In [663... import numpy as np import pandas as pd import matplotlib.pyplot as plt In [664... df = pd.read_csv("/Users/dev/Personal/DS & AI Class Notes/Data Sets/Adaboos In [665... df.drop("Loan_ID",axis=1,inplace=True) Gender Married Dependents Education Self_Employed ApplicantIncome Coapplicat Out [665... 0 5849 Male No 0 Graduate No Male 1 Yes Graduate No 4583 2 Male 0 Graduate 3000 Yes Yes Not Male 0 2583 Yes No Graduate Male No 0 Graduate No 6000 609 Female No 0 Graduate No 2900

3+

1

2

0

Graduate

Graduate

Graduate

Graduate

No

No

No

Yes

4106

8072

7583

4583

614 rows × 12 columns

Male

Male

Male

Female

Yes

Yes

Yes

No

In [666...

df.info()

610

611

612

613

<class 'pandas.core.frame.DataFrame'> RangeIndex: 614 entries, 0 to 613 Data columns (total 12 columns): Non-Null Count Dtype Column 0 Gender 601 non-null object Married 611 non-null object 1 599 non-null object 614 non-null object 2 Dependents 3 Education Self_Employed 582 non-null object 5 ApplicantIncome 614 non-null int64 CoapplicantIncome 614 non-null 6 float64 7 LoanAmount 592 non-null float64 Loan Amount Term 600 non-null float64 564 non-null 9 Credit History float64 10 Property_Area 614 non-null object object 11 Loan Status 614 non-null dtypes: float64(4), int64(1), object(7) memory usage: 57.7+ KB

Replacing df["Dependents"] values by str to int. by Regex

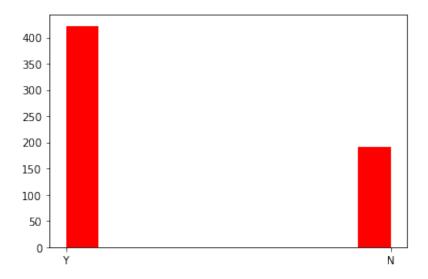
```
In [667... df["Dependents"].replace("[^0-9.]","",regex=True,inplace=True)
```

Removing Nans

```
In [668...
          df.isna().sum()
         Gender
                                13
Out [668...
                                 3
         Married
          Dependents
                                15
          Education
                                 0
          Self Employed
          ApplicantIncome
                                 0
          CoapplicantIncome
                                 0
          LoanAmount
                                22
                                14
          Loan_Amount_Term
          Credit_History
                                50
                                 0
          Property Area
          Loan Status
                                 0
          dtype: int64
In [669...
          df["Gender"].fillna(method='pad',inplace=True)
In [670...
          df["Married"].fillna(method='pad',inplace=True)
In [671...
          df["Dependents"].fillna(method='pad',inplace=True)
```

```
In [672...
           df["Self_Employed"].fillna(method='pad',inplace=True)
In [673...
           def nanremove(df):
               m = round(df.mean(), 2)
               df.replace(np.nan,m,inplace = True)
In [674...
           nanremove(df["LoanAmount"])
In [675...
           nanremove(df["Loan_Amount_Term"])
In [676...
           nanremove(df["Credit History"])
In [677...
           df.isna().sum()
          Gender
                                0
Out [677...
                                0
          Married
          Dependents
                                0
          Education
                                0
          Self_Employed
          ApplicantIncome
          CoapplicantIncome
          LoanAmount
          Loan Amount Term
                                0
          Credit_History
          Property_Area
                                0
          Loan_Status
                                0
          dtype: int64
```

