```
In [1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
```

In [42]: kisan=pd.read_csv('data_folder/kisan_net_log.csv')
kisan.head()

C:\Users\Atul kumar\anaconda3\lib\site-packages\IPython\core\interactiveshell.p
y:3063: DtypeWarning: Columns (19,29,30) have mixed types.Specify dtype option
on import or set low_memory=False.

interactivity=interactivity, compiler=compiler, result=result)

Out[42]:

	loan_amnt	term	int_rate	installment	grade	sub_grade	emp_title	emp_length	home_ow
0	10000.0	36 months	11.44	329.48	В	B4	Marketing	10+ years	
1	Mendozaberg	OK 22690"	NaN	NaN	NaN	NaN	NaN	NaN	
2	8000.0	36 months	11.99	265.68	В	B5	Credit analyst	4 years	MOR
3	Loganmouth	SD 05113"	NaN	NaN	NaN	NaN	NaN	NaN	
4	15600.0	36 months	10.49	506.97	В	В3	Statistician	< 1 year	

5 rows × 31 columns

localhost:8888/notebooks/Kisan_logistic_regression.ipynb

In [43]: kisan1=kisan.iloc[0::2,0:27]
kisan1.head()

Out[43]:

	loan_amnt	term	int_rate	installment	grade	sub_grade	emp_title	emp_length	home_ow
0	10000.0	36 months	11.44	329.48	В	В4	Marketing	10+ years	
2	8000.0	36 months	11.99	265.68	В	B5	Credit analyst	4 years	MOR
4	15600.0	36 months	10.49	506.97	В	ВЗ	Statistician	< 1 year	
6	7200.0	36 months	6.49	220.65	Α	A2	Client Advocate	6 years	
8	24375.0	60 months	17.27	609.33	С	C5	Destiny Management Inc.	9 years	MOR

5 rows × 27 columns

EDA

Target Variable-Loan_Status

```
In [44]: kisan1['loan_status'].value_counts()
```

```
Out[44]: Fully Paid 78498
Charged Off 19294
Jun-13 125
Aug-13 107
May-13 99
```

35000 1 40000 1 Sep-07 1 42000 1 81090 1

Name: loan_status, Length: 127, dtype: int64

```
In [45]: kisan2=kisan1[(kisan1['loan status']=='Fully Paid')|(kisan1['loan status']=='Char
         kisan2.shape
Out[45]: (97792, 27)
In [46]: kisan2['loan status'].value counts(normalize=True)
         #Highly Imbalanced dataset
Out[46]: Fully Paid
                        0.802704
         Charged Off
                        0.197296
         Name: loan status, dtype: float64
In [47]: kisan2.info()
         <class 'pandas.core.frame.DataFrame'>
         Int64Index: 97792 entries, 0 to 200488
         Data columns (total 27 columns):
              Column
                                    Non-Null Count Dtype
              -----
                                     -----
                                                     ____
              loan amnt
          0
                                    97792 non-null object
          1
              term
                                    97792 non-null
                                                    object
          2
              int_rate
                                    97792 non-null float64
          3
              installment
                                    97792 non-null float64
          4
              grade
                                    97792 non-null object
          5
              sub_grade
                                    97792 non-null
                                                    object
          6
              emp title
                                    92061 non-null
                                                    object
          7
              emp length
                                    93190 non-null
                                                    object
          8
              home_ownership
                                    97792 non-null
                                                    object
          9
              annual inc
                                    97792 non-null
                                                    object
          10
             verification status
                                    97792 non-null
                                                    object
          11
              issue d
                                    97792 non-null
                                                    object
          12
              loan status
                                    97792 non-null
                                                    object
          13
                                    97792 non-null
                                                    object
              purpose
          14 title
                                    97351 non-null
                                                    object
          15
              dti
                                    97792 non-null
                                                    object
          16
              earliest cr line
                                    97792 non-null
                                                    object
          17
              open acc
                                    97792 non-null
                                                    object
          18 pub rec
                                    97792 non-null
                                                    object
          19 revol bal
                                    97792 non-null
                                                    object
          20 revol util
                                    97723 non-null
                                                    float64
          21 total acc
                                    97792 non-null
                                                    float64
          22
              initial list status
                                    97792 non-null
                                                    object
          23 application_type
                                    97792 non-null
                                                    object
          24
              mort acc
                                    88924 non-null
                                                    object
          25
              pub rec bankruptcies 97607 non-null
                                                    object
          26 address
                                    97782 non-null
                                                    object
         dtypes: float64(4), object(23)
         memory usage: 20.9+ MB
```

```
In [48]: kisan2.isnull().sum()
Out[48]: loan amnt
                                      0
                                      0
         term
         int rate
                                      0
         installment
                                      0
                                      0
         grade
         sub_grade
                                      0
         emp title
                                   5731
                                   4602
         emp length
         home_ownership
                                      0
         annual inc
                                      0
         verification_status
                                      0
         issue d
                                      0
         loan status
                                      0
         purpose
                                      0
         title
                                    441
         dti
                                      0
         earliest_cr_line
                                      0
         open_acc
                                      0
         pub rec
                                      0
         revol bal
                                      0
         revol util
                                     69
         total_acc
                                      0
         initial list status
                                      0
         application_type
                                      0
                                   8868
         mort_acc
                                    185
         pub rec bankruptcies
                                     10
         address
         dtype: int64
In [49]: import re
         pattern='[a-zA-Z]'
         cols=['loan_amnt', 'annual_inc', 'dti', 'revol_bal', 'mort_acc']
         for i in cols:
              kisan2[i]=list(map(lambda x:'nan'if re.search(pattern,str(x))else x,kisan2[i]
              kisan2=kisan2[kisan2[i]!='nan']
         kisan2.shape
         C:\Users\Atul kumar\anaconda3\lib\site-packages\ipykernel launcher.py:5: Settin
         gWithCopyWarning:
         A value is trying to be set on a copy of a slice from a DataFrame.
         Try using .loc[row indexer,col indexer] = value instead
         See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/sta
         ble/user guide/indexing.html#returning-a-view-versus-a-copy (https://pandas.pyd
         ata.org/pandas-docs/stable/user guide/indexing.html#returning-a-view-versus-a-c
         opy)
            .. .. ..
Out[49]: (88741, 27)
```

Changing the data types of some features to floats

<class 'pandas.core.frame.DataFrame'>
Int64Index: 88741 entries, 0 to 200488
Data columns (total 27 columns):

Data	columns (total 27 columns)	umns):	
#	Column	Non-Null Count	Dtype
0	loan_amnt	88741 non-null	float64
1	term	88741 non-null	object
2	int_rate	88741 non-null	float64
3	installment	88741 non-null	float64
4	grade	88741 non-null	object
5	sub_grade	88741 non-null	object
6	emp_title	83607 non-null	object
7	emp_length	84403 non-null	object
8	home_ownership	88741 non-null	object
9	annual_inc	88741 non-null	float64
10	verification_status	88741 non-null	object
11	issue_d	88741 non-null	object
12	loan_status	88741 non-null	object
13	purpose	88741 non-null	object
14	title	88302 non-null	object
15	dti	88741 non-null	float64
16	earliest_cr_line	88741 non-null	object
17	open_acc	88741 non-null	object
18	pub_rec	88741 non-null	object
19	revol_bal	88741 non-null	float64
20	revol_util	88684 non-null	float64
21	total_acc	88741 non-null	float64
22	<pre>initial_list_status</pre>	88741 non-null	object
23	application_type	88741 non-null	object
24	mort_acc	88741 non-null	object
25	<pre>pub_rec_bankruptcies</pre>	88741 non-null	object
26	address	88741 non-null	object
dtype	es: float64(8), object	(19)	
memoi	∽y usage: 19.0+ MB		

