```
## import libraries
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import warnings
warnings.filterwarnings('ignore')
#dataload
data_loan = pd.read_csv("loan.csv")
#see the data
data_loan.head(5)
\overline{\Rightarrow}
              id member_id loan_amnt funded_amnt funded_amnt_inv
                                                                                int_rate instal
      0 1077501
                    1296599
                                   5000
                                                5000
                                                                 4975.0
                                                                                   10.65%
                                                                        months
                                                                            60
      1 1077430
                     1314167
                                   2500
                                                2500
                                                                 2500.0
                                                                                   15.27%
                                                                        months
                                                                 2400.0
      2 1077175
                    1313524
                                   2400
                                                2400
                                                                                   15.96%
                                                                        months
      3 1076863
                    1277178
                                  10000
                                               10000
                                                                10000.0
                                                                                   13.49%
                                                                        months
                                                                 3000.0
      4 1075358
                    1311748
                                   3000
                                                3000
                                                                                   12.69%
                                                                        months
     5 rows × 111 columns
```

data\_loan.shape

→ (39717, 111)

#getting the null values
data\_loan.isnull().sum()

```
id
                                    0
member_id
                                    0
loan_amnt
                                    0
{\sf funded\_amnt}
                                    0
funded_amnt_inv
                                    0
tax_liens
                                   39
tot_hi_cred_lim
                                39717
total_bal_ex_mort
                                39717
total_bc_limit
                                39717
 total_il_high_credit_limit
                                39717
Length: 111, dtype: int64
```

### \*FROM the above analysis lots of data is null so revoving null characters \*

```
Start coding or generate with AI.

data_loan.dropna(axis = 1, how = 'all', inplace = True)
data_loan.head(5)
```

<b>→</b>		id	member_id	loan_amnt	funded_amnt	funded_amnt_inv	term	int_rate	instal
	0	1077501	1296599	5000	5000	4975.0	36 months	10.65%	,
	1	1077430	1314167	2500	2500	2500.0	60 months	15.27%	
	2	1077175	1313524	2400	2400	2400.0	36 months	15.96%	
	3	1076863	1277178	10000	10000	10000.0	36 months	13.49%	:
	4	1075358	1311748	3000	3000	3000.0	60 months	12.69%	

5 rows × 57 columns

<pre>data_loan.info()</pre>							
_	1	member_id	39717 non-null	int64			
	2	loan amnt	39717 non-null	int64			
	3	funded amnt	39717 non-null	int64			
	4	funded_amnt_inv	39717 non-null	float64			
	5	term	39717 non-null	object			
	6	int_rate	39717 non-null	object			
	7	installment	39717 non-null	float64			
	8	grade	39717 non-null	object			
	9	sub_grade	39717 non-null	object			
	10	emp_title	37258 non-null	object			
	11	emp_length	38642 non-null	object			
	12	home_ownership	39717 non-null	object			
	13	annual_inc	39717 non-null	float64			
	14	verification_status	39717 non-null	object			
	15	issue_d	397 <b>1</b> 7 non-null	object			
	16	loan_status	39717 non-null	object			
	17	pymnt_plan	39717 non-null	object			
	18	url	39717 non-null	object			
	19	desc	26775 non-null	object			
	20	purpose	39717 non-null	object			
	21	title	39706 non-null	object			
	22	zip_code	39717 non-null	object			
	23	addr_state	39717 non-null	object			
	24	dti	39717 non-null	float64			
	25	delinq_2yrs	39717 non-null	int64			
	26	earliest_cr_line	39717 non-null	object			
	27	inq_last_6mths	39717 non-null 14035 non-null	int64			
	28 29	<pre>mths_since_last_delinq mths_since_last_record</pre>	2786 non-null	float64 float64			
	30	open acc	39717 non-null	int64			
	31	pub_rec	39717 non-null	int64			
	32	revol_bal	39717 non-null	int64			
	33	revol_util	39667 non-null	object			
	34	total acc	39717 non-null	int64			
	35	initial_list_status	39717 non-null	object			
	36	out_prncp	39717 non-null	float64			
	37	out_prncp_inv	39717 non-null	float64			
	38	total_pymnt	39717 non-null	float64			
	39	total_pymnt_inv	39717 non-null	float64			
	40	total_rec_prncp	39717 non-null	float64			
	41	total_rec_int	39717 non-null	float64			
	42	total_rec_late_fee	39717 non-null	float64			
	43	recoveries	39717 non-null	float64			
	44	collection_recovery_fee	39717 non-null	float64			
	45	last_pymnt_d	39646 non-null	object			
	46	last_pymnt_amnt	39717 non-null	float64			
	47	next_pymnt_d	1140 non-null	object			
	48	last_credit_pull_d	39715 non-null	object			
	49	collections_12_mths_ex_med					
	50	policy_code	39717 non-null	int64			
	51	application type	39717 non-null	object			
	52	acc_now_delinq	39717 non-null	int64			
	53	chargeoff_within_12_mths	39661 non-null	float64			
	54	delinq_amnt	39717 non-null	int64			
	55	pub_rec_bankruptcies	39020 non-null	float64			
	56	tax liens	39678 non-null	float64			
		es: float64(20), int64(13),					
		ry usage: 17.3+ MB					

