

PROJECT–Computer hardware

February 1, 2023

```
[1]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import warnings
warnings.filterwarnings('ignore')
```

```
[133]: df=pd.read_table('https://archive.ics.uci.edu/ml/machine-learning-databases/
↳cpu-performance/machine.
↳data',sep=',',header=None,names=['Vendors','Model','MYCT','MMIN','MMAX','CACH','CHMIN','CHMAX','PRP','ERP'])
```

```
[134]: df.head()
```

```
[134]:
```

	Vendors	Model	MYCT	MMIN	MMAX	CACH	CHMIN	CHMAX	PRP	ERP
0	adviser	32/60	125	256	6000	256	16	128	198	199
1	amdahl	470v/7	29	8000	32000	32	8	32	269	253
2	amdahl	470v/7a	29	8000	32000	32	8	32	220	253
3	amdahl	470v/7b	29	8000	32000	32	8	32	172	253
4	amdahl	470v/7c	29	8000	16000	32	8	16	132	132

```
[148]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 209 entries, 0 to 208
Data columns (total 10 columns):
#   Column      Non-Null Count  Dtype
---  -
0   Vendors      209 non-null    object
1   Model        209 non-null    object
2   MYCT         209 non-null    int64
3   MMIN         209 non-null    int64
4   MMAX         209 non-null    int64
5   CACH         209 non-null    int64
6   CHMIN        209 non-null    int64
7   CHMAX        209 non-null    int64
8   PRP          209 non-null    int64
9   ERP          209 non-null    int64
dtypes: int64(8), object(2)
```

memory usage: 16.5+ KB

```
[149]: df.isna().sum()
```

```
[149]: Vendors      0
      Model      0
      MYCT      0
      MMIN      0
      MMAX      0
      CACH      0
      CHMIN     0
      CHMAX     0
      PRP       0
      ERP       0
      dtype: int64
```

1 EDA

```
[150]: df.head()
```

```
[150]:   Vendors  Model MYCT  MMIN  MMAX  CACH  CHMIN  CHMAX  PRP  ERP
0  adviser  32/60  125   256  6000   256    16   128  198  199
1  amdahl  470v/7   29  8000  32000    32     8    32  269  253
2  amdahl  470v/7a   29  8000  32000    32     8    32  220  253
3  amdahl  470v/7b   29  8000  32000    32     8    32  172  253
4  amdahl  470v/7c   29  8000  16000    32     8    16  132  132
```

```
[151]: df.isna().sum()
```

```
[151]: Vendors      0
      Model      0
      MYCT      0
      MMIN      0
      MMAX      0
      CACH      0
      CHMIN     0
      CHMAX     0
      PRP       0
      ERP       0
      dtype: int64
```

```
[145]: df['ERP'].unique()
```

```
[145]: array([ 199,  253,  132,  290,  381,  749, 1238,   23,   24,   70,  117,
         15,   64,   29,   22,  124,   35,   39,   40,   45,   28,   21,
         27,  102,   74,  138,  136,   44,   30,   41,   54,   18,   36,
         38,   34,   19,   72,   56,   42,   75,  113,  157,   20,   33,
```

```

47, 25, 52, 50, 53, 73, 32, 175, 57, 181, 82,
171, 361, 350, 220, 17, 26, 31, 76, 59, 65, 101,
116, 128, 37, 46, 80, 88, 86, 95, 107, 119, 120,
48, 126, 266, 270, 426, 151, 267, 603, 62, 78, 142,
281, 190, 67, 43, 99, 81, 149, 183, 275, 382, 182,
227, 341, 360, 919, 978], dtype=int64)