localhost:8888/notebooks/Kisan_logistic_regression.ipynb

```
In [51]: cols=['open_acc', 'pub_rec', 'mort_acc',
                pub rec bankruptcies']
         for i in cols:
             kisan2[i]=list(map(lambda x:int(x),kisan2[i]))
         kisan2.info()
         <class 'pandas.core.frame.DataFrame'>
         Int64Index: 88741 entries, 0 to 200488
         Data columns (total 27 columns):
              Column
                                    Non-Null Count Dtype
              -----
                                    _____
                                                   ----
          0
              loan amnt
                                    88741 non-null float64
                                    88741 non-null object
          1
              term
          2
              int rate
                                    88741 non-null float64
          3
              installment
                                    88741 non-null float64
          4
              grade
                                    88741 non-null object
          5
              sub_grade
                                    88741 non-null object
          6
              emp title
                                    83607 non-null object
          7
              emp length
                                    84403 non-null object
          8
              home ownership
                                    88741 non-null object
          9
              annual inc
                                    88741 non-null float64
          10 verification status
                                    88741 non-null object
          11 issue d
                                    88741 non-null object
          12 loan_status
                                    88741 non-null object
          13 purpose
                                    88741 non-null object
          14 title
                                    88302 non-null object
          15 dti
                                    88741 non-null
                                                   float64
                                    88741 non-null object
          16 earliest cr line
          17 open_acc
                                    88741 non-null
                                                   int64
          18 pub_rec
                                    88741 non-null int64
          19 revol bal
                                    88741 non-null float64
          20 revol util
                                    88684 non-null float64
          21 total acc
                                    88741 non-null float64
          22 initial_list_status
                                    88741 non-null object
          23 application_type
                                    88741 non-null object
                                    88741 non-null
                                                   int64
          24 mort_acc
          25
              pub rec bankruptcies 88741 non-null
                                                   int64
          26 address
                                    88741 non-null
                                                   object
         dtypes: float64(8), int64(4), object(15)
         memory usage: 19.0+ MB
In [52]: kisan2['title'].value_counts()
Out[52]: Debt consolidation
                                                37962
         Credit card refinancing
                                                12970
         Home improvement
                                                 3865
         Other
                                                 3179
         Debt Consolidation
                                                 2191
         Home Improvement/Consolidation
                                                    1
         loan payback then purchase new shop
                                                    1
         consolidation of credit cards
                                                    1
         Debt Consolidation & CC Reduction
                                                    1
         Name: title, Length: 10289, dtype: int64
```

```
In [53]: kisan2['purpose'].value counts()
Out[53]: debt consolidation
                                 53511
          credit card
                                 19352
          home_improvement
                                  5359
          other
                                  4275
                                  1723
          major_purchase
          small_business
                                   939
          medical
                                   889
          car
                                   808
                                   599
          moving
          vacation
                                   549
          house
                                   468
          wedding
                                   215
          renewable energy
                                    53
          educational
                                     1
          Name: purpose, dtype: int64
          Categorising purpose into:debt consolidation,credit card and others
In [54]: | z=['debt_consolidation','credit_card']
          kisan2['purpose']=list(map(lambda x:x if x in z else 'others',kisan2['purpose']))
          kisan2['purpose'].value_counts()
Out[54]: debt consolidation
                                 53511
          credit card
                                 19352
          others
                                 15878
          Name: purpose, dtype: int64
In [55]: kisan2['home ownership'].value counts()
Out[55]: MORTGAGE
                      45023
          RENT
                      35083
                       8613
          OWN
          OTHER
                         13
          NONE
                          7
          ANY
                          2
          Name: home_ownership, dtype: int64
          Categorising home ownership into:Mortgage,Rent and Others
In [56]: z=['MORTGAGE', 'RENT']
          kisan2['home ownership']=list(map(lambda x:x if x in z else 'OTHERS', kisan2['home
          kisan2['home ownership'].value counts()
Out[56]: MORTGAGE
                      45023
          RENT
                      35083
          OTHERS
                       8635
          Name: home_ownership, dtype: int64
```

```
In [57]: kisan2['verification status'].value counts()
Out[57]: Verified
                              31144
          Source Verified
                              30265
          Not Verified
                              27332
          Name: verification_status, dtype: int64
          Merging the Veified and Source Verified for verification status
         z=['Verified','Source Verified']
In [58]:
          kisan2['verification_status']=list(map(lambda x:'Verified' if x in z else x,kisar
          kisan2['verification_status'].value_counts()
Out[58]: Verified
                          61409
          Not Verified
                          27332
          Name: verification_status, dtype: int64
          pub rec:No. of derogatory public records
In [59]: kisan2['pub_rec'].value_counts()
Out[59]: 0
                74769
                11950
          1
          2
                 1393
          3
                  360
          4
                  147
          5
                   63
          6
                   27
                   15
          7
                    8
          10
                    3
                    2
          11
          9
                    2
          19
                    1
          13
                    1
          Name: pub_rec, dtype: int64
In [60]:
         z = [0, 1]
          kisan2['pub_rec']=list(map(lambda x:x if x in z else 1,kisan2['pub_rec']))
          kisan2['pub rec'].value counts()
Out[60]: 0
               74769
               13972
          Name: pub rec, dtype: int64
```

```
In [61]: kisan2['term'].value counts()
Out[61]:
                                              36 months
                                                                                                       67798
                                              60 months
                                                                                                       20943
                                         Name: term, dtype: int64
In [62]: kisan2['is_36mnths']=list(map(lambda x:1 if x==' 36 months'else 0,kisan2['term'])
                                         kisan2['is_36mnths'].value_counts()
Out[62]: 1
                                                               67798
                                                               20943
                                         Name: is_36mnths, dtype: int64
In [63]: kisan2['is_verified']=list(map(lambda x:1 if x=='Verified'else 0,kisan2['verification of the content of the cont
                                         kisan2['is_verified'].value_counts()
Out[63]: 1
                                                               61409
                                                               27332
                                         Name: is_verified, dtype: int64
In [64]: |kisan2['grade'].value_counts()
Out[64]: B
                                                               25968
                                                               24431
                                         C
                                         D
                                                               14598
                                         Α
                                                               13215
                                                                    7207
                                         Ε
                                                                    2642
                                         G
                                                                        680
                                         Name: grade, dtype: int64
```

```
In [65]: d={'A':0,'B':1,'C':2,'D':3,'E':4,'F':5,'G':6}
    kisan2['mod_grade']=list(map(lambda x:d[x],kisan2['grade']))
    kisan2[['grade','mod_grade']].head(20)
```

Out[65]:

	grade	mod_grade
0	В	1
2	В	1
4	В	1
6	Α	0
8	С	2
10	С	2
12	Α	0
14	В	1
16	В	1
18	С	2
20	В	1
22	С	2
24	В	1
26	С	2
28	Α	0
30	Α	0
32	Е	4
34	С	2
36	Α	0
38	Α	0

```
In [66]: kisan2=pd.get_dummies(kisan2,columns=['home_ownership','purpose'],drop_first=True
```

```
In [67]: cols=['home ownership OTHERS','home ownership RENT','purpose debt consolidation'.
               purpose others'
         for i in cols:
             kisan2[i]=list(map(lambda x:int(x),kisan2[i]))
         kisan2.info()
         <class 'pandas.core.frame.DataFrame'>
         Int64Index: 88741 entries, 0 to 200488
         Data columns (total 32 columns):
          #
              Column
                                          Non-Null Count Dtype
          0
              loan amnt
                                          88741 non-null
                                                          float64
              term
                                          88741 non-null object
          1
          2
                                          88741 non-null float64
              int rate
          3
              installment
                                          88741 non-null float64
          4
              grade
                                          88741 non-null object
          5
              sub_grade
                                          88741 non-null
                                                          object
          6
                                          83607 non-null object
              emp title
          7
              emp length
                                          84403 non-null object
          8
                                          88741 non-null float64
              annual inc
          9
              verification status
                                          88741 non-null object
          10
                                          88741 non-null object
              issue d
          11 loan_status
                                          88741 non-null object
          12
             title
                                          88302 non-null
                                                          obiect
                                          88741 non-null float64
          13
             dti
                                          88741 non-null object
          14 earliest_cr_line
          15
              open acc
                                          88741 non-null int64
                                          88741 non-null int64
          16 pub rec
          17
              revol_bal
                                          88741 non-null float64
             revol util
                                          88684 non-null float64
          18
          19
              total acc
                                          88741 non-null float64
          20
              initial_list_status
                                          88741 non-null object
          21 application type
                                          88741 non-null object
          22
              mort acc
                                          88741 non-null int64
          23
              pub rec bankruptcies
                                          88741 non-null int64
          24
             address
                                          88741 non-null
                                                          object
          25 is 36mnths
                                                          int64
                                          88741 non-null
          26 is verified
                                          88741 non-null
                                                         int64
          27
              mod_grade
                                          88741 non-null
                                                         int64
                                                         int64
          28
              home ownership OTHERS
                                          88741 non-null
          29
              home_ownership_RENT
                                          88741 non-null
                                                          int64
          30
              purpose debt consolidation 88741 non-null
                                                          int64
          31 purpose others
                                          88741 non-null
                                                          int64
         dtypes: float64(8), int64(11), object(13)
         memory usage: 22.3+ MB
```