data\_loan.columns

```
Index(['id', 'member_id', 'loan_amnt', 'funded_amnt', 'funded_amnt_inv',
                 'term', 'int_rate', 'installment', 'grade', 'sub_grade', 'emp_title',
                term', 'int_rate', 'installment', 'grade', 'sub_grade', 'emp_title',
'emp_length', 'home_ownership', 'annual_inc', 'verification_status',
'issue_d', 'loan_status', 'pymnt_plan', 'url', 'desc', 'purpose',
'title', 'zip_code', 'addr_state', 'dti', 'delinq_2yrs',
'earliest_cr_line', 'inq_last_6mths', 'mths_since_last_delinq',
'mths_since_last_record', 'open_acc', 'pub_rec', 'revol_bal',
'revol_util', 'total_acc', 'initial_list_status', 'out_prncp',
'out_prncp_inv', 'total_pymnt', 'total_pymnt_inv', 'total_rec_prncp',
'total_rec_int', 'total_rec_late_fee', 'recoveries',
'collection recovery fee'. 'last pymnt d'. 'last pymnt amnt'
                 'collection_recovery_fee', 'last_pymnt_d', 'last_pymnt_amnt', 'next_pymnt_d', 'last_credit_pull_d', 'collections_12_mths_ex_med', 'policy_code', 'application_type', 'acc_now_delinq',
                 'chargeoff_within_12_mths', 'delinq_amnt', 'pub_rec_bankruptcies',
                 'tax_liens'],
               dtype='object')
data loan =data loan[data loan.loan status != "Current"]
data_loan.loan_status.unique()
→ array(['Fully Paid', 'Charged Off'], dtype=object)
(data_loan.isna().sum()/len(data_loan.index))*100
                                                    9 999999
      id
       member_id
                                                    0.000000
       loan amnt
                                                    0.000000
       funded amnt
                                                   0.000000
       funded_amnt_inv
                                                    0.000000
       term
                                                    0.000000
       int rate
                                                    0.000000
       installment
                                                    0.000000
       grade
                                                    0.000000
       sub_grade
                                                    0.000000
       emp_title
                                                    6.185033
       emp_length
                                                    2.677761
       home_ownership
                                                    0.000000
                                                    0.000000
       annual inc
       verification_status
                                                    0.000000
       issue_d
                                                    0.000000
       loan_status
                                                    0.000000
                                                    0.000000
       pymnt_plan
       url
                                                   0.000000
                                                  32.477901
       desc
                                                   0.000000
       purpose
       title
                                                    0.028514
       zip_code
                                                    0.000000
                                                   0.000000
       addr_state
       dti
                                                   0.000000
       delinq_2yrs
                                                    0.000000
       earliest_cr_line
                                                    0.000000
       inq_last_6mths
                                                   0.000000
       mths_since_last_delinq
                                                  64.559193
      mths_since_last_record
                                                  92.897322
                                                    0.000000
       open_acc
       pub_rec
                                                    0.000000
       revol_bal
                                                    0.000000
                                                    0.129611
       revol util
       total acc
                                                    0.000000
       initial_list_status
                                                    0.000000
       out_prncp
                                                    0.000000
       out prncp inv
                                                   0.000000
       total_pymnt
                                                    0.000000
                                                    0.000000
       total_pymnt_inv
                                                    0.000000
       total_rec_prncp
       total_rec_int
                                                    0.000000
                                                   0.000000
       total_rec_late_fee
       recoveries
                                                   0.000000
                                                    0.000000
       collection_recovery_fee
       last_pymnt_d
                                                   0.184047
       last pymnt amnt
                                                    0.000000
                                                 100.000000
       next pymnt d
       last_credit_pull_d
                                                   0.005184
                                                    0.145164
       collections_12_mths_ex_med
       policy code
                                                    0.000000
       application_type
                                                    0.000000
       acc_now_delinq
                                                    0.000000
       chargeoff_within_12_mths
                                                    0.145164
       delinq_amnt
                                                    0.000000
       pub_rec_bankruptcies
                                                    1.806776
```