```

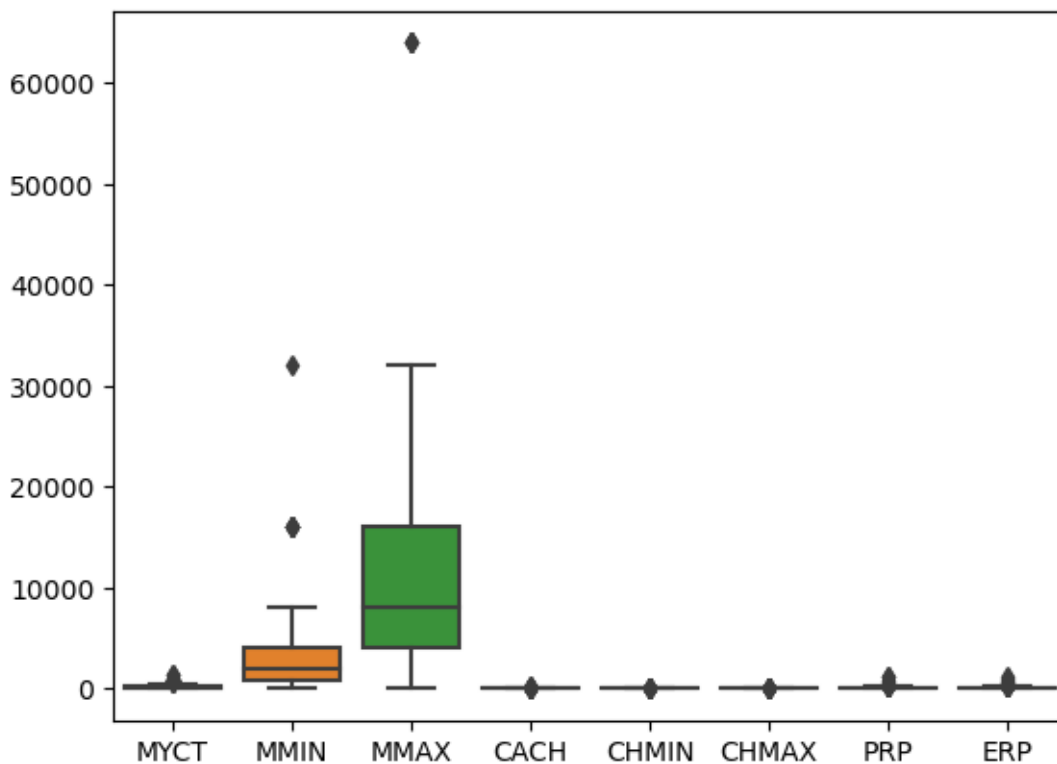
```
[162]: df.replace([np.inf, -np.inf], np.nan, inplace=True)
```

```
[147]: from sklearn.impute import SimpleImputer
```

```
[167]: si=SimpleImputer(missing_values=np.nan, strategy='mean')
df[['CHMAX']] = si.fit_transform(df[['CHMAX']])
```

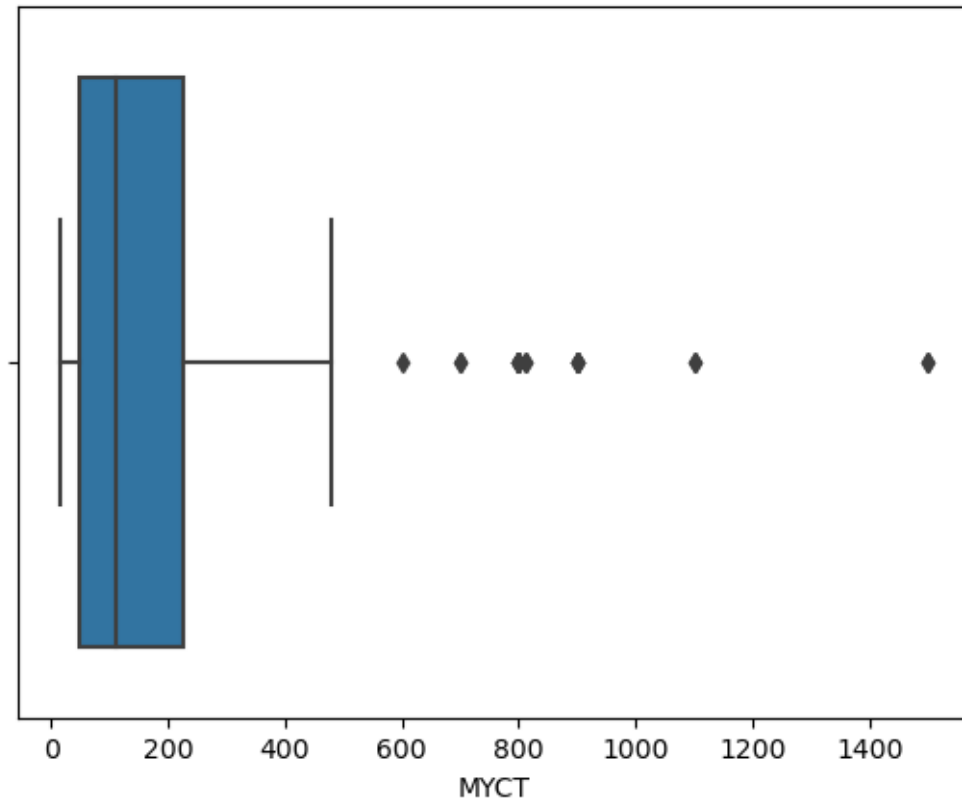
```
[152]: sns.boxplot(data=df)
```

```
[152]: <AxesSubplot:>
```



```
[153]: sns.boxplot(x='MYCT', data=df)
```

```
[153]: <AxesSubplot: xlabel='MYCT'>
```



```
[154]: colname=df.select_dtypes(['int64']).columns
```

```
[155]: df[colname]
```

```
[155]:
```