Type *Markdown* and LaTeX: α^2

```
In [68]: kisan2['application type'].value counts(normalize=True)
Out[68]: INDIVIDUAL
                        0.998073
         JOINT
                        0.001127
         DIRECT PAY
                        0.000800
         Name: application_type, dtype: float64
In [69]: kisan2['initial_list_status'].value_counts(normalize=True)
Out[69]: f
              0.555775
              0.444225
         Name: initial_list_status, dtype: float64
In [70]:
         d={'f':0,'w':1}
         kisan2['initial_list_status']=list(map(lambda x:d[x] ,kisan2['initial_list_status
         kisan2['initial list status'].value counts()
Out[70]:
              49320
         1
               39421
         Name: initial list status, dtype: int64
In [72]: kisan2['pub rec bankruptcies'].value counts()
Out[72]: 0
              77846
              10303
         1
         2
                 475
                  79
         3
                  25
         4
         5
                   8
         6
                   4
         Name: pub_rec_bankruptcies, dtype: int64
         pub rec bankruptcies: Number of public record bankruptcies
In [73]: z=[0,1]
         kisan2['pub rec bankruptcies']=list(map(lambda x:x if x in z else 1,kisan2['pub r
         kisan2['pub rec bankruptcies'].value counts()
Out[73]:
              77846
         0
              10895
         Name: pub rec bankruptcies, dtype: int64
In [86]: kisan3=kisan2.dropna()
         kisan3.shape
Out[86]: (83135, 32)
```

```
In [87]: kisan3.isnull().sum()
Out[87]: loan_amnt
                                          0
                                          0
          term
          int_rate
                                          0
          installment
                                          0
                                          0
          grade
          sub_grade
                                          0
                                          0
          emp title
          emp length
                                          0
                                          0
          annual_inc
          verification_status
                                          0
                                          0
          issue_d
          loan_status
                                          0
          title
                                          0
          dti
                                          0
          earliest_cr_line
                                          0
          open acc
                                          0
          pub_rec
                                          0
          revol_bal
                                          0
          revol_util
                                          0
          total acc
                                          0
          initial_list_status
                                          0
          application_type
                                          0
          mort_acc
                                          0
          pub_rec_bankruptcies
                                          0
          address
                                          0
          is 36mnths
                                          0
          is_verified
                                          0
          mod_grade
                                          0
          home_ownership_OTHERS
                                          0
          home ownership RENT
                                          0
                                          0
          purpose debt consolidation
          purpose others
                                          0
          dtype: int64
```

Splitting the Kisan3 dataset into train, validation and test sets

```
In [91]: XS.info()
```

<class 'pandas.core.frame.DataFrame'>
Int64Index: 83135 entries, 0 to 200488
Data columns (total 21 columns):

```
Column
 #
                                Non-Null Count Dtype
     _____
                                -----
 0
    loan amnt
                                83135 non-null float64
 1
    int rate
                                83135 non-null float64
 2
    installment
                                83135 non-null float64
 3
    mod_grade
                                83135 non-null int64
 4
    sub grade
                                83135 non-null object
 5
    annual inc
                                83135 non-null float64
 6
    is verified
                                83135 non-null int64
 7
    dti
                                83135 non-null float64
 8
    open acc
                                83135 non-null int64
                                83135 non-null int64
 9
    pub rec
 10 revol_bal
                                83135 non-null float64
 11
    revol util
                                83135 non-null float64
 12
    total acc
                                83135 non-null
                                                float64
 13 initial list status
                                83135 non-null
                                               int64
 14
    mort acc
                                83135 non-null int64
 15
    pub_rec_bankruptcies
                                83135 non-null int64
 16
    is_36mnths
                                83135 non-null int64
 17
    home ownership OTHERS
                                83135 non-null int64
 18 home ownership RENT
                                83135 non-null int64
    purpose_debt_consolidation 83135 non-null
 19
                                                int64
                                83135 non-null int64
 20 purpose others
dtypes: float64(8), int64(12), object(1)
memory usage: 14.0+ MB
```

```
In [92]: y=kisan3['loan_status']
```

In [96]: kisan3['y_obs']=list(map(lambda x:0 if x=='Fully Paid' else 1,kisan3['loan_status'
kisan3[['loan_status','y_obs']].head(14)

C:\Users\Atul kumar\anaconda3\lib\site-packages\ipykernel_launcher.py:1: Settin
gWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame. Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy (https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)

"""Entry point for launching an IPython kernel.

Out[96]:

	loan_status	y_obs
0	Fully Paid	0
2	Fully Paid	0
4	Fully Paid	0
6	Fully Paid	0
8	Charged Off	1
10	Fully Paid	0
12	Fully Paid	0
14	Fully Paid	0
16	Fully Paid	0
18	Fully Paid	0
20	Fully Paid	0
22	Fully Paid	0
24	Fully Paid	0
26	Fully Paid	0

In [97]: y=kisan3['y_obs']
In [98]: from sklearn.model_selection import train_test_split
 X_train,X_test,y_train,y_test=train_test_split(XS,y,test_size=0.15,random_state=2)
In [99]: X_trainn,X_dev,y_trainn,y_dev=train_test_split(X_train,y_train,test_size=0.15,random_state=2)
In [99]: X_trainn,X_dev,y_trainn,y_test_size=2)
In [99]: X_trainn,X_trainn,Y_train_test_split(X_train,y_train,test_size=2)
In [99]: X_trainn,X_train_test_split(X_train,y_train,test_size=3)
In [99]: X_trainn,X_train_test_split(X_train,y_train,test_size=3)
In [99]: X_train_test_split(X_train,y_train,test_size=3)
In [99]: X_train_test_split(X_train,y_train,test_size=3)
In [99]: I

```
In [100]: print(X_trainn.shape)
          print(X_dev.shape)
          print(X_test.shape)
          print(y_trainn.shape)
          print(y_dev.shape)
          print(y_test.shape)
           (60064, 21)
           (10600, 21)
           (12471, 21)
           (60064,)
           (10600,)
           (12471,)
In [101]: X_trainn=pd.concat([X_trainn,y_trainn],axis=1)
In [103]: | z=X_trainn.groupby('sub_grade')['y_obs'].mean().to_dict()
          X_trainn['sub_grade']=list(map(lambda x:z[x],X_trainn['sub_grade']))
          X_trainn['sub_grade']
Out[103]: 14488
                     0.192450
          94254
                     0.154802
          24670
                     0.341498
                     0.168598
          20048
          55286
                     0.075792
          71550
                     0.108300
          86534
                     0.347036
          117018
                     0.085448
          190226
                     0.213446
          184816
                     0.125066
          Name: sub_grade, Length: 60064, dtype: float64
```

```
In [104]: X_dev['sub_grade']=list(map(lambda x:z[x],X_dev['sub_grade']))
X_dev['sub_grade']
```

C:\Users\Atul kumar\anaconda3\lib\site-packages\ipykernel_launcher.py:1: Settin
gWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame.

Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy (https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)

"""Entry point for launching an IPython kernel.

```
Out[104]:
                     0.404795
          39684
           16006
                     0.122754
           165846
                     0.154802
                     0.062593
           151720
           15474
                     0.192450
                       . . .
           192714
                     0.154802
           38598
                     0.122754
           132708
                     0.028833
           90502
                     0.241294
           179298
                     0.322265
           Name: sub grade, Length: 10600, dtype: float64
```

```
In [105]: X_test['sub_grade']=list(map(lambda x:z[x],X_test['sub_grade']))
X_test['sub_grade']
```

C:\Users\Atul kumar\anaconda3\lib\site-packages\ipykernel_launcher.py:1: Settin
gWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame.

Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy (https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)

"""Entry point for launching an IPython kernel.

```
Out[105]: 56938
                     0.322265
          63494
                     0.154802
                     0.240930
          120154
          26788
                     0.265314
          77820
                     0.108300
          143328
                     0.125066
          177034
                     0.265314
          48176
                     0.240930
          49746
                     0.101064
          110436
                     0.154802
```

Name: sub_grade, Length: 12471, dtype: float64

{'loan_amnt': [14491.336324587108, 8388.981997556977], 'int_rate': [13.78460392 2482301, 4.490693801881828], 'installment': [444.271407332167, 251.863757655670 16], 'annual_inc': [75394.66977274243, 48671.130821498744], 'dti': [17.68224127 597239, 8.108662503973125], 'open_acc': [11.59971030900373, 5.213968658647141], 'revol_bal': [16033.318560202451, 19607.84551585026], 'revol_util': [54.1856752 7970166, 23.89877657577768], 'total_acc': [25.88725359616409, 11.92963296750360 4], 'mort_acc': [1.8127164358018113, 2.148206675715537], 'mod_grade': [1.852124 400639318, 1.321668894368757]}