tax\_liens ט.נטוטי. dtype: float64

data loan.info()

<<class 'pandas.core.frame.DataFrame'> Index: 38577 entries, 0 to 39716 Data columns (total 57 columns): Non-Null Count Dtype Column # -----0 id 38577 non-null int64 member\_id 38577 non-null int64 1 2 loan\_amnt 38577 non-null int64 funded\_amnt 38577 non-null int64 4 funded\_amnt\_inv 38577 non-null float64 5 38577 non-null object int\_rate 38577 non-null installment 38577 non-null float64 38577 non-null object 8 grade 9 sub\_grade 38577 non-null object 10 emp title 36191 non-null object 37544 non-null emp\_length object 11 12 home\_ownership 38577 non-null object 38577 non-null 13 annual\_inc float64 verification\_status 38577 non-null obiect 14 15 issue\_d 38577 non-null object 16 loan\_status 38577 non-null object 38577 non-null 17 pymnt\_plan object url 38577 non-null object 18 19 desc 26048 non-null object 38577 non-null purpose object 21 38566 non-null title obiect 38577 non-null 22 zip\_code object 23 addr\_state 38577 non-null 24 dti 38577 non-null float64 25 delinq\_2yrs 38577 non-null int64 26 earliest\_cr\_line 38577 non-null object 38577 non-null 27 ing last 6mths mths\_since\_last\_delinq 13672 non-null float64 28 29 mths\_since\_last\_record 2740 non-null float64 30 38577 non-null open\_acc int64 38577 non-null int64 31 pub rec 38577 non-null 32 revol bal int64 33 revol\_util 38527 non-null object 38577 non-null total\_acc 35 initial\_list\_status 38577 non-null obiect out\_prncp 36 38577 non-null float64 37 out\_prncp\_inv 38577 non-null float64 38 total\_pymnt 38577 non-null float64 38577 non-null float64 39 total\_pymnt\_inv 40 total\_rec\_prncp 38577 non-null float64 41 total rec int 38577 non-null float64 42 total\_rec\_late\_fee 38577 non-null float64 43 recoveries 38577 non-null float64 collection\_recovery\_fee 38577 non-null 38506 non-null last\_pymnt\_d obiect 45 38577 non-null float64 46 last\_pymnt\_amnt next\_pymnt\_d 47 0 non-null object 48 last credit pull d 38575 non-null object

print("Mode : " + data\_loan.emp\_length.mode()[0])
data\_loan.emp\_length.value\_counts()

collections\_12\_mths\_ex\_med 38521 non-null float64

38577 non-null int64 38577 non-null object

38577 non-null int64

→ Mode : 10+ years emp\_length 8488 10+ years < 1 year 4508 2 years 4291 4012 3 vears 4 years 3342 5 years 3194 1 vear 3169 6 years 2168 7 vears 1711 1435 8 years 1226 9 vears Name: count, dtype: int64

policy\_code

51 application\_type
52 acc\_now\_delinq

49 50 Start coding or generate with AI.

