In [3]:

import pandas as pd
import numpy as np
import seaborn as sns

In [4]:

df=pd.read_csv('datasets/loan_data.csv')
df

Out[4]:

	credit.policy	purpose	int.rate	installment	log.annual.inc	dti	fico	days.with
0	1	debt_consolidation	0.1189	829.10	11.350407	19.48	737	5639.
1	1	credit_card	0.1071	228.22	11.082143	14.29	707	2760.
2	1	debt_consolidation	0.1357	366.86	10.373491	11.63	682	4710.
3	1	debt_consolidation	0.1008	162.34	11.350407	8.10	712	2699.
4	1	credit_card	0.1426	102.92	11.299732	14.97	667	4066.
9573	0	all_other	0.1461	344.76	12.180755	10.39	672	10474.
9574	0	all_other	0.1253	257.70	11.141862	0.21	722	4380.
9575	0	debt_consolidation	0.1071	97.81	10.596635	13.09	687	3450.
9576	0	home_improvement	0.1600	351.58	10.819778	19.18	692	1800.
9577	0	debt_consolidation	0.1392	853.43	11.264464	16.28	732	4740.

9578 rows × 14 columns

→

In [5]:

```
df.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 9578 entries, 0 to 9577
Data columns (total 14 columns):
     Column
                        Non-Null Count
                                        Dtype
     credit.policy
                                        int64
 0
                        9578 non-null
 1
     purpose
                        9578 non-null
                                        object
 2
     int.rate
                        9578 non-null
                                        float64
 3
     installment
                        9578 non-null
                                        float64
 4
     log.annual.inc
                        9578 non-null
                                        float64
 5
     dti
                        9578 non-null
                                        float64
 6
     fico
                        9578 non-null
                                        int64
 7
     days.with.cr.line 9578 non-null
                                        float64
     revol.bal
                        9578 non-null
                                        int64
     revol.util
                                        float64
                        9578 non-null
 10 inq.last.6mths
                        9578 non-null
                                        int64
 11 deling.2yrs
                        9578 non-null
                                        int64
     pub.rec
                        9578 non-null
                                        int64
                        9578 non-null
     not.fully.paid
                                        int64
dtypes: float64(6), int64(7), object(1)
```

work with missing data

memory usage: 1.0+ MB

```
In [6]:
```

```
df.isnull().sum()
Out[6]:
credit.policy
                       0
                       0
purpose
int.rate
                       0
installment
log.annual.inc
                       0
dti
fico
                       0
days.with.cr.line
revol.bal
                       0
revol.util
                       0
                       0
inq.last.6mths
                       0
delinq.2yrs
pub.rec
                       0
not.fully.paid
dtype: int64
```

categorical data

In [7]:

df=pd.get_dummies(df,columns=['purpose'],drop_first=True)
df

Out[7]:

	credit.policy	int.rate	installment	log.annual.inc	dti	fico	days.with.cr.line	revol.bal	ı
0	1	0.1189	829.10	11.350407	19.48	737	5639.958333	28854	
1	1	0.1071	228.22	11.082143	14.29	707	2760.000000	33623	
2	1	0.1357	366.86	10.373491	11.63	682	4710.000000	3511	
3	1	0.1008	162.34	11.350407	8.10	712	2699.958333	33667	
4	1	0.1426	102.92	11.299732	14.97	667	4066.000000	4740	
9573	0	0.1461	344.76	12.180755	10.39	672	10474.000000	215372	
9574	0	0.1253	257.70	11.141862	0.21	722	4380.000000	184	
9575	0	0.1071	97.81	10.596635	13.09	687	3450.041667	10036	
9576	0	0.1600	351.58	10.819778	19.18	692	1800.000000	0	
9577	0	0.1392	853.43	11.264464	16.28	732	4740.000000	37879	

9578 rows × 19 columns

In [8]:

df.describe()