Out[108]:

	loan_amnt	int_rate	installment	mod_grade	sub_grade	annual_inc	is_verified	d
14488	-0.869156	0.339679	-0.768596	0.111886	0.192450	-0.735439	0	0.55468
94254	0.418247	0.068006	0.681792	-0.644734	0.154802	0.053529	0	-0.05330
24670	0.060635	0.907520	-0.256136	0.868505	0.341498	-0.110839	1	0.43012
20048	-0.296977	-0.176945	-0.158822	0.111886	0.168598	-0.234116	0	0.85436
55286	1.610286	-1.310400	1.714652	-1.401353	0.075792	1.347931	1	-0.07057
138636	0.418247	0.442113	0.740554	0.868505	0.265314	-0.521760	1	0.63114
47806	-1.155246	-1.595434	-1.178778	-1.401353	0.040411	-0.850497	0	-1.59363
151938	-1.369813	-1.726816	-1.401398	-1.401353	0.028833	-0.727221	0	0.99372
153512	0.626854	-0.150223	0.882376	-0.644734	0.125066	-0.460122	1	-0.36901
151486	-1.012201	-1.582073	-1.031833	-1.401353	0.062593	0.505543	0	-0.58483
106528	2.444714	1.381835	1.917023	1.625124	0.391566	0.711003	1	0.50535
167006	-0.589027	0.713786	-0.412252	0.868505	0.298160	-0.932682	1	1.51415
129208	0.418247	0.994367	0.060622	0.868505	0.268987	-0.460122	1	0.56825
162968	-0.907898	1.604517	-0.735641	1.625124	0.391566	-1.035412	1	1.06401
163026	-0.726112	-0.319462	-0.650595	-0.644734	0.125066	0.074075	1	-0.53674
77624	1.610286	1.381835	1.180831	0.868505	0.322265	0.197352	1	0.20074
17694	-0.654589	0.402031	-0.514847	0.111886	0.241294	-0.809405	1	1.41302
20454	0.954665	-0.266018	1.228476	0.111886	0.192450	-0.521760	1	-1.61829
23510	-0.535385	-0.522103	-0.455768	-0.644734	0.125066	0.505543	1	-1.25695
161670	-0.535385	-0.176945	-0.426307	0.111886	0.168598	-0.829951	1	1.53635

20 rows × 22 columns

```
cols=['loan amnt','int rate','installment','annual inc','dti',
                  open_acc','revol_bal','revol_util','total_acc','mort_acc',
                'mod grade']
           for i in cols:
               X_{dev[i]=(X_{dev[i]-d[i][0])/d[i][1]}
           print(X dev)
                   loan amnt
                               int_rate
                                          installment
                                                        mod_grade
                                                                    sub grade
                                                                                annual_inc
                                                                     0.404795
                                                                                  7.285743
           39684
                    2.444714
                               1.929634
                                             2.109746
                                                         1.625124
           16006
                   -1.310211 -0.733651
                                            -1.312302
                                                        -0.644734
                                                                     0.122754
                                                                                 -0.398484
           165846
                    0.060635 -0.399627
                                             0.213920
                                                        -0.644734
                                                                     0.154802
                                                                                 -0.213570
                    1.610286 -1.372751
                                             1.700318
                                                        -1.401353
                                                                     0.062593
                                                                                  1.841037
           151720
           15474
                   -0.177773 -0.176945
                                            -0.589769
                                                         0.111886
                                                                     0.192450
                                                                                  0.197352
           . . .
                   -0.296977 -0.502061
                                                                                 -0.727221
           192714
                                            -0.715353
                                                        -0.644734
                                                                     0.154802
           38598
                   -0.714191 -0.399627
                                            -0.643131
                                                        -0.644734
                                                                     0.122754
                                                                                 -0.419030
           132708
                   -0.535385 -1.884921
                                            -0.568249
                                                        -1.401353
                                                                     0.028833
                                                                                 -0.012218
           90502
                    1.133471
                               0.667023
                                             0.593013
                                                         0.111886
                                                                     0.241294
                                                                                  0.300082
                    1.783132 0.842942
                                             1.178012
                                                         0.868505
                                                                     0.322265
                                                                                 -0.336846
           179298
                   is verified
                                       dti
                                            open acc
                                                       pub rec
                                                                      revol util total acc
                                                                . . .
           39684
                              1 -0.561405
                                            0.268565
                                                             0
                                                                        1.172207
                                                                                    1.434474
                                                                 . . .
           16006
                              1 -0.758725
                                                             0
                                                                 . . .
                                                                       -1.601993
                                                                                   -0.912623
                                            0.652150
                                                             0
           165846
                              0
                                 0.622514
                                            0.076773
                                                                       -0.288118
                                                                                    0.512400
                                                                 . . .
           151720
                              1 -0.996742
                                            1.035735
                                                                       -0.555914
                                                                                    1.182999
                                                                 . . .
                                                                                    ~ ~ ~ ~ ~ ~ ~
          cols=['loan_amnt','int_rate','installment','annual_inc','dti',
In [110]:
                  open_acc', 'revol_bal', 'revol_util', 'total_acc', 'mort_acc',
                'mod_grade']
           for i in cols:
               X_{\text{test}[i]=(X_{\text{test}[i]-d[i][0])/d[i][1]}
           print(X_test)
                   loan amnt
                               int rate
                                          installment
                                                        mod grade
                                                                    sub grade
                                                                                annual inc
                    1.875515
                                                                                 -0.069747
           56938
                               1.210369
                                             2.648371
                                                         0.868505
                                                                     0.322265
           63494
                    1.133471 -0.502061
                                             1.379709
                                                        -0.644734
                                                                     0.154802
                                                                                 -0.143230
                    0.179839
                               0.339679
                                            -0.242279
                                                         0.111886
                                                                     0.240930
                                                                                 -0.316300
           120154
                   -0.392340
           26788
                               0.406484
                                            -0.691729
                                                         0.868505
                                                                     0.265314
                                                                                 -0.710784
           77820
                    0.060635 -0.588908
                                             0.189819
                                                        -0.644734
                                                                     0.108300
                                                                                  0.094621
           . . .
           143328
                   -0.535385 -0.176945
                                            -0.426307
                                                        -0.644734
                                                                     0.125066
                                                                                 -0.521760
           177034
                   -0.296977
                               0.406484
                                            -0.615140
                                                         0.868505
                                                                     0.265314
                                                                                 -0.768313
           48176
                   -0.773793
                               0.045738
                                            -0.678468
                                                         0.111886
                                                                     0.240930
                                                                                 -0.747767
           49746
                   -0.535385 -0.844993
                                            -0.482965
                                                        -0.644734
                                                                     0.101064
                                                                                  1.121924
           110436
                   -0.010888
                               0.068006
                                             0.192638
                                                        -0.644734
                                                                     0.154802
                                                                                 -0.656337
                   is verified
                                                                      revol util total acc
                                       dti
                                            open acc
                                                       pub rec ...
           \
           56938
                              1
                                 0.951792 -0.882190
                                                             0
                                                                        1.327027
                                                                                   -0.912623
                                                                 . . .
           63494
                                 0.967824
                                                                       -1.003636
                              1
                                            1.227527
                                                             0
                                                                 . . .
                                                                                    0.847700
           120154
                              1 -1.043605 -0.690397
                                                             0
                                                                 . . .
                                                                        0.732854
                                                                                   -0.158199
                                 1.932225
                                                             0
                                                                        0.448321
                                                                                   -0.912623
           26788
                              1
                                            0.268565
                                                                 . . .
```

Making a Logistic regression model

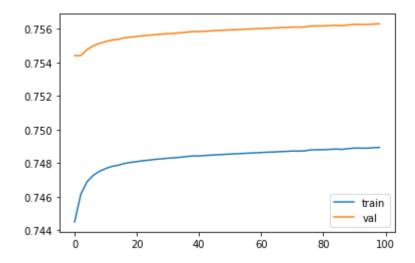
```
In [121]: train cost=[]
          val cost=[]
          lambdaa=[]
          for i in np.arange(0.01,100,2):
              lambdaa.append(i)
              model=LogisticRegression(C=1/i,class_weight={0:0.1,1:0.6})
              model.fit(X trainn,y trainn)
              y train probs=model.predict proba(X trainn)
              y val probs=model.predict proba(X dev)
              train_loss=metrics.log_loss(y_trainn,y_train_probs)
              val loss=metrics.log loss(y dev,y val probs)
              train cost.append(train loss)
              val cost.append(val loss)
          plt.plot(lambdaa,train cost,label='train')
          plt.plot(lambdaa,val cost,label='val')
          plt.legend()
          plt.show()
```

C:\Users\Atul kumar\anaconda3\lib\site-packages\sklearn\linear_model_logistic.
py:940: ConvergenceWarning: lbfgs failed to converge (status=1):
STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.