# We can come to the conclusion that the mode value is higher than most frequent value

```
data_loan.emp_length.fillna(data_loan.emp_length.mode()[0], inplace = True)
data_loan.emp_length.isna().sum()

data_loan.emp_length.fillna(data_loan.emp_length.mode()[0], inplace = True)
data_loan.emp_length.isna().sum()

0
```

#### STANDERDIZE THE DATA

```
data_loan.int_rate = pd.to_numeric(data_loan.int_rate.apply(lambda x : x.split('%')[0]))

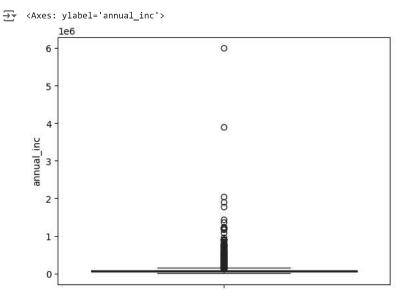
data_loan.emp_length = pd.to_numeric(data_loan.emp_length.apply(lambda x: 0 if "<" in x else (x.split('+')[0] if "+" in x else x.split()[0])

data_loan.head(20)</pre>
```

							ū		
₹		id	member_id	loan_amnt	funded_amnt	funded_amnt_inv	term	int_rate	insta
	0	1077501	1296599	5000	5000	4975.00000	36 months	10.65	
	1	1077430	1314167	2500	2500	2500.00000	60 months	15.27	
	2	1077175	1313524	2400	2400	2400.00000	36 months	15.96	
	3	1076863	1277178	10000	10000	10000.00000	36 months	13.49	
	5	1075269	1311441	5000	5000	5000.00000	36 months	7.90	
	6	1069639	1304742	7000	7000	7000.00000	60 months	15 <u>.</u> 96	
	7	1072053	1288686	3000	3000	3000.00000	36 months	18.64	
	8	1071795	1306957	5600	5600	5600.00000	60 months	21.28	
	9	1071570	1306721	5375	5375	5350.00000	60 months	12.69	
	10	1070078	1305201	6500	6500	6500.00000	60 months	14.65	
	11	1069908	1305008	12000	12000	12000.00000	36 months	12.69	
	12	1064687	1298717	9000	9000	9000.00000	36 months	13.49	
	13	1069866	1304956	3000	3000	3000.00000	36 months	9.91	
	14	1069057	1303503	10000	10000	10000.00000	36 months	10.65	
	15	1069759	1304871	1000	1000	1000.00000	36 months	16.29	
	16	1065775	1299699	10000	10000	10000.00000	36 months	15.27	
	17	1069971	1304884	3600	3600	3600.00000	36 months	6.03	
	18	1062474	1294539	6000	6000	6000.00000	36 months	11.71	
	19	1069742	1304855	9200	9200	9200.00000	36 months	6.03	
	20	1069740	1284848	20250	20250	19142.16108	60 months	15.27	
	20 rc	ws × 57 cc	olumns						

## Outlier boxplot

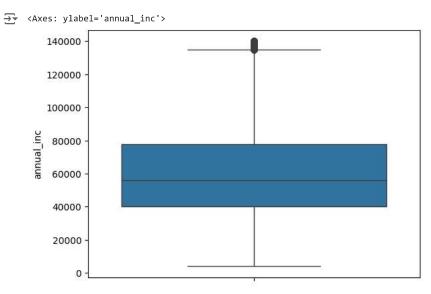
sns.boxplot(data\_loan['annual\_inc'], color ="green")



```
0.50
               58868.0
\overline{\pm}
               82000.0
     0.75
     0.90
             115000.0
             140004.0
     0.95
    0.97
             165000.0
     0.98
             187000.0
     0.99
             234144.0
     Name: annual_inc, dtype: float64
```

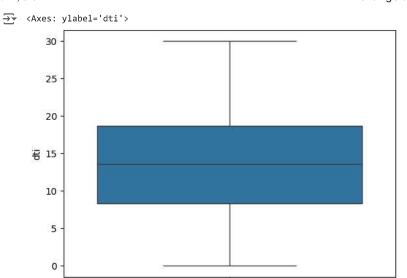
```
per_95_annual_inc = data_loan['annual_inc'].quantile(0.95)
data_loan = data_loan[data_loan.annual_inc <= per_95_annual_inc]</pre>
```

sns.boxplot(data\_loan.annual\_inc)