	MYCT	MMIN	MMAX	CACH	CHMIN	CHMAX	PRP	ERP
0	125	256	6000	256	16	128	198	199
1	29	8000	32000	32	8	32	269	253
2	29	8000	32000	32	8	32	220	253
3	29	8000	32000	32	8	32	172	253
4	29	8000	16000	32	8	16	132	132
..
204	124	1000	8000	0	1	8	42	37
205	98	1000	8000	32	2	8	46	50
206	125	2000	8000	0	2	14	52	41
207	480	512	8000	32	0	0	67	47
208	480	1000	4000	0	0	0	45	25

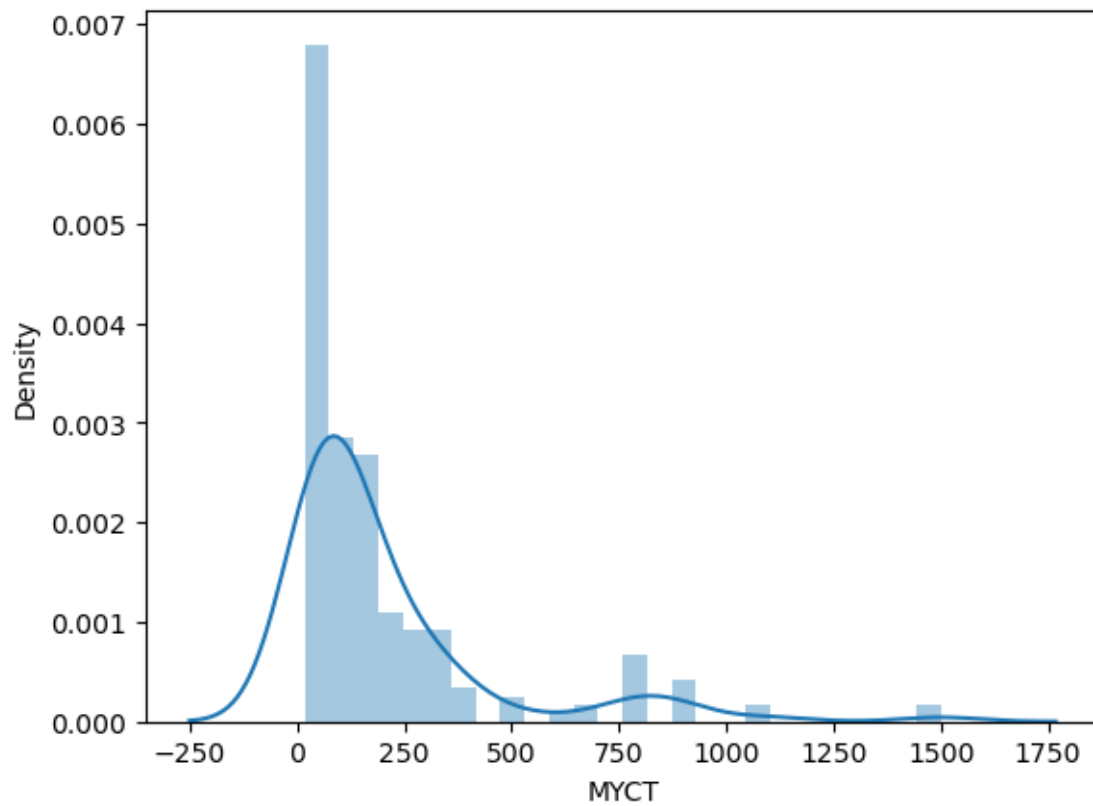
[209 rows x 8 columns]

```
[156]: from scipy.stats import skew
```

```
[157]: for i in df[colname]:  
        print(i)  
        print(skew(df[i]))  
        sns.distplot(df[i])  
        plt.show()
```