Increase the number of iterations (max_iter) or scale the data as shown in:
 https://scikit-learn.org/stable/modules/preprocessing.html (https://scikit-learn.org/stable/modules/preprocessing.html)

Please also refer to the documentation for alternative solver options:
 https://scikit-learn.org/stable/modules/linear_model.html#logistic-regressi
on (https://scikit-learn.org/stable/modules/linear_model.html#logistic-regressi
on)

extra_warning_msg=_LOGISTIC_SOLVER_CONVERGENCE_MSG)



```
In [122]: z=np.argmin(val_cost)
z
```

Out[122]: 0

```
In [123]: |model=LogisticRegression(C=1,class weight={0:0.1,1:0.6})
          model.fit(X trainn,y trainn)
          print(model.score(X trainn,y trainn))
          print(model.score(X dev,y dev))
          print(model.score(X_test,y_test))
          0.5457844965370272
          0.5323584905660378
          0.5452650148344158
In [124]: from sklearn.metrics import confusion matrix, precision score, recall score
          ypred=model.predict(X test)
          confusion_matrix(y_test,ypred)
Out[124]: array([[4788, 5176],
                  [ 495, 2012]], dtype=int64)
In [125]: |print(precision_score(y_test,ypred))
          print(recall_score(y_test,ypred))
          0.27991096271563715
          0.8025528520143598
          Removing the outliers which are on the wrong side of the plane theta T*x+theta not
In [126]: | z=np.array(list(model.intercept )+list(model.coef [0]))
          z=z.T
          Z
Out[126]: array([-0.37986903, -0.09505128, -0.24157436, 0.16026868,
                                                                       0.29393671,
                   3.91784125, -0.07684596, 0.10735067, 0.21085049, 0.10616443,
                  0.08805733, -0.07201472, 0.08561362, -0.1274847, -0.0507649,
                  -0.07308559, -0.08033209, -0.49011554, 0.10939622, 0.26325411,
                  0.02387176, 0.12101527])
In [127]: X trainn.insert(0,'x not',1)
```

In [128]: X_trainn

Out[128]:

	x_not	loan_amnt	int_rate	installment	mod_grade	sub_grade	annual_inc	is_verified
14488	1	-0.869156	0.339679	-0.768596	0.111886	0.192450	-0.735439	0
94254	1	0.418247	0.068006	0.681792	-0.644734	0.154802	0.053529	0 -
24670	1	0.060635	0.907520	-0.256136	0.868505	0.341498	-0.110839	1
20048	1	-0.296977	-0.176945	-0.158822	0.111886	0.168598	-0.234116	0
55286	1	1.610286	-1.310400	1.714652	-1.401353	0.075792	1.347931	1 -
71550	1	0.656655	-0.956334	0.779384	-0.644734	0.108300	-0.117003	1
86534	1	0.299043	1.159152	-0.013386	1.625124	0.347036	-0.291645	1
117018	1	0.662615	-1.087717	0.763860	-1.401353	0.085448	-0.624491	1
190226	1	-1.369813	0.266194	-1.351133	0.111886	0.213446	0.197352	0 -
184816	1	-1.143325	-0.622310	-1.127083	-0.644734	0.125066	-0.172477	1

60064 rows × 22 columns

```
In [129]: z2=X_trainn.dot(z)
          z3=1/(1+np.exp(-z2))
          z3
Out[129]: 14488
                    0.555917
          94254
                    0.504452
          24670
                    0.769651
          20048
                    0.539214
          55286
                    0.339882
          71550
                    0.491943
          86534
                    0.854377
          117018
                    0.431970
          190226
                    0.471579
                    0.391628
          184816
          Length: 60064, dtype: float64
In [130]: X_trainn['log_loss']=-1*(y_trainn*np.log(z3)+(1-y_trainn)*(1-np.log(z3)))
```

```
In [131]: X trainn['log loss'].describe()
Out[131]: count
                    60064.000000
                       -1.268352
          mean
                        0.921445
          std
                       -5.582546
          min
          25%
                       -1.851313
          50%
                       -1.563946
          75%
                       -1.232072
                        2.381357
          max
          Name: log_loss, dtype: float64
In [132]: X trainn=pd.concat([X trainn,y trainn],axis=1)
In [133]: |q05=X_trainn['log_loss'].quantile(0.05)
          q95=X_trainn['log_loss'].quantile(0.95)
          X_trainn=X_trainn[(X_trainn['log_loss']>q05)&(X_trainn['log_loss']<q95)]</pre>
In [134]: y_trainn=X_trainn['y_obs']
          y trainn
Out[134]: 14488
                     0
          94254
                     0
          24670
                     0
          20048
          55286
                     0
          71550
                     0
          86534
                     1
          117018
          190226
                     0
          184816
          Name: y_obs, Length: 54056, dtype: int64
In [136]: X_trainn=X_trainn.drop(['x_not','y_obs','log_loss'],axis=1)
```

Performing LogisticRegression after removing outliers

```
In [173]: train cost=[]
          val cost=[]
          lambdaa=[]
          for i in np.arange(0.1,1000,5):
              lambdaa.append(i)
              model=LogisticRegression(C=1/i,class_weight={0:0.1,1:0.9})
              model.fit(X trainn,y trainn)
              y train probs=model.predict proba(X trainn)
              y val probs=model.predict proba(X dev)
              train_loss=metrics.log_loss(y_trainn,y_train_probs)
              val loss=metrics.log loss(y dev,y val probs)
              train cost.append(train loss)
              val cost.append(val loss)
          plt.plot(lambdaa,train cost,label='train')
          plt.plot(lambdaa,val cost,label='val')
          plt.legend()
          plt.show()
```

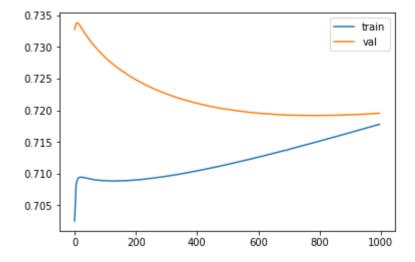
C:\Users\Atul kumar\anaconda3\lib\site-packages\sklearn\linear_model_logistic.
py:940: ConvergenceWarning: lbfgs failed to converge (status=1):
STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.

Increase the number of iterations (max_iter) or scale the data as shown in:
 https://scikit-learn.org/stable/modules/preprocessing.html (https://scikit-learn.org/stable/modules/preprocessing.html)

Please also refer to the documentation for alternative solver options:

https://scikit-learn.org/stable/modules/linear_model.html#logistic-regressi
on (https://scikit-learn.org/stable/modules/linear_model.html#logistic-regressi
on)

extra warning msg= LOGISTIC SOLVER CONVERGENCE MSG)



```
In [ ]:
In [174]: | z=np.argmin(val_cost)
          lbest=0.1+z*5
          lbest
Out[174]: 785.1
In [175]: | model=LogisticRegression(C=1/lbest,class_weight={0:0.1,1:0.9})
          model.fit(X_trainn,y_trainn)
          print(model.score(X_trainn,y_trainn))
          print(model.score(X_dev,y_dev))
          print(model.score(X_test,y_test))
          0.6240565339647772
          0.6033962264150944
          0.6118194210568519
In [178]: ypred=model.predict(X_test)
          confusion_matrix(y_test,ypred)
Out[178]: array([[5850, 4114],
                  [ 727, 1780]], dtype=int64)
In [177]: print(precision score(y test,ypred))
          print(recall_score(y_test,ypred))
          0.30200203596878183
          0.7100119664938173
```