Checking Balance of DF



Checking And Removing 0's

```
In [680...
           def checkz(df):
               return df[df == 0].value_counts()
In [681...
           for i in df.columns:
               print(checkz(df[i]))
          Series([], Name: Gender, dtype: int64)
          Series([], Name: Married, dtype: int64)
          Series([], Name: Dependents, dtype: int64)
          Series([], Name: Education, dtype: int64)
          Series([], Name: Self_Employed, dtype: int64)
          Series([], Name: ApplicantIncome, dtype: int64)
          0.0
                  273
          Name: CoapplicantIncome, dtype: int64
          Series([], Name: LoanAmount, dtype: int64)
          Series([], Name: Loan_Amount_Term, dtype: int64)
          0.0
                 89
          Name: Credit_History, dtype: int64
          Series([], Name: Property Area, dtype: int64)
          Series([], Name: Loan_Status, dtype: int64)
In [682...
           df.columns
          Index(['Gender', 'Married', 'Dependents', 'Education', 'Self_Employed',
Out [682...
                  'ApplicantIncome', 'CoapplicantIncome', 'LoanAmount', 'Loan_Amount_Term', 'Credit_History', 'Property_Area', 'Loan_Status'
          ],
                dtype='object')
In [683...
           clist = [ 'Dependents', 'ApplicantIncome', 'CoapplicantIncome',
                     'LoanAmount', 'Loan_Amount_Term', 'Credit_History', ]
```

```
In [684...
          for i in clist:
              zeroremove(df[i])
In [685...
          for i in df.columns:
              print(checkz(df[i]))
         Series([], Name: Gender, dtype: int64)
         Series([], Name: Married, dtype: int64)
         Series([], Name: Dependents, dtype: int64)
         Series([], Name: Education, dtype: int64)
         Series([], Name: Self_Employed, dtype: int64)
         Series([], Name: ApplicantIncome, dtype: int64)
         Series([], Name: CoapplicantIncome, dtype: int64)
         Series([], Name: LoanAmount, dtype: int64)
         Series([], Name: Loan_Amount_Term, dtype: int64)
         Series([], Name: Credit_History, dtype: int64)
         Series([], Name: Property_Area, dtype: int64)
         Series([], Name: Loan Status, dtype: int64)
         OD_Tech With The help of Skew
In [686...
          def odiqr(df):
              q1 = df.quantile(0.25)
              q3 = df.quantile(0.75)
              iqr = q3 - q1
              low = q1 - (1.5 * iqr)
              high = q3 + (1.5 * iqr)
              m = df.mean()
              df = df.apply(lambda x : m if x < low else (m if x > high else x ) )
              return df
In [687...
          def odmsd(df):
              m = round(df.mean(), 2)
              s = round(df.std(),2)
              low = round(m-(3*s), 2)
              high = round(m+(3*s),2)
              ft1 = df[df<low]</pre>
              ft2 = df[df>high]
              df = df.map(lambda x : low if x < low else (high if x > high else x ))
In [688...
          for i in clist:
              print(f'{i} is {df[i].skew()}')
         Dependents is 0.9764220531079079
         ApplicantIncome is 6.539513113994625
         CoapplicantIncome is 9.677702794378638
         LoanAmount is 2.7266039186591304
         Loan Amount Term is -2.3896799467122745
         Credit History is -1.3108351339315514
```

```
In [689...
    for i in clist:
        print(f'{i} is {df[i].dtype}')

Dependents is object
    ApplicantIncome is int64
    CoapplicantIncome is float64
    LoanAmount is float64
    Loan Amount Term is float64
```

Bcz df["Dependents"] is str type

Credit_History is float64

```
In [690...
          df["Dependents"] = df["Dependents"].astype(int)
In [691...
          for i in clist:
               if df[i].skew() <= 0.5:</pre>
                   odmsd(df[i])
               else:
                   df[i] = odiqr(df[i])
In [692...
          for i in clist:
              print(f'{i} is {df[i].skew()}')
          Dependents is 0.9764220531079079
         ApplicantIncome is 0.9131963245496909
         CoapplicantIncome is 0.6991764314798369
         LoanAmount is 0.36795693965510146
         Loan Amount Term is -2.3896799467122745
         Credit History is -1.3108351339315514
In [693...
          df
```

Out[693		Gender	Married	Dependents	Education	Self_Employed	ApplicantIncome	Coapplica
	0	Male	No	0	Graduate	No	5849.0	162
	1	Male	Yes	1	Graduate	No	4583.0	150
	2	Male	Yes	0	Graduate	Yes	3000.0	162
	3	Male	Yes	0	Not Graduate	No	2583.0	235
	4	Male	No	0	Graduate	No	6000.0	162
	•••							
	609	Female	No	0	Graduate	No	2900.0	162
	610	Male	Yes	3	Graduate	No	4106.0	162
	611	Male	Yes	1	Graduate	No	8072.0	234
	612	Male	Yes	2	Graduate	No	7583.0	162
	613	Female	No	0	Graduate	Yes	4583.0	162