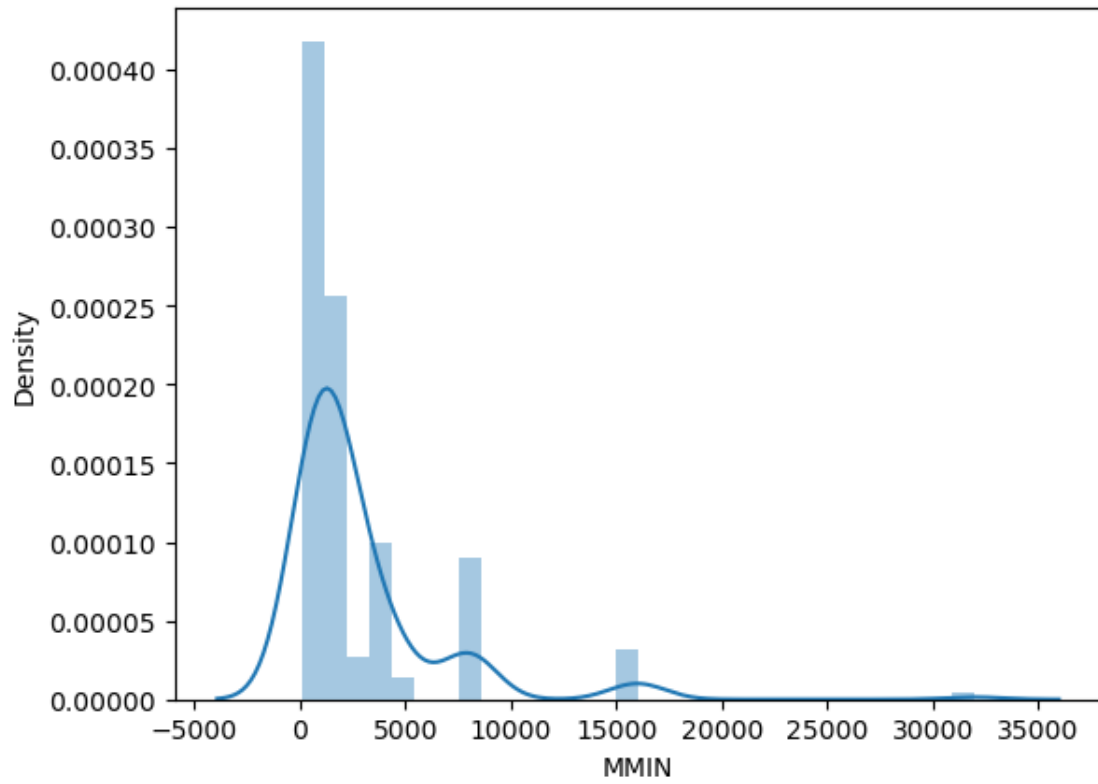
MYCT

2.5258570096766686

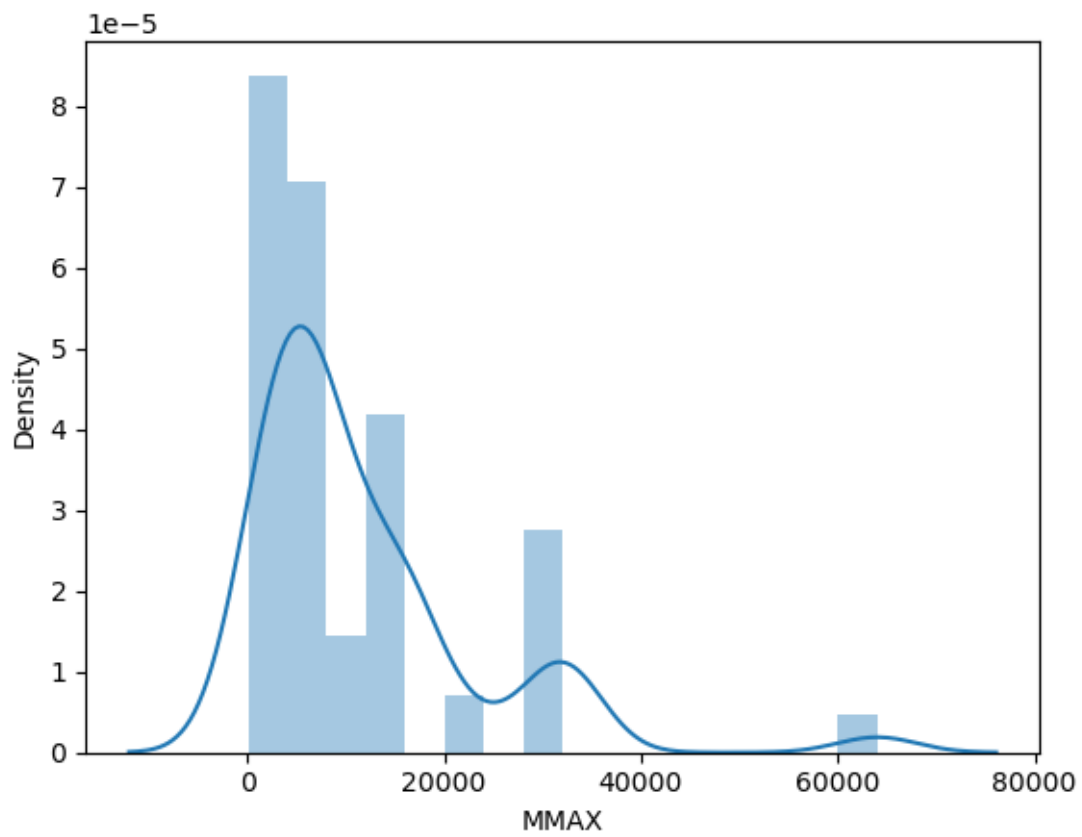


MMIN

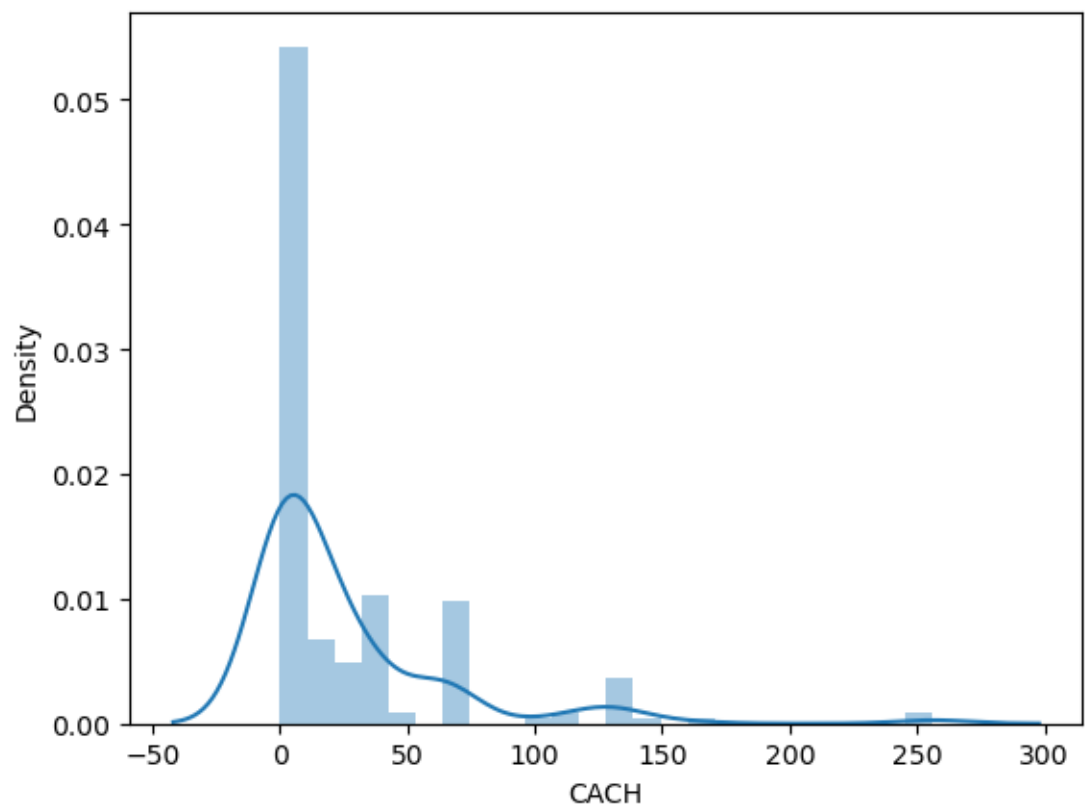