Introducing polynomial terms in the logistic regression model

```
In [180]: cols=['loan amnt','int rate','installment','annual inc','dti',
                   open_acc','revol_bal','revol_util','total_acc','mort_acc',
                 'mod grade']
           t=0
           for i in cols:
                z='f'+str(t)
               XS[z]=XS[i]**2
               t+=1
           print(XS)
                    loan amnt
                                int rate
                                           installment
                                                          mod grade sub grade
                                                                                 annual inc
           0
                      10000.0
                                    11.44
                                                 329.48
                                                                             B4
                                                                                    117000.0
                                                                   1
           2
                       8000.0
                                    11.99
                                                 265.68
                                                                   1
                                                                             B5
                                                                                     65000.0
           4
                      15600.0
                                    10.49
                                                 506.97
                                                                   1
                                                                             В3
                                                                                     43057.0
           6
                                     6.49
                                                                   0
                       7200.0
                                                 220.65
                                                                             Α2
                                                                                     54000.0
           8
                      24375.0
                                    17.27
                                                 609.33
                                                                   2
                                                                             C5
                                                                                     55000.0
                                                     . . .
                           . . .
                                      . . .
                                                                 . . .
                                                                            . . .
           200476
                      30225.0
                                    17.27
                                                 755.57
                                                                   2
                                                                             C5
                                                                                     68000.0
           200478
                                     5.32
                                                 180.69
                                                                   0
                                                                             Α1
                       6000.0
                                                                                     48000.0
           200480
                      15450.0
                                     7.90
                                                 483.44
                                                                   0
                                                                             Α4
                                                                                     42000.0
           200486
                       5000.0
                                    13.53
                                                 169.75
                                                                   1
                                                                             B5
                                                                                     43000.0
                                                                   2
                                    14.30
                                                                             C1
                                                                                     44000.0
           200488
                      15000.0
                                                 514.86
                    is_verified
                                     dti
                                          open_acc
                                                     pub_rec
                                                                            f1
                                                                                          f2
                                                                                               \
           0
                                  26.24
                                                                     130.8736
                                                                                108557.0704
                               0
                                                 16
                                                            0
                                                                . . .
           2
                               0
                                  22.05
                                                 17
                                                            0
                                                                     143.7601
                                                                                 70585.8624
                                                                . . .
                                  12.79
           4
                               1
                                                 13
                                                                     110.0401
                                                                                257018.5809
                                                            0
           6
                               0
                                   2.60
                                                  6
                                                            0
                                                                . . .
                                                                      42.1201
                                                                                 48686.4225
                                  33.95
                                                 13
                                                                                371283.0489
           8
                               1
                                                            0
                                                                     298.2529
                                  24.55
           200476
                               1
                                                  7
                                                            0
                                                                     298.2529
                                                                                570886.0249
                                                                . . .
                                                  7
           200478
                               1
                                    3.78
                                                            0
                                                                      28.3024
                                                                                 32648.8761
           200480
                                  28.09
                                                 13
                                                            0
                               0
                                                                . . .
                                                                      62.4100
                                                                                233714.2336
           200486
                                  18.39
                                                            0
                                                                     183.0609
                                                                                 28815.0625
                               1
                                                  6
           200488
                               1
                                  26.02
                                                 10
                                                                     204.4900
                                                                                265080.8196
                                                            1
                                                                . . .
                               f3
                                           f4
                                                 f5
                                                                 f6
                                                                           f7
                                                                                   f8
                                                                                        f9
                                                                                            f10
           0
                    1.368900e+10
                                     688.5376
                                                256
                                                     1.322704e+09
                                                                     1747.24
                                                                                625.0
                                                                                         0
                                                                                               1
           2
                    4.225000e+09
                                     486.2025
                                                289
                                                     4.052572e+08
                                                                     2840.89
                                                                                729.0
                                                                                         9
                                                                                               1
           4
                                     163.5841
                                                169
                                                                                         0
                                                                                               1
                    1.853905e+09
                                                     1.436882e+08
                                                                     8500.84
                                                                                676.0
           6
                    2.916000e+09
                                       6.7600
                                                 36
                                                     2.994278e+07
                                                                      462.25
                                                                                169.0
                                                                                         0
                                                                                               0
           8
                    3.025000e+09
                                    1152.6025
                                                169
                                                     6.043731e+08
                                                                     4872.04
                                                                               1849.0
                                                                                         1
                                                                                               4
                                                                                   . . .
                                                . . .
           200476
                    4.624000e+09
                                     602.7025
                                                 49
                                                     1.494905e+09
                                                                     7072.81
                                                                                289.0
                                                                                        16
                                                                                               4
           200478
                    2.304000e+09
                                      14.2884
                                                 49
                                                     5.346534e+07
                                                                     1705.69
                                                                                         0
                                                                                               0
                                                                                841.0
           200480
                    1.764000e+09
                                     789.0481
                                                169
                                                     2.196028e+08
                                                                     3203.56
                                                                                729.0
                                                                                         0
                                                                                               0
           200486
                    1.849000e+09
                                     338.1921
                                                 36
                                                     5.290000e+06
                                                                     1747.24
                                                                                289.0
                                                                                               1
                                                                                        16
                                     677.0404
                                                                                               4
           200488
                    1.936000e+09
                                                100
                                                     1.937107e+08
                                                                     3757.69
                                                                                529.0
                                                                                         0
```

[83135 rows x 32 columns]

C:\Users\Atul kumar\anaconda3\lib\site-packages\ipykernel_launcher.py:7: Sett
ingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame. Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/s table/user_guide/indexing.html#returning-a-view-versus-a-copy (https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)

import sys

```
In [183]: XS.info()
```

<class 'pandas.core.frame.DataFrame'>
Int64Index: 83135 entries, 0 to 200488
Data columns (total 32 columns):

#	Column	Non-Null Count	Dtype
0	loan_amnt	83135 non-null	
1	int_rate	83135 non-null	float64
2	installment	83135 non-null	float64
3	mod_grade	83135 non-null	int64
4	sub_grade	83135 non-null	object
5	annual_inc	83135 non-null	float64
6	is_verified	83135 non-null	int64
7	dti	83135 non-null	float64
8	open_acc	83135 non-null	int64
9	pub_rec	83135 non-null	int64
10	revol_bal	83135 non-null	float64
11	revol_util	83135 non-null	float64
12	total_acc	83135 non-null	float64
13	initial_list_status	83135 non-null	int64
14	mort_acc	83135 non-null	int64
15	<pre>pub_rec_bankruptcies</pre>	83135 non-null	int64
16	is_36mnths	83135 non-null	int64
17	home_ownership_OTHERS	83135 non-null	int64
18	home_ownership_RENT	83135 non-null	int64
19	<pre>purpose_debt_consolidation</pre>	83135 non-null	int64
20	purpose_others	83135 non-null	int64
21	f0	83135 non-null	float64
22	f1	83135 non-null	float64
23	f2	83135 non-null	float64
24	f3	83135 non-null	float64
25	f4	83135 non-null	float64
26	f5	83135 non-null	int64
27	f6	83135 non-null	float64
28	f7	83135 non-null	float64
29	f8	83135 non-null	float64
30	f9	83135 non-null	int64
31	f10	83135 non-null	int64
dtype	es: float64(16), int64(15),	object(1)	

```
In [184]: X_train,X_test,y_train,y_test=train_test_split(XS,y,test_size=0.15,random_state=2
```

```
In [185]: X_trainn,X_dev,y_trainn,y_dev=train_test_split(X_train,y_train,test_size=0.15,rar
```

memory usage: 20.9+ MB

```
In [186]: |print(X_trainn.shape)
          print(X_dev.shape)
          print(X_test.shape)
          print(y_trainn.shape)
          print(y_dev.shape)
          print(y_test.shape)
           (60064, 32)
           (10600, 32)
           (12471, 32)
           (60064,)
           (10600,)
           (12471,)
In [187]: X_trainn=pd.concat([X_trainn,y_trainn],axis=1)
In [188]: | z=X_trainn.groupby('sub_grade')['y_obs'].mean().to_dict()
          X_trainn['sub_grade']=list(map(lambda x:z[x],X_trainn['sub_grade']))
          X_trainn['sub_grade']
Out[188]: 14488
                     0.192450
          94254
                     0.154802
          24670
                     0.341498
          20048
                     0.168598
          55286
                     0.075792
          71550
                     0.108300
          86534
                     0.347036
          117018
                     0.085448
          190226
                     0.213446
                     0.125066
          184816
          Name: sub_grade, Length: 60064, dtype: float64
```

```
In [189]: X_dev['sub_grade']=list(map(lambda x:z[x],X_dev['sub_grade']))
X_dev['sub_grade']
```

C:\Users\Atul kumar\anaconda3\lib\site-packages\ipykernel_launcher.py:1: Settin
gWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame.

Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy (https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)

"""Entry point for launching an IPython kernel.

```
Out[189]:
                     0.404795
          39684
           16006
                     0.122754
           165846
                     0.154802
                     0.062593
           151720
           15474
                     0.192450
                       . . .
           192714
                     0.154802
           38598
                     0.122754
           132708
                     0.028833
           90502
                     0.241294
           179298
                     0.322265
           Name: sub grade, Length: 10600, dtype: float64
```

```
In [190]: X_test['sub_grade']=list(map(lambda x:z[x],X_test['sub_grade']))
X_test['sub_grade']
```

C:\Users\Atul kumar\anaconda3\lib\site-packages\ipykernel_launcher.py:1: Settin
gWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame.

Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy (https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)

"""Entry point for launching an IPython kernel.

```
Out[190]:
          56938
                     0.322265
          63494
                     0.154802
          120154
                     0.240930
          26788
                     0.265314
          77820
                     0.108300
          143328
                     0.125066
          177034
                     0.265314
                     0.240930
          48176
          49746
                     0.101064
          110436
                     0.154802
          Name: sub_grade, Length: 12471, dtype: float64
```

{'loan_amnt': [14491.336324587108, 8388.981997556977], 'int_rate': [13.78460392 2482301, 4.490693801881828], 'installment': [444.271407332167, 251.863757655670 16], 'annual_inc': [75394.66977274243, 48671.130821498744], 'dti': [17.68224127 597239, 8.108662503973125], 'open_acc': [11.59971030900373, 5.213968658647141], 'revol_bal': [16033.318560202451, 19607.84551585026], 'revol_util': [54.1856752 7970166, 23.89877657577768], 'total_acc': [25.88725359616409, 11.92963296750360 4], 'mort_acc': [1.8127164358018113, 2.148206675715537], 'mod_grade': [1.852124 400639318, 1.321668894368757]}