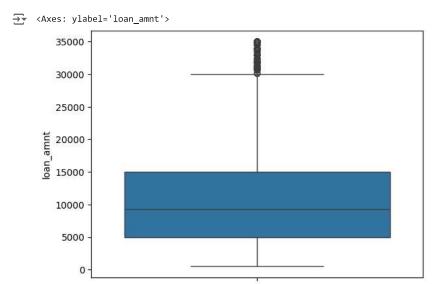


## Now the "annual\_inc" data looks good

sns.boxplot(data\_loan.dti)



sns.boxplot(data\_loan.loan\_amnt)



data\_loan.loan\_amnt.quantile([0.75,0.90,0.95,0.97,0.975, 0.98, 0.99, 1.0])

```
→ 0.750
             15000.0
             20000.0
    0.900
    0.950
             25000.0
    0.970
             25475.0
    0.975
             28000.0
    0.980
             30000.0
    0.990
             35000.0
    1.000
             35000.0
    Name: loan_amnt, dtype: float64
```

sns.boxplot(data\_loan.funded\_amnt\_inv)

```
Axes: ylabel='funded_amnt_inv'>

35000 -

25000 -

25000 -

15000 -

5000 -

0 -
```

data\_loan.sub\_grade = pd.to\_numeric(data\_loan.sub\_grade.apply(lambda x : x[-1]))

0.950 23941.875589 0.970 24975.000000 0.975 25350.000000 0.980 27950.000000 0.985 29925.000000 0.990 31954.028576 1.000 35000.000000

Name: funded\_amnt\_inv, dtype: float64

#### Visualizing Categorical Data

```
data_loan.sub_grade.head(5)

### 0 2

1 4

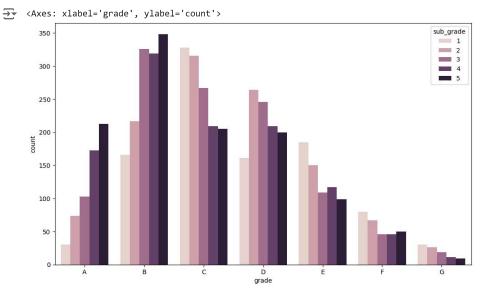
2 5

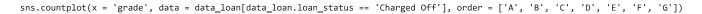
3 1

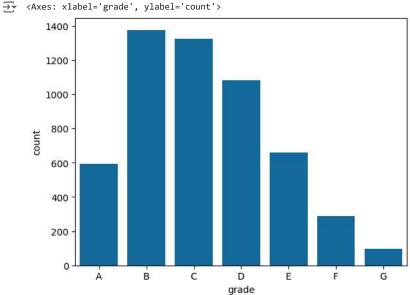
5 4

Name: sub_grade, dtype: int64

fig, ax = plt.subplots(figsize=(12,7))
sns.set_palette('colorblind')
sns.countplot(x = 'grade', order = ['A', 'B', 'C', 'D', 'E', 'F', 'G'] , hue = 'sub_grade',data = data_loan[data_loan.loan_status == 'Charge')
```





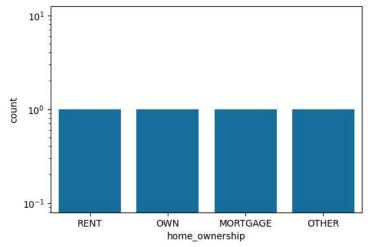


```
data_loan['home_ownership'].unique()

array(['RENT', 'OWN', 'MORTGAGE', 'OTHER', 'NONE'], dtype=object)

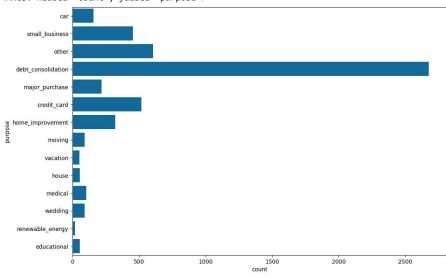
# replacing none values
data_loan['home_ownership'].replace(to_replace = ['NONE'],value='OTHER',inplace = True)

fig, ax = plt.subplots(figsize = (6,4))
ax.set(yscale = 'log')
sns.countplot(x='home_ownership', data=data_loan[data_loan['loan_status']=='Charged Off'])
```



```
fig, ax = plt.subplots(figsize = (12,8))
ax.set(xscale = 'linear')
sns.countplot(y = 'purpose', data=data_loan[data_loan.loan_status == 'Charged Off'])
```