3.4906489998203925



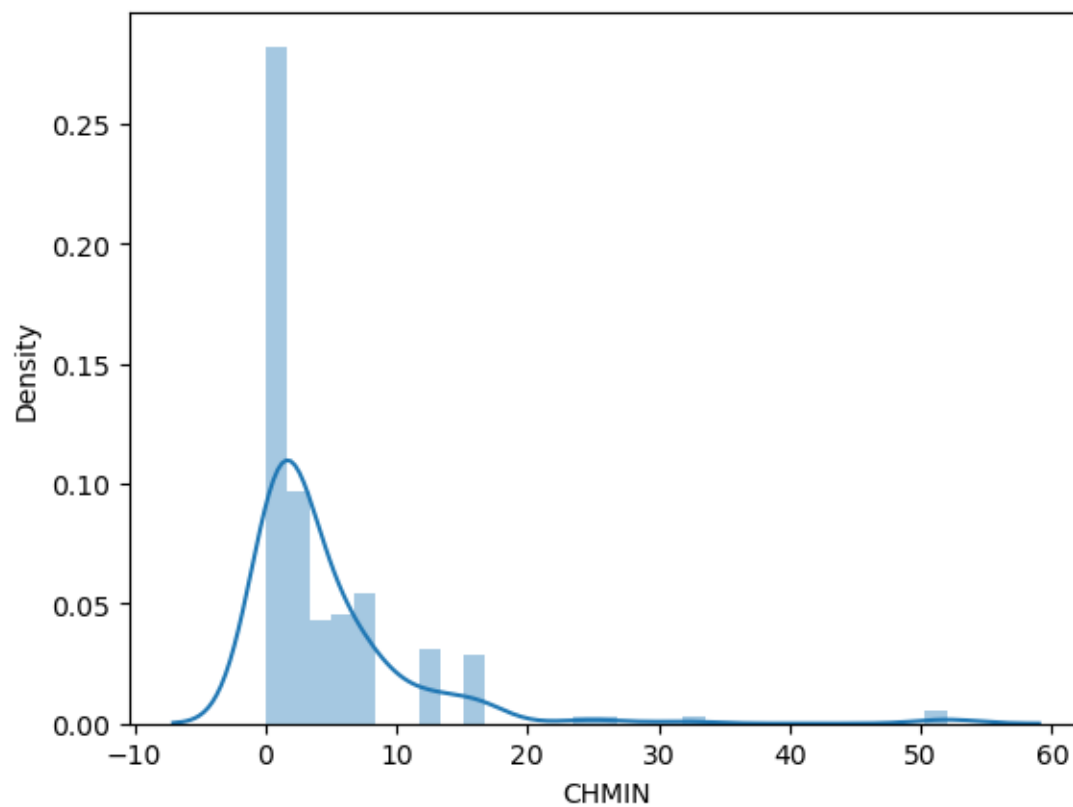
MMAX
2.1252682972972194



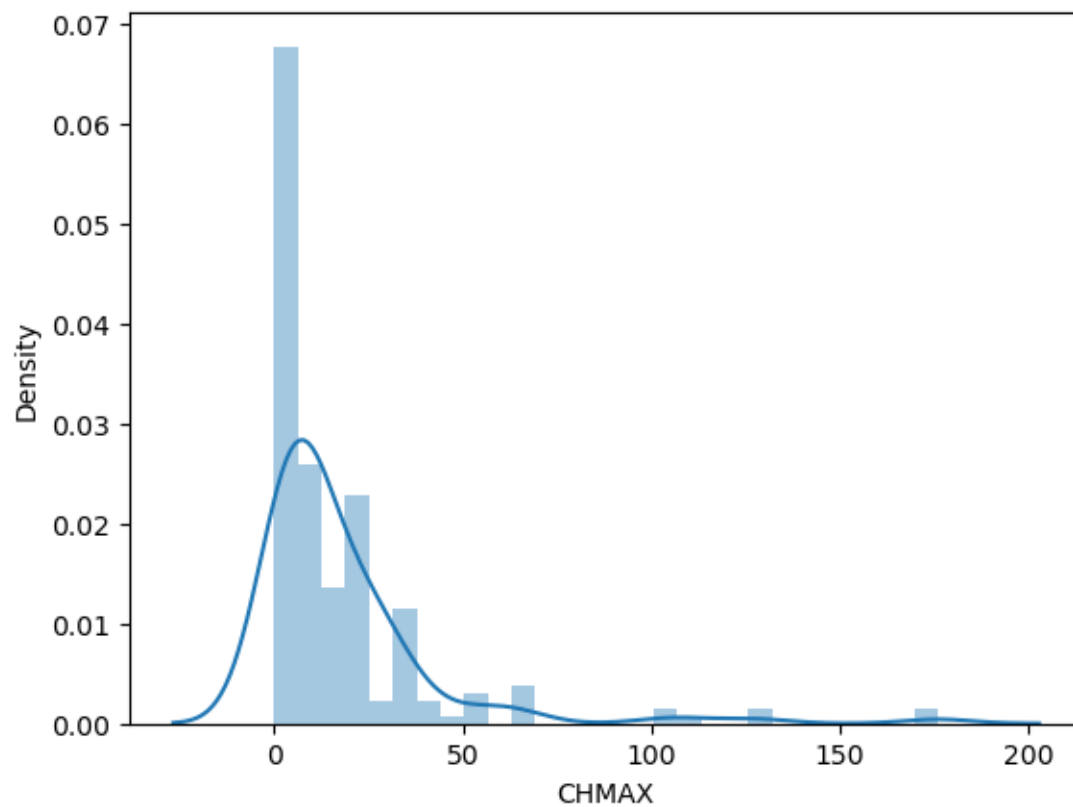
CACH
2.8044632567259247