Out[698		Married	Dependents	Education	Self_Employed	ApplicantIncome	CoapplicantIncome
	0	No	0	Graduate	No	5849.0	1621.250000
	1	Yes	1	Graduate	No	4583.0	1508.000000
	2	Yes	0	Graduate	Yes	3000.0	1621.250000
	3	Yes	0	Not Graduate	No	2583.0	2358.000000
	4	No	0	Graduate	No	6000.0	1621.250000
	•••	•••					
	609	No	0	Graduate	No	2900.0	1621.250000
	610	Yes	3	Graduate	No	4106.0	1621.250000
	611	Yes	1	Graduate	No	8072.0	2342.094739
	612	Yes	2	Graduate	No	7583.0	1621.250000
	613	No	0	Graduate	Yes	4583.0	1621.250000

ut[702		Dependents	Education	Self_Employed	ApplicantIncome	CoapplicantIncome	LoanAm
	0	0	Graduate	No	5849.0	1621.250000	14
	1	1	Graduate	No	4583.0	1508.000000	12
	2	0	Graduate	Yes	3000.0	1621.250000	6
	3	0	Not Graduate	No	2583.0	2358.000000	12
	4	0	Graduate	No	6000.0	1621.250000	14
	•••						
	609	0	Graduate	No	2900.0	1621.250000	5
	610	3	Graduate	No	4106.0	1621.250000	4
	611	1	Graduate	No	8072.0	2342.094739	25
	612	2	Graduate	No	7583.0	1621.250000	18
	613	0	Graduate	Yes	4583.0	1621.250000	13

614 rows × 12 columns

0u

Out [707...

	Dependents	Self_Employed	ApplicantIncome	CoapplicantIncome	LoanAmount	Loan_
0	0	No	5849.0	1621.250000	146.41	
1	1	No	4583.0	1508.000000	128.00	
2	0	Yes	3000.0	1621.250000	66.00	
3	0	No	2583.0	2358.000000	120.00	
4	0	No	6000.0	1621.250000	141.00	
•••						
609	0	No	2900.0	1621.250000	71.00	
610	3	No	4106.0	1621.250000	40.00	
611	1	No	8072.0	2342.094739	253.00	
612	2	No	7583.0	1621.250000	187.00	
613	0	Yes	4583.0	1621.250000	133.00	

\cap	11.1	+		7	1	7
U	u	L	L	/	т.	Z

	Dependents	ApplicantIncome	CoapplicantIncome	LoanAmount	Loan_Amount_Term
0	0	5849.0	1621.250000	146.41	360.0
1	1	4583.0	1508.000000	128.00	360.0
2	0	3000.0	1621.250000	66.00	360.0
3	0	2583.0	2358.000000	120.00	360.0
4	0	6000.0	1621.250000	141.00	360.0
•••					
609	0	2900.0	1621.250000	71.00	360.0
610	3	4106.0	1621.250000	40.00	180.0
611	1	8072.0	2342.094739	253.00	360.0
612	2	7583.0	1621.250000	187.00	360.0
613	0	4583.0	1621.250000	133.00	360.0

Out[717		Dependents	ApplicantIncome	CoapplicantIncome	LoanAmount	Loan_Amount_Term
	0	0	5849.0	1621.250000	146.41	360.0
	1	1	4583.0	1508.000000	128.00	360.0
	2	0	3000.0	1621.250000	66.00	360.0
	3	0	2583.0	2358.000000	120.00	360.0
	4	0	6000.0	1621.250000	141.00	360.0
	•••					
	609	0	2900.0	1621.250000	71.00	360.0
	610	3	4106.0	1621.250000	40.00	180.0
	611	1	8072.0	2342.094739	253.00	360.0
	612	2	7583.0	1621.250000	187.00	360.0
	613	0	4583.0	1621.250000	133.00	360.0

614 rows × 14 columns

Encoding of df["Loan_Status"] with Label Encoder

```
In [718...
           from sklearn.preprocessing import LabelEncoder
In [719...
           le = LabelEncoder()
In [720...
           le.fit(df["Loan_Status"])
         LabelEncoder()
Out [720...
In [721...
           df = pd.concat([ df, pd.DataFrame(le.fit_transform(df["Loan_Status"]),column

                           ,axis=1).drop("Loan_Status",axis=1)
In [722...
           le.classes_
          array(['N', 'Y'], dtype=object)
Out[722...
In [723...
           df = df[['Male','Married_Yes','Dependents','Not Graduate',
                     'Self_Employed_Yes','ApplicantIncome','CoapplicantIncome',
                   'LoanAmount', 'Loan_Amount_Term', 'Credit_History', 'Rural', 'Semiurbar
```

In [724...

df

Out [724...