Out[192]:

	loan_amnt	int_rate	installment	mod_grade	sub_grade	annual_inc	is_verified	d
14488	-0.869156	0.339679	-0.768596	0.111886	0.192450	-0.735439	0	0.55468
94254	0.418247	0.068006	0.681792	-0.644734	0.154802	0.053529	0	-0.05330
24670	0.060635	0.907520	-0.256136	0.868505	0.341498	-0.110839	1	0.43012
20048	-0.296977	-0.176945	-0.158822	0.111886	0.168598	-0.234116	0	0.85436
55286	1.610286	-1.310400	1.714652	-1.401353	0.075792	1.347931	1	-0.07057
138636	0.418247	0.442113	0.740554	0.868505	0.265314	-0.521760	1	0.63114
47806	-1.155246	-1.595434	-1.178778	-1.401353	0.040411	-0.850497	0	-1.59363
151938	-1.369813	-1.726816	-1.401398	-1.401353	0.028833	-0.727221	0	0.99372
153512	0.626854	-0.150223	0.882376	-0.644734	0.125066	-0.460122	1	-0.36901
151486	-1.012201	-1.582073	-1.031833	-1.401353	0.062593	0.505543	0	-0.58483
106528	2.444714	1.381835	1.917023	1.625124	0.391566	0.711003	1	0.50535
167006	-0.589027	0.713786	-0.412252	0.868505	0.298160	-0.932682	1	1.51415
129208	0.418247	0.994367	0.060622	0.868505	0.268987	-0.460122	1	0.56825
162968	-0.907898	1.604517	-0.735641	1.625124	0.391566	-1.035412	1	1.06401
163026	-0.726112	-0.319462	-0.650595	-0.644734	0.125066	0.074075	1	-0.53674
77624	1.610286	1.381835	1.180831	0.868505	0.322265	0.197352	1	0.20074
17694	-0.654589	0.402031	-0.514847	0.111886	0.241294	-0.809405	1	1.41302
20454	0.954665	-0.266018	1.228476	0.111886	0.192450	-0.521760	1	-1.61829
23510	-0.535385	-0.522103	-0.455768	-0.644734	0.125066	0.505543	1	-1.25695
161670	-0.535385	-0.176945	-0.426307	0.111886	0.168598	-0.829951	1	1.53635

20 rows × 33 columns

```
In [193]: cols=['loan amnt','int rate','installment','annual inc','dti',
                 'open_acc','revol_bal','revol_util','total_acc','mort_acc',
                'mod grade']
          for i in cols:
               X_{dev[i]=(X_{dev[i]-d[i][0])/d[i][1]}
          print(X_dev)
                   loan_amnt int_rate installment mod_grade
                                                                  sub grade
                                                                              annual_inc
          39684
                                                                    0.404795
                    2.444714
                              1.929634
                                            2.109746
                                                        1.625124
                                                                                7.285743
                                                                    0.122754
          16006
                   -1.310211 -0.733651
                                           -1.312302
                                                       -0.644734
                                                                               -0.398484
          165846
                    0.060635 -0.399627
                                            0.213920
                                                       -0.644734
                                                                    0.154802
                                                                               -0.213570
          151720
                    1.610286 -1.372751
                                            1.700318
                                                      -1.401353
                                                                    0.062593
                                                                                1.841037
          15474
                   -0.177773 -0.176945
                                           -0.589769
                                                        0.111886
                                                                    0.192450
                                                                                0.197352
           . . .
                   -0.296977 -0.502061
                                           -0.715353
                                                                               -0.727221
          192714
                                                       -0.644734
                                                                    0.154802
          38598
                   -0.714191 -0.399627
                                           -0.643131
                                                       -0.644734
                                                                    0.122754
                                                                               -0.419030
          132708
                  -0.535385 -1.884921
                                           -0.568249
                                                       -1.401353
                                                                    0.028833
                                                                               -0.012218
          90502
                    1.133471 0.667023
                                            0.593013
                                                        0.111886
                                                                    0.241294
                                                                                0.300082
          179298
                    1.783132 0.842942
                                            1.178012
                                                        0.868505
                                                                    0.322265
                                                                               -0.336846
                   is verified
                                      dti
                                           open acc
                                                     pub rec
                                                                           f1
                                                                                         f2
                                                                                             \
                              1 -0.561405
                                                                    504.0025
          39684
                                           0.268565
                                                                               951873.4096
                                                            0
          16006
                              1 -0.758725
                                           0.652150
                                                            0
                                                                    110.0401
                                                                                12939.0625
                                                                . . .
                             0 0.622514
                                           0.076773
                                                            0
                                                                    143.7601
                                                                               248153.4225
          165846
                                                                . . .
                             1 -0.996742
                                                                               761291,1504
          151720
                                           1.035735
                                                            0
                                                                     58.0644
                                                                . . .
          15474
                             0 -0.737759 -1.073982
                                                            0
                                                                    168.7401
                                                                                87456.2329
                                                                . . .
           . . .
                                      . . .
                                                 . . .
                                                          . . .
                                                                . . .
                                                                          . . .
                                                                                        . . .
                            . . .
          192714
                             0
                                0.390664 -0.498605
                                                            0
                                                                    132.9409
                                                                                69748.8100
          38598
                             0 -0.192663
                                           1.035735
                                                            0
                                                                . . .
                                                                    143.7601
                                                                                79687.6441
          132708
                             1 -1.187895 -0.690397
                                                            1
                                                                     28.3024
                                                                                90691.3225
                                                                . . .
                                                                               352396.5769
          90502
                             1 1.224340
                                          0.076773
                                                            0
                                                               . . .
                                                                    281.5684
          179298
                                0.521388 -0.498605
                                                            0
                                                                     308.7049
                                                                               549036.5409
                              f3
                                        f4
                                             f5
                                                            f6
                                                                      f7
                                                                              f8
                                                                                  f9
                                                                                      f10
          39684
                   1.849000e+11
                                 172.3969
                                            169
                                                 2.655959e+09
                                                                6756.84
                                                                          1849.0
                                                                                   9
                                                                                        16
          16006
                   3.136000e+09
                                  132.9409
                                            225
                                                  1.227101e+07
                                                                 252.81
                                                                           225.0
                                                                                         1
                                                                          1024.0
          165846
                   4.225000e+09
                                  516.6529
                                            144
                                                  2.757260e+08
                                                                2237.29
                                                                                         1
                                                                                   0
                                            289
          151720
                   2.722500e+10
                                   92.1600
                                                 4.789970e+08
                                                                1672.81
                                                                          1600.0
                                                                                         0
          15474
                   7.225000e+09
                                  136.8900
                                             36
                                                 1.100191e+08
                                                                5227.29
                                                                           324.0
                                                                                   1
                                                                                         4
                                             . . .
                                                                          1296.0
          192714
                   1.600000e+09 434.7225
                                             81
                                                 1.029961e+09
                                                                4747.21
                                                                                   9
                                                                                         1
          38598
                   3.025000e+09
                                  259.8544
                                            289
                                                 9.329628e+07
                                                                1962.49
                                                                          1024.0
                                                                                   1
                                                                                         1
          132708
                   5.595040e+09
                                   64.8025
                                             64
                                                 1.651610e+07
                                                                2520.04
                                                                           144.0
                                                                                         0
          90502
                   8.100000e+09
                                  762.3121
                                            144
                                                 7.464917e+08
                                                                4692.25
                                                                           900.0
                                                                                   4
                                                                                         4
          179298
                   3.481000e+09
                                  480.0481
                                             81
                                                 3.013002e+08
                                                                3660.25
                                                                           289.0
                                                                                         9
```

[10600 rows x 32 columns]

C:\Users\Atul kumar\anaconda3\lib\site-packages\ipykernel_launcher.py:5: Settin
gWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame. Try using .loc[row indexer,col indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy (https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-c