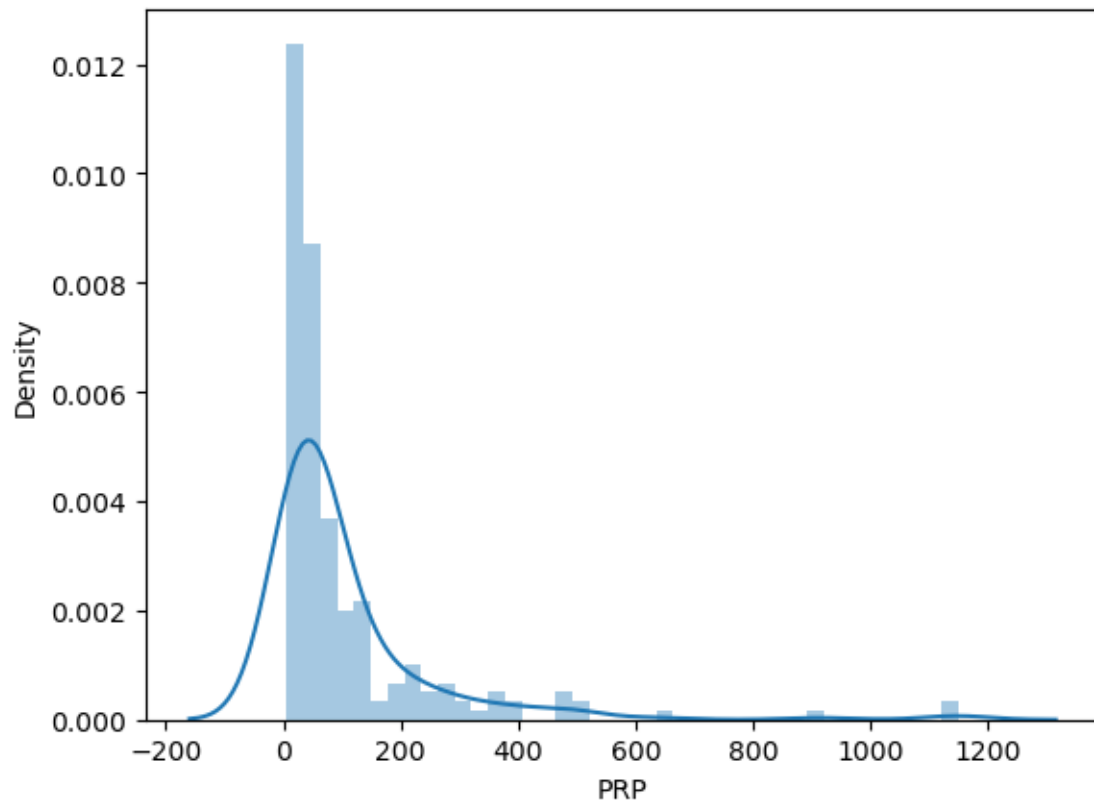
CHMIN
3.998370744836009



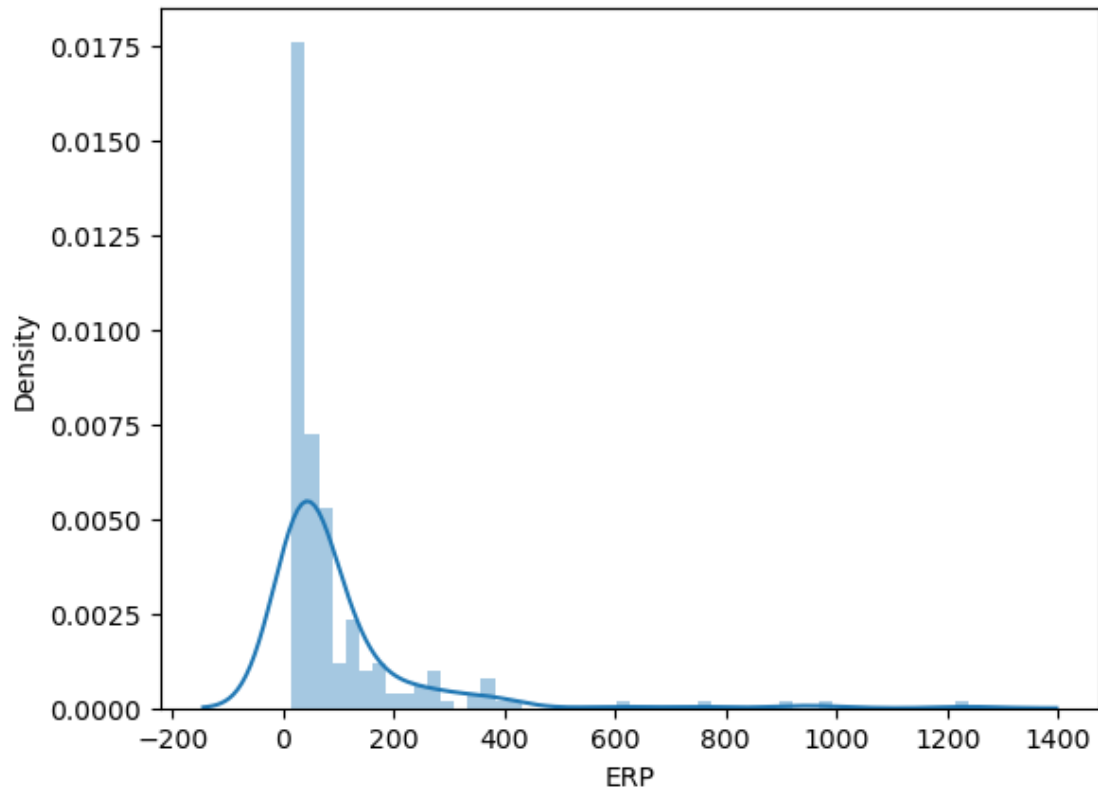
CHMAX
3.570045822323458



PRP