Out[8]:

fico	days.with.cr.line	revol.bal	revol.util	inq.last.6mths	delinq.2yrs	pub.rec
78.000000	9578.000000	9.578000e+03	9578.000000	9578.000000	9578.000000	9578.000000
10.846314	4560.767197	1.691396e+04	46.799236	1.577469	0.163708	0.062122
37.970537	2496.930377	3.375619e+04	29.014417	2.200245	0.546215	0.262126
12.000000	178.958333	0.000000e+00	0.000000	0.000000	0.000000	0.000000
82.000000	2820.000000	3.187000e+03	22.600000	0.000000	0.000000	0.000000
07.000000	4139.958333	8.596000e+03	46.300000	1.000000	0.000000	0.000000
37.000000	5730.000000	1.824950e+04	70.900000	2.000000	0.000000	0.000000
27.000000	17639.958330	1.207359e+06	119.000000	33.000000	13.000000	5.000000
4						>

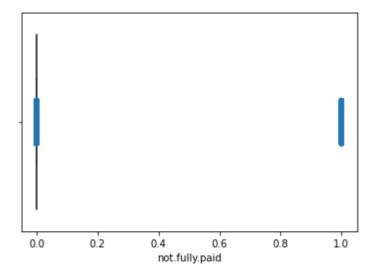
outliers

In [9]:

```
sns.boxplot(x='not.fully.paid',data=df)
sns.stripplot(x='not.fully.paid',data=df)
```

Out[9]:

<AxesSubplot:xlabel='not.fully.paid'>



In [10]:

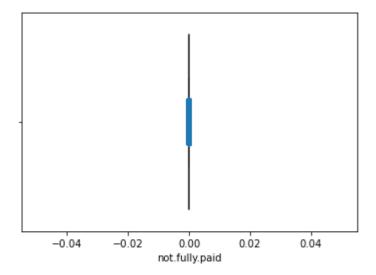
```
from datasist.structdata import detect_outliers
outlier_indices=detect_outliers(df,0,['not.fully.paid'])
outlier_indices
143,
 145,
 150,
 160,
 161,
 164,
 165,
 182,
 187,
 193,
 201,
 204,
 205,
 207,
 211,
 218,
 222,
 226,
 233,
```

In [11]:

```
df.drop(outlier_indices,inplace=True)
sns.boxplot(x='not.fully.paid',data=df)
sns.stripplot(x='not.fully.paid',data=df)
```

Out[11]:

<AxesSubplot:xlabel='not.fully.paid'>



In [13]:

```
x=df.drop('credit.policy',axis=1)
x
```

Out[13]:

	int.rate	installment	log.annual.inc	dti	fico	days.with.cr.line	revol.bal	revol.util	inq.last.6mths	de
0	0.1189	829.10	11.350407	19.48	737	5639.958333	28854	52.1	0	
1	0.1071	228.22	11.082143	14.29	707	2760.000000	33623	76.7	0	
2	0.1357	366.86	10.373491	11.63	682	4710.000000	3511	25.6	1	
3	0.1008	162.34	11.350407	8.10	712	2699.958333	33667	73.2	1	
4	0.1426	102.92	11.299732	14.97	667	4066.000000	4740	39.5	0	
9561	0.0788	115.74	10.999095	10.17	722	4410.000000	11586	61.6	4	
9562	0.1348	508.87	10.933107	17.76	717	3870.041667	8760	28.2	6	
9564	0.1385	511.56	12.323856	12.33	687	6420.041667	385489	51.2	4	
0E67	N 1911	101 24	10 060100	0 72	607	9700 044667	1511	120	E	>

```
In [14]:
```

```
y=df['credit.policy']
y
```

Out[14]:

Name: credit.policy, Length: 8045, dtype: int64

In [15]:

```
from sklearn.model_selection import train_test_split
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.2)
x_train
```

Out[15]:

	int.rate	installment	log.annual.inc	dti	fico	days.with.cr.line	revol.bal	revol.util	inq.l
7504	0.1422	342.85	10.714418	12.48	682	7050.041667	4719	77.4	
1176	0.0832	157.43	10.335010	10.13	742	4307.000000	20337	39.2	
9059	0.1505	180.40	10.126471	5.95	692	2279.958333	225	5.0	
1444	0.1051	243.81	10.858922	7.82	722	4379.958333	5911	20.9	
2574	0.0932	351.42	11.095398	19.15	747	3749.958333	4148	14.3	
473	0.1324	405.72	11.314475	15.12	672	5760.041667	27905	81.8	
2909	0.0768	124.77	10.499573	0.93	742	1919.958333	1410	67.1	
8813	0.1507	52.05	9.035987	10.00	652	1019.958333	1764	53.5	
6806	0.1253	368.13	10.645425	18.89	702	3060.041667	6811	25.7	
6884	0.1357	244.58	11.350407	6.45	692	12061.000000	7983	43.6	