opy)

```
In [194]: cols=['loan amnt','int rate','installment','annual inc','dti',
                  open_acc','revol_bal','revol_util','total_acc','mort_acc',
                 'mod grade']
           for i in cols:
               X_{\text{test}[i]=(X_{\text{test}[i]-d[i][0])/d[i][1]}
           print(X test)
                    loan_amnt int_rate
                                          installment
                                                        mod_grade
                                                                    sub_grade
                                                                                annual inc
           56938
                     1.875515
                               1.210369
                                                         0.868505
                                                                     0.322265
                                                                                 -0.069747
                                             2.648371
           63494
                    1.133471 -0.502061
                                             1.379709
                                                        -0.644734
                                                                     0.154802
                                                                                 -0.143230
           120154
                    0.179839
                               0.339679
                                             -0.242279
                                                         0.111886
                                                                     0.240930
                                                                                 -0.316300
           26788
                    -0.392340
                               0.406484
                                             -0.691729
                                                         0.868505
                                                                     0.265314
                                                                                 -0.710784
                                                                     0.108300
                                             0.189819
           77820
                    0.060635 -0.588908
                                                        -0.644734
                                                                                  0.094621
           . . .
                                            -0.426307
                                                                     0.125066
                                                                                 -0.521760
           143328
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                                                        -0.644734
           177034
                   -0.296977
                               0.406484
                                             -0.615140
                                                         0.868505
                                                                     0.265314
                                                                                 -0.768313
           48176
                    -0.773793
                               0.045738
                                             -0.678468
                                                         0.111886
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                                               49
                                                    3.330329e+09
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           63494
                   4.681778e+09
                                    651.7809
                                              324
                                                    1.038565e+08
                                                                    912.04
                                                                             1296.0
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                   3.600000e+09
                                     85.0084
                                                    2.815364e+07
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           26788
                                   1112.2225
                                              169
                                                    2.984256e+08
                                                                   4212.01
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                    1.664640e+09
           77820
                    6.400000e+09
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           143328
                   2.500000e+09
                                    394.0225
                                              169
                                                    3.053668e+07
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           177034
                   1.444000e+09
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                                                    9.672100e+06
                                                                   2034.01
                                                                              841.0
                                                                                            9
                                               64
           48176
                    1.521000e+09
                                   1104.2329
                                              256
                                                    5.406661e+07
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           49746
                    1.690000e+10
                                     48.1636
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                                                    1.314233e+08
                                                                   3080.25
           110436
                   1.887902e+09
                                  1206.8676
                                              169
                                                    3.661460e+07
                                                                   7259.04
                                                                             2116.0
                                                                                     81
                                                                                            1
           [12471 rows x 32 columns]
           C:\Users\Atul kumar\anaconda3\lib\site-packages\ipykernel launcher.py:5: Sett
           ingWithCopyWarning:
```

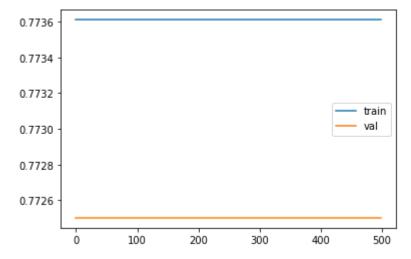
A value is trying to be set on a copy of a slice from a DataFrame. Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/s

table/user_guide/indexing.html#returning-a-view-versus-a-copy (https://panda s.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-ver sus-a-copy)

Making a LogisticRegression model

```
In [196]:
          train cost=[]
          val_cost=[]
          lambdaa=[]
          for i in np.arange(0.01,500,2):
              lambdaa.append(i)
              model=LogisticRegression(C=1/i,class weight={0:0.1,1:0.6})
              model.fit(X trainn,y trainn)
              y_train_probs=model.predict_proba(X_trainn)
              y val probs=model.predict proba(X dev)
              train_loss=metrics.log_loss(y_trainn,y_train_probs)
              val_loss=metrics.log_loss(y_dev,y_val_probs)
              train cost.append(train loss)
              val cost.append(val loss)
          plt.plot(lambdaa,train cost,label='train')
          plt.plot(lambdaa,val cost,label='val')
          plt.legend()
          plt.show()
```



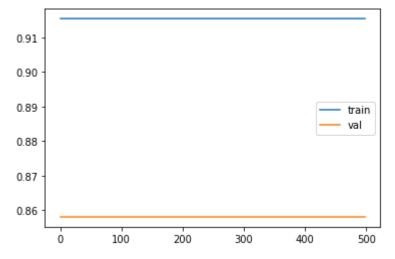
```
In [197]: | z=np.argmin(val cost)
Out[197]: 0
In [198]: |model=LogisticRegression(C=1,class weight={0:0.1,1:0.6})
          model.fit(X_trainn,y_trainn)
          print(model.score(X_trainn,y_trainn))
          print(model.score(X_dev,y_dev))
          print(model.score(X_test,y_test))
          0.2590570058604156
          0.25669811320754715
          0.26124609093095985
In [199]: | ypred=model.predict(X test)
          confusion_matrix(y_test,ypred)
Out[199]: array([[ 889, 9075],
                  [ 138, 2369]], dtype=int64)
In [200]: |print(precision_score(y_test,ypred))
          print(recall score(y test,ypred))
          0.20700803914715135
          0.944954128440367
          Removing the outliers which are on the wrong side of the plane theta T*x+theta not
In [201]:
          z=np.array(list(model.intercept_)+list(model.coef_[0]))
          z=z.T
          z
Out[201]: array([ 2.73464847e-18,
                                   9.53859513e-19,
                                                     5.42256037e-18,
                                                                       5.66917095e-19,
                   5.71174031e-18, 1.17718627e-18, -1.77511311e-18,
                                                                       2.65818229e-18,
                   2.89722012e-18, 3.63015482e-19, 4.71306960e-19, -4.59448671e-19,
                  1.59365616e-18, -7.35330550e-19,
                                                     1.11152473e-18, -1.78406958e-18,
                  3.26092317e-19, 4.97488431e-19, 2.77981799e-19, 1.88506989e-18,
                  1.91517594e-18, 5.73489833e-19,
                                                     9.90760756e-10, 1.29122727e-15,
                  7.95629367e-13, -4.02482995e-12,
                                                     1.95358871e-15,
                                                                      4.97865851e-16,
                  -1.74604483e-11, 1.33647727e-14,
                                                     1.74795612e-15,
                                                                       5.97045872e-19,
                  4.80592142e-17])
In [202]: X_trainn.insert(0,'x_not',1)
```

```
In [203]: | z2=X trainn.dot(z)
           z3=1/(1+np.exp(-z2))
           z3
Out[203]: 14488
                     0.510121
           94254
                     0.570642
           24670
                     0.549522
           20048
                     0.529288
           55286
                     0.660947
                       . . .
           71550
                     0.592478
           86534
                     0.566896
           117018
                     0.596300
           190226
                     0.492246
           184816
                     0.497507
           Length: 60064, dtype: float64
In [204]: | X_trainn['log_loss']=-1*(y_trainn*np.log(z3)+(1-y_trainn)*(1-np.log(z3)))
In [205]: X_trainn['log_loss'].describe()
Out[205]: count
                    60064.000000
                       -1.165980
           mean
           std
                        0.884166
           min
                      -25.048947
           25%
                       -1.665072
           50%
                       -1.598013
           75%
                       -1.367070
           max
                        4.028935
           Name: log loss, dtype: float64
In [206]: X trainn=pd.concat([X trainn,y trainn],axis=1)
In [207]: |q10=X_trainn['log_loss'].quantile(0.10)
           q90=X trainn['log loss'].quantile(0.90)
          X_trainn=X_trainn[(X_trainn['log_loss']>q10)&(X_trainn['log_loss']<q90)]</pre>
In [208]: X_trainn['log_loss'].describe()
Out[208]: count
                    48050.000000
                       -1.324091
           mean
           std
                        0.687728
                       -1.688692
           min
           25%
                       -1.655102
           50%
                       -1.598013
           75%
                       -1.457196
                        0.607496
           Name: log loss, dtype: float64
```

```
In [209]: |y_trainn=X_trainn['y_obs']
          y_trainn
Out[209]: 14488
                     0
           94254
                     0
           24670
                     0
           20048
                     0
           55286
           162398
                     0
           2676
                     0
           71550
                     0
           86534
                     1
           117018
           Name: y_obs, Length: 48050, dtype: int64
In [210]: X_trainn=X_trainn.drop(['x_not','y_obs','log_loss'],axis=1)
```

Performing LogisticRegression after removing outliers

```
In [221]:
          train_cost=[]
          val cost=[]
          lambdaa=[]
          for i in np.arange(0.1,500,2):
              lambdaa.append(i)
              model=LogisticRegression(C=1/i,class_weight={0:0.1,1:1.2})
              model.fit(X_trainn,y_trainn)
              y train probs=model.predict proba(X trainn)
              y val probs=model.predict proba(X dev)
              train_loss=metrics.log_loss(y_trainn,y_train_probs)
              val loss=metrics.log loss(y dev,y val probs)
              train_cost.append(train_loss)
              val cost.append(val loss)
          plt.plot(lambdaa,train cost,label='train')
          plt.plot(lambdaa,val cost,label='val')
          plt.legend()
          plt.show()
```



```
In [222]: z=np.argmin(val_cost)
          lbest=0.1+z*5
          1best
Out[222]: 240.1
In [223]: |model=LogisticRegression(C=1/lbest,class_weight={0:0.1,1:1.2})
          model.fit(X trainn,y trainn)
          print(model.score(X_trainn,y_trainn))
          print(model.score(X_dev,y_dev))
          print(model.score(X_test,y_test))
          0.24276795005202914
          0.36933962264150944
          0.36612942025499157
In [224]: | ypred=model.predict(X_test)
          confusion_matrix(y_test,ypred)
Out[224]: array([[2471, 7493],
                  [ 412, 2095]], dtype=int64)
In [225]: print(precision_score(y_test,ypred))
          print(recall_score(y_test,ypred))
          0.21850229453483522
          0.8356601515755884
```

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