	Male	Married_Yes	Dependents	Not Graduate	Self_Employed_Yes	ApplicantIncome	Coal
0	1	0	0	0	0	5849.0	
1	1	1	1	0	0	4583.0	
2	1	1	0	0	1	3000.0	
3	1	1	0	1	0	2583.0	
4	1	0	0	0	0	6000.0	
•••							
609	0	0	0	0	0	2900.0	
610	1	1	3	0	0	4106.0	
611	1	1	1	0	0	8072.0	
612	1	1	2	0	0	7583.0	
613	0	0	0	0	1	4583.0	

614 rows × 14 columns

Splitting

```
In [725...
          X = df.drop("LOAN_Status",axis=1)
In [726...
           X.sample()
Out [726...
                                                   Not
               Male Married_Yes Dependents
                                                        Self_Employed_Yes ApplicantIncome Coap
                                              Graduate
                                           0
                                                                        0
          281
                                                                                    3927.0
In [727...
          y = df["LOAN_Status"]
Out[728... 363 1
Name: LOAN_Status, dtype: int64
```

```
In [729...
          from sklearn.model_selection import train_test_split
          from sklearn.preprocessing import StandardScaler
          from sklearn.model_selection import KFold , cross_val_score
          from sklearn.svm import SVC
          from sklearn.tree import DecisionTreeClassifier
          from sklearn.ensemble import RandomForestClassifier , AdaBoostClassifier
In [730...
          Xtrain, Xtest, ytrain, ytest = train_test_split(X, y, test_size=.20)
In [731...
          X.shape , Xtrain.shape , Xtest.shape
         ((614, 13), (491, 13), (123, 13))
Out [731...
In [732...
          y.shape , ytrain.shape , ytest.shape
         ((614,), (491,), (123,))
Out [732...
In [733...
          kf = KFold(n splits=11)
In [734...
          dct = DecisionTreeClassifier()
In [735...
          rfc = RandomForestClassifier()
In [736...
          algo = [ dct , rfc ]
         Without Feature Scaling
```

With Feature Scaling

```
In [738... ss = StandardScaler()

In [739... ss.fit(Xtrain)
```

With Cross Validation

RandomForestClassifier() = 0.7507969303423849

Boosting

Without Feature Scaling

```
for i in algo1:
    i.fit(Xtrain,ytrain)
    s = i.score(Xtest,ytest)
    print(f'{i} = {s}')
```

```
RandomForestClassifier(max_depth=2, max_leaf_nodes=3, n_estimators=150) = 0
.6910569105691057
AdaBoostClassifier() = 0.7642276422764228
```

With Feature Scaling

```
for i in algo1:
    i.fit(Xtrain_ss,ytrain)
    s = i.score(Xtest_ss,ytest)
    print(f'{i} = {s}')

RandomForestClassifier(max_depth=2, max_leaf_nodes=3, n_estimators=150) = 0
```

.6910569105691057 AdaBoostClassifier() = 0.7642276422764228

With Cross Validation (Boosting)

With GridSearch CV

```
In [751... from sklearn.model_selection import GridSearchCV

In [752... dic = { 'n_estimators' : [100,125,180], 'criterion': ['gini', 'entropy'], 'r , 'min_samples_leaf' : [ 1,5] }

In [753... kf1 = KFold(n_splits=12)

In [754... gvc = GridSearchCV(RandomForestClassifier(),param_grid=dic,cv = kf1)

In [755... %%time gvc.fit(X,y)
```

```
CPU times: user 46.1 s, sys: 399 ms, total: 46.5 s
         Wall time: 46.7 s
         GridSearchCV(cv=KFold(n splits=12, random state=None, shuffle=False),
Out [755...
                       estimator=RandomForestClassifier(),
                       param_grid={'criterion': ['gini', 'entropy'],
                                    'max_depth': [2, 3, 10], 'min_samples_leaf': [1, 5
          ],
                                    'n_estimators': [100, 125, 180]})
In [756...
          gvc.best params
Out[756... {'criterion': 'entropy',
           'max_depth': 10,
           'min_samples_leaf': 5,
           'n estimators': 180}
In [757...
          gvc.best estimator
Out[757... RandomForestClassifier(criterion='entropy', max_depth=10, min_samples_leaf=
                                 n estimators=180)
In [758...
          gvc.best_score_
          0.763826043237808
Out [758...
 In []:
```