6436 rows × 18 columns

In [16]:

```
x_test
```

Out[16]:

	int.rate	installment	log.annual.inc	dti	fico	days.with.cr.line	revol.bal	revol.util	inq.l
6094	0.1114	518.30	11.512925	21.88	747	4620.000000	29860	47.6	
146	0.0964	90.68	11.156251	18.00	732	3691.000000	55720	10.0	
8078	0.1122	472.94	11.407565	15.00	687	1950.041667	11220	72.4	
5081	0.0894	317.72	11.407565	10.47	767	9630.000000	93093	1.4	
1001	0.1229	166.77	10.471638	14.96	677	4829.958333	18099	69.2	
4925	0.1635	883.23	11.884461	14.32	682	4619.958333	30427	86.9	
2236	0.1347	508.84	11.842287	6.76	707	11550.041670	0	0.0	
4829	0.1357	101.91	10.596635	22.98	672	4500.000000	17590	80.7	
1056	0.1292	168.28	10.596535	16.29	667	2729.958333	14244	85.8	
2931	0.1284	33.62	10.085809	24.35	687	3480.000000	2533	38.4	

1609 rows × 18 columns

In [18]:

```
y_test.value_counts()
```

Out[18]:

1 1342
 267

Name: credit.policy, dtype: int64

In [19]:

```
from imblearn.over_sampling import SMOTE
sampler=SMOTE()
x_train,y_train=sampler.fit_resample(x_train,y_train)
y_train.value_counts()
```

Out[19]:

5354
 5354

Name: credit.policy, dtype: int64

In [25]:

```
x=df.drop('installment',axis=1)
y=df['installment']
```

In [26]:

```
y.value_counts()
```

Out[26]:

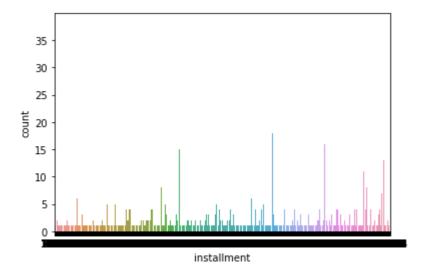
217 7					
317.72	2 38				
316.11	L 33				
319.47	7 29				
381.26	5 27				
156.10	24				
	• •				
680.69	1				
127.82	2 1				
759.31	l 1				
241.55	5 1				
35.83	1				
Name:	$\verb"installment",\\$	Length:	4111,	dtype:	int64

In [27]:

```
sns.countplot(x='installment',data=df)
```

Out[27]:

<AxesSubplot:xlabel='installment', ylabel='count'>



In [28]:

```
from sklearn.model_selection import train_test_split
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.2)
y_train.value_counts()
```

Out[28]:

317.72 30 316.11 29 319.47 23 381.26 22 156.10 20 377.25 1 252.85 1 477.92 1 827.41 1 1 71.00

Name: installment, Length: 3507, dtype: int64

In [30]:

df

Out[30]:

credit.policy	int.rate	installment	log.annual.inc	dti	fico	days.with.cr.line	revol.bal	ı
1	0.1189	829.10	11.350407	19.48	737	5639.958333	28854	
1	0.1071	228.22	11.082143	14.29	707	2760.000000	33623	
1	0.1357	366.86	10.373491	11.63	682	4710.000000	3511	
1	0.1008	162.34	11.350407	8.10	712	2699.958333	33667	
1	0.1426	102.92	11.299732	14.97	667	4066.000000	4740	
0	0.0788	115.74	10.999095	10.17	722	4410.000000	11586	
0	0.1348	508.87	10.933107	17.76	717	3870.041667	8760	
0	0.1385	511.56	12.323856	12.33	687	6420.041667	385489	
0	0.1311	101.24	10.968198	8.23	687	2790.041667	1514	
0	0.1979	37.06	10.645425	22.17	667	5916.000000	28854	
	1 1 1 1 0 0	1 0.1189 1 0.1071 1 0.1357 1 0.1008 1 0.1426 0 0.0788 0 0.1348 0 0.1385 0 0.1311	1 0.1189 829.10 1 0.1071 228.22 1 0.1357 366.86 1 0.1008 162.34 1 0.1426 102.92 0 0.0788 115.74 0 0.1348 508.87 0 0.1385 511.56 0 0.1311 101.24	1 0.1189 829.10 11.350407 1 0.1071 228.22 11.082143 1 0.1357 366.86 10.373491 1 0.1008 162.34 11.350407 1 0.1426 102.92 11.299732 0 0.0788 115.74 10.999095 0 0.1348 508.87 10.933107 0 0.1385 511.56 12.323856 0 0.1311 101.24 10.968198	1 0.1189 829.10 11.350407 19.48 1 0.1071 228.22 11.082143 14.29 1 0.1357 366.86 10.373491 11.63 1 0.1008 162.34 11.350407 8.10 1 0.1426 102.92 11.299732 14.97 0 0.0788 115.74 10.999095 10.17 0 0.1348 508.87 10.933107 17.76 0 0.1385 511.56 12.323856 12.33 0 0.1311 101.24 10.968198 8.23	1 0.1189 829.10 11.350407 19.48 737 1 0.1071 228.22 11.082143 14.29 707 1 0.1357 366.86 10.373491 11.63 682 1 0.1008 162.34 11.350407 8.10 712 1 0.1426 102.92 11.299732 14.97 667 0 0.0788 115.74 10.999095 10.17 722 0 0.1348 508.87 10.933107 17.76 717 0 0.1385 511.56 12.323856 12.33 687 0 0.1311 101.24 10.968198 8.23 687	1 0.1189 829.10 11.350407 19.48 737 5639.958333 1 0.1071 228.22 11.082143 14.29 707 2760.000000 1 0.1357 366.86 10.373491 11.63 682 4710.000000 1 0.1008 162.34 11.350407 8.10 712 2699.958333 1 0.1426 102.92 11.299732 14.97 667 4066.000000 0 0.0788 115.74 10.999095 10.17 722 4410.000000 0 0.1348 508.87 10.933107 17.76 717 3870.041667 0 0.1385 511.56 12.323856 12.33 687 6420.041667 0 0.1311 101.24 10.968198 8.23 687 2790.041667	1 0.1189 829.10 11.350407 19.48 737 5639.958333 28854 1 0.1071 228.22 11.082143 14.29 707 2760.000000 33623 1 0.1357 366.86 10.373491 11.63 682 4710.000000 3511 1 0.1008 162.34 11.350407 8.10 712 2699.958333 33667 1 0.1426 102.92 11.299732 14.97 667 4066.000000 4740 0 0.0788 115.74 10.999095 10.17 722 4410.000000 11586 0 0.1348 508.87 10.933107 17.76 717 3870.041667 8760 0 0.1385 511.56 12.323856 12.33 687 6420.041667 385489 0 0.1311 101.24 10.968198 8.23 687 2790.041667 1514

8045 rows × 19 columns

In [31]:

```
x=df[['credit.policy','not.fully.paid','purpose_credit_card']]
y=df['installment']
from sklearn.model_selection import train_test_split
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.25)
x_train
```

Out[31]:

	credit.policy	not.fully.paid	purpose_credit_card
3565	1	0	0
9441	0	0	0
731	1	0	0
9101	0	0	0
2638	1	0	0
5885	1	0	0
4917	1	0	0
1032	1	0	0
1918	1	0	0
5537	1	0	0

6033 rows × 3 columns

In [32]:

```
from sklearn.preprocessing import MinMaxScaler
scaler=MinMaxScaler()
scaler.fit(x_train)
x_train=scaler.transform(x_train)
x_test=scaler.transform(x_test)
x_train
```

Out[32]:

In [33]:

```
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.25)
x_train
```

Out[33]:

credit.policy	not.fully.paid	purpose_credit_card
1	0	0
1	0	0
1	0	0
1	0	0
1	0	0
1	0	0
1	0	0
0	0	0
1	0	0
1	0	0
	1 1 1 1 1 1 0	1 0 1 0 1 0 1 0 1 0 1 0 1 0 1 0 1 0 1 0

6033 rows × 3 columns

In [34]:

```
from sklearn.preprocessing import StandardScaler
scaler=StandardScaler()
x_train=scaler.fit_transform(x_train)
x_test=scaler.transform(x_test)
x_train
```

Out[34]:

In []: