

LOAN STATUS PREDICTION

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Problem Statement

- *Sometimes it's hard to know whether a person can get loan or not, from banks and other lenders beforehand.*
- *This model aims to make it easier for both lenders and borrowers to understand if a loan will be approved or not, making the whole process simpler.*

Data Exploration

- **Importing dataset:** Taking required dataset from kaggle and linking directly.
- **Packages:** Importing required packages like pandas, numpy, seaborn etc...
- **Data cleaning:** Cleaning the dataset by removing the missing values.
- **Data Processing:** Changing the labels of the columns values and dropping.
- **Data Visualization:** Generating graphs for analysis.

Model Building

- **Train and Test:** *Divided the data set into 80:20 ratio for training and testing sets.*
- **Algorithm:** *Chose Support Vector Machine (SVM) as the algorithm for loan status prediction.*
- **Optimization:** *Worked on different kernel types for more accuracy rate.*

Model Selection

- **Decision Tree:**

`DecisionTreeClassifier()` accuracy is 0.7604166666666666

`DecisionTreeClassifier()` Avg cross val score is 0.7020833333333334

- **Logistic Regression:**

`LogisticRegression()` accuracy is 0.8229166666666666

`LogisticRegression()` Avg cross val score is 0.8020833333333334

- **Support Vector Machine:**

Accuracy on training data : 80.46875

Accuracy on test data : 83.33333333333334

Performance Analysis

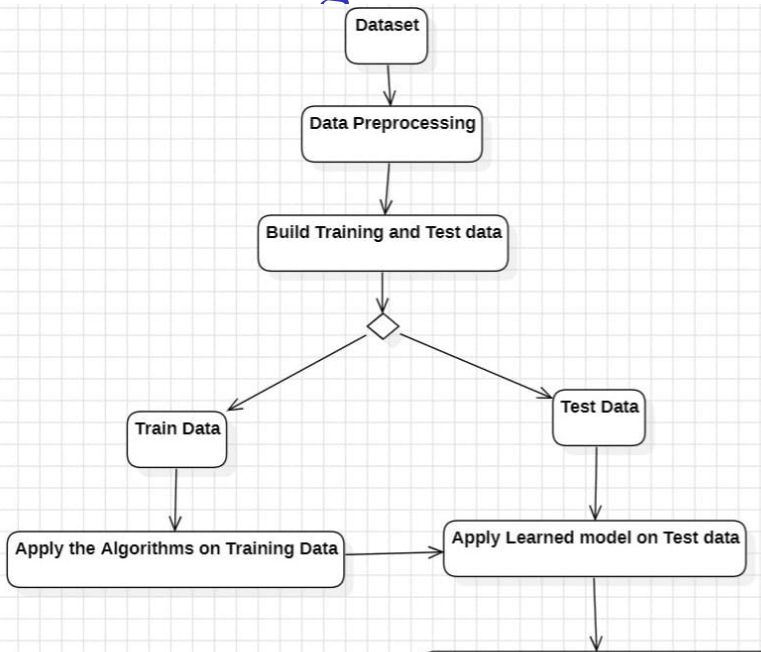
- **Accuracy rate for loan approvals by svm:**

```
Accuracy on training data : 70.3125  
Accuracy on test data : 68.75
```

- **Optimized accuracy rate:**

```
Accuracy on training data : 80.46875  
Accuracy on test data : 83.33333333333334
```

Data Flow Diagram



Challenges

- *Faced challenges in dropping the missing values and labeling in dataset.*
- *Worked on many model building algorithms and optimization techniques for getting more accuracy rates..*
- *Faced challenges in linking model to the user interface.*

Output

Loan Status Prediction

Gender:

Married:

Dependents:

Education:

Self Employed:

Applicant Income:

Coapplicant Income:

Yes



Applicant Income:

1000

Coapplicant Income:

2160

Loan Amount:

3400

Loan Amount Term:

360

Credit History:

1



Property Area:

Urban



Predict Loan

Prediction Result:

Loan Approved

[Go back to home page](#)

Loan Status Prediction

Gender:

Female



Married:

Yes



Dependents:

0

Education:

Not Graduate



Self Employed:

No



Applicant Income:

100

Coapplicant Income:

0

Self Employed:

No

Applicant Income:

100

Coapplicant Income:

0

Loan Amount:

344

Loan Amount Term:

360

Credit History:

0

Property Area:

Rural

Predict Loan

Prediction Result:

Loan Not Approved

[Go back to home page](#)

Bibliography

- *Selected dataset from kaggle*
 - ① *Ninzaami-loan prediction*
 - ② *abdelruhmanessam-loan-status-prediction*
- *Took reference from the Youtube videos*
 - ① *Loan status prediction by Siddardhan*
 - ② *ML project-loan prediction by Data Thinkers*
 - ③ *Loan Eligibility Prediction Tutorial by edureka!*
- *Used ChatGPT,Bard and Google for understanding terms and resolving challenges encountered during the project.*

THANK YOU