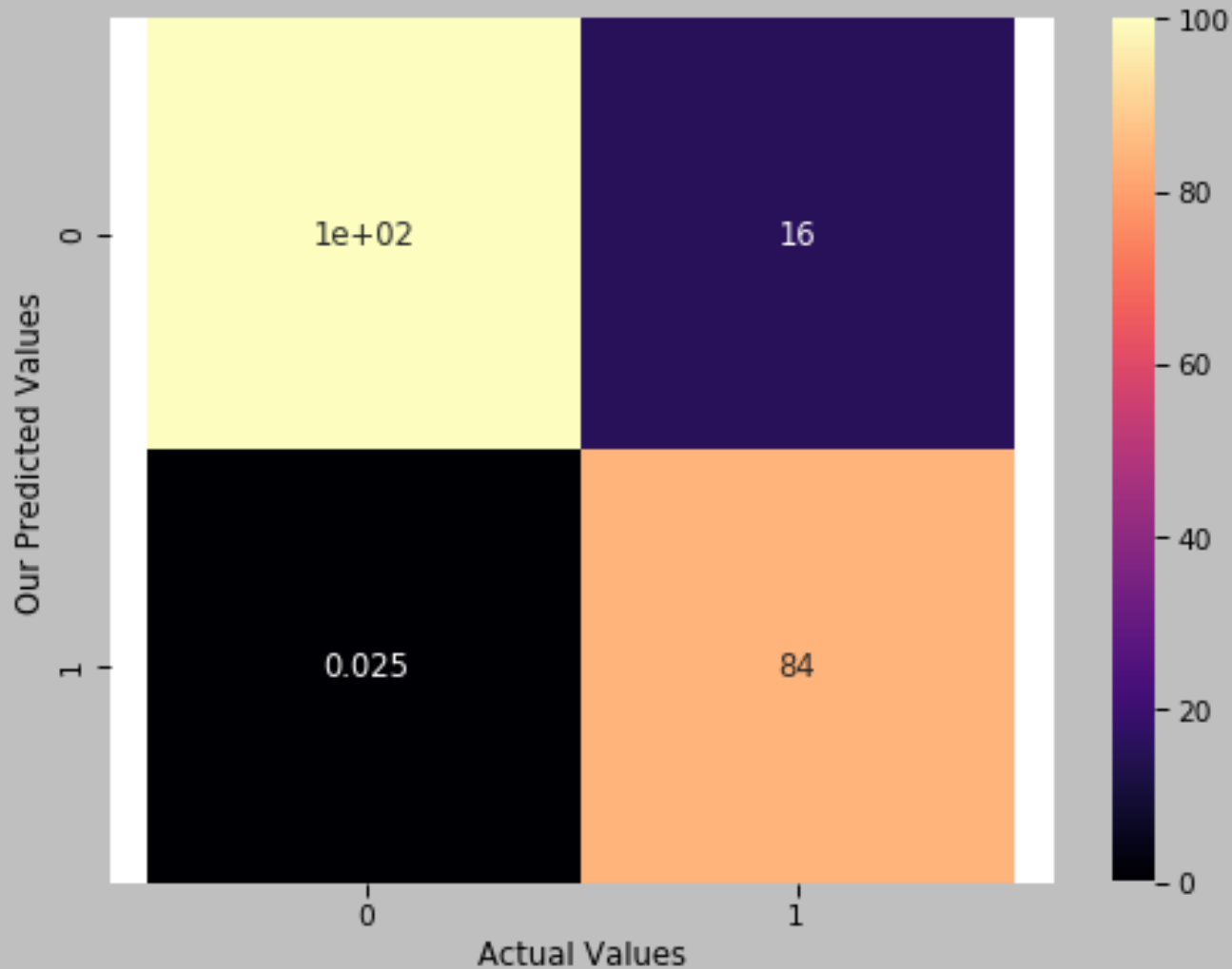
**Decision  
Tree**

**Adaptive Boost  
Classifier**

**Extra Trees  
Classifier**

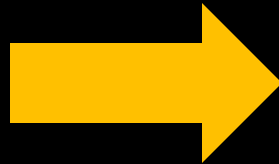
## Credit Risk Analysis

# Heat Map of Logistic Regression



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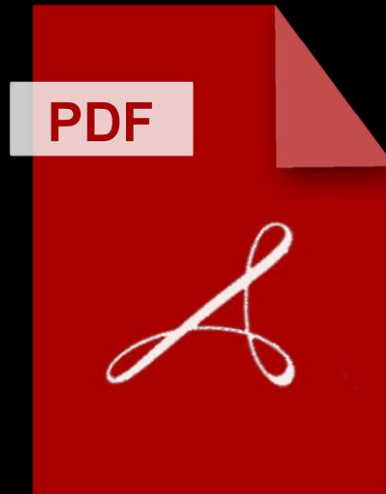
## Visualisation of Data



## Conclusion

Credit Risk Analysis is a very crucial part of the banking sector and it plays an important role in the growth of the bank's profit. Using analysing techniques one can predict or analyse that a person applying for loan will repay the loan or not. So, multiples algorithms have been implemented to analyse a defaulter. The best model selected out of all models that have been tested is Logistic Regression model with an accuracy of 99.957%, which is cross verified with K-fold cross validation technique having the same accuracy of 99.956%. Further insights gained using visualizations from Tableau is that the bank is really working hard to come up from crisis of loss of revenue. To overcome this crisis, credit risk analysis will help them to grow and earn profit.

## Documentation for the Credit Risk Analysis



**Thank You !**