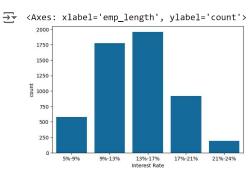


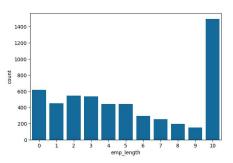


```
#creating bins for int_rate,open_acc,revol_util,total_acc
data_loan['int_rate_groups'] = pd.cut(data_loan['int_rate'], bins=5,precision =0,labels=['5%-9%','9%-13%','13%-17%','17%-21%','21%-24%'])
data_loan['open_acc_groups'] = pd.cut(data_loan['open_acc'],bins = 5,precision =0,labels=['2-10','10-19','19-27','27-36','36-44'])
data_loan['total_acc_groups'] = pd.cut(data_loan['total_acc'], bins=5,precision =0,labels=['2-20','20-37','37-55','55-74','74-90'])
data_loan['annual_inc_groups'] = pd.cut(data_loan['annual_inc'], bins=5,precision =0,labels =['3k-31k','31k-58k','58k-85k','85k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k-112k','112k','112k','112k-112k','112k','112k','112k','112k','112k','112k','112k','112k','112k','11
```

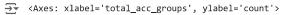
#### Intrest rate wrt bins

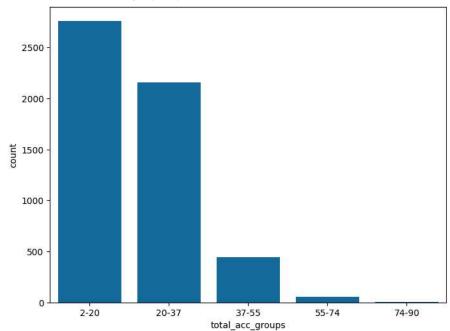
```
fig, ax = plt.subplots(figsize = (15,10))
plt.subplot(221)
sns.countplot(x='int_rate_groups', data=data_loan[data_loan.loan_status == 'Charged Off'])
plt.xlabel('Interest Rate')
plt.subplot(222)
sns.countplot(x='emp_length', data=data_loan[data_loan.loan_status == 'Charged Off'])
```



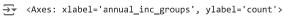


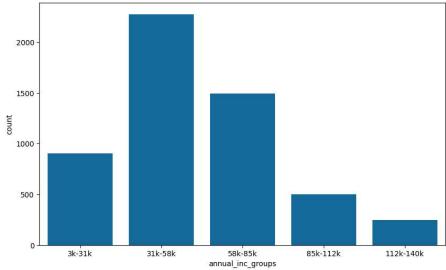
fig, ax = plt.subplots(figsize = (8,6))
ax.set\_yscale('linear')
sns.countplot(x='total\_acc\_groups', data=data\_loan[data\_loan.loan\_status == 'Charged Off'])





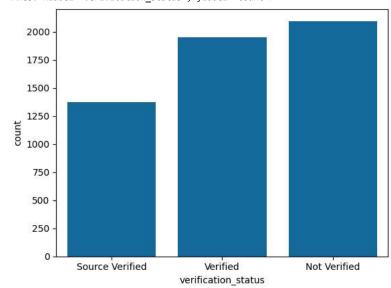
fig, ax = plt.subplots(figsize = (10,6))
sns.countplot(x='annual\_inc\_groups', data=data\_loan[data\_loan.loan\_status == 'Charged Off'])





 $sns.countplot(x='verification\_status', \ data=data\_loan[data\_loan['loan\_status']=='Charged \ Off'])$ 





sns.countplot(y='term', data=data\_loan[data\_loan['loan\_status']=='Charged Off'])

```
<Axes: xlabel='count', ylabel='term'>
60 months -
```

fig,ax = plt.subplots(figsize = (10,8))
ax.set\_yscale('linear')
sns.countplot(x='inq\_last\_6mths', data=data\_loan[data\_loan['loan\_status']=='Charged Off'])