3.864819547035671



ERP
4.273072105948175



```
[158]: df.corr().style.background_gradient()
```

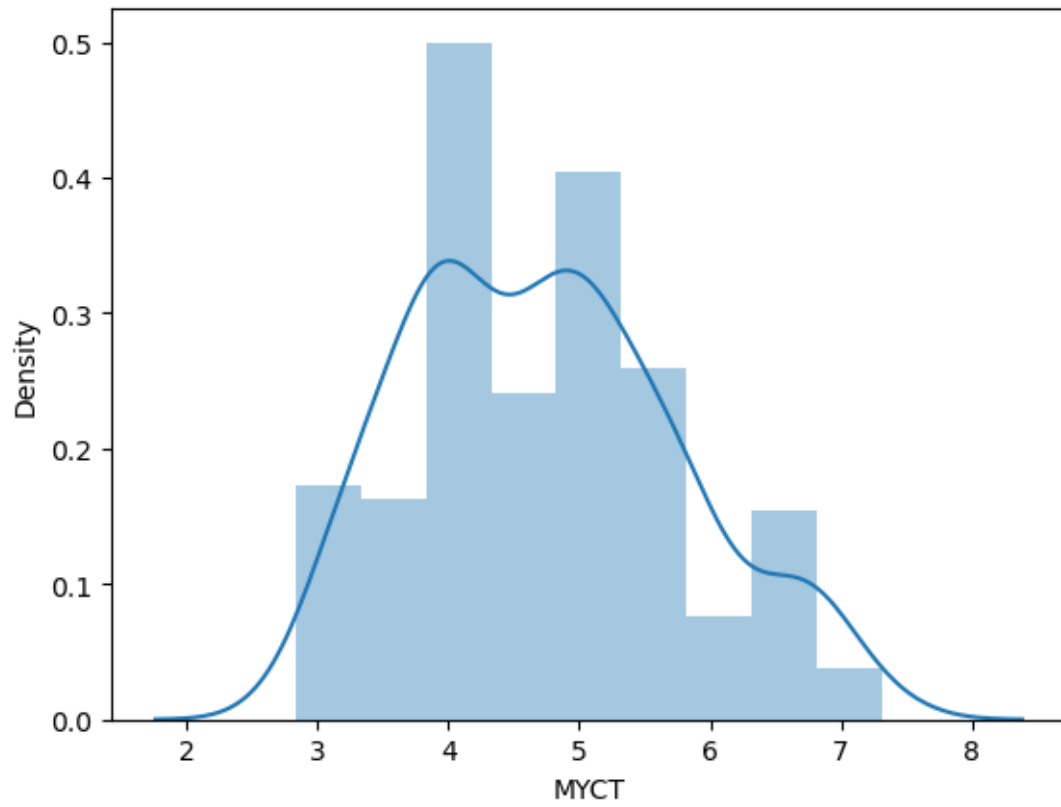
```
[158]: <pandas.io.formats.style.Styler at 0x2428479b910>
```

```
[160]: df[['MYCT', 'CACH', 'CHMIN', 'CHMAX']] = np.log(df[['MYCT', 'CACH', 'CHMIN', 'CHMAX']])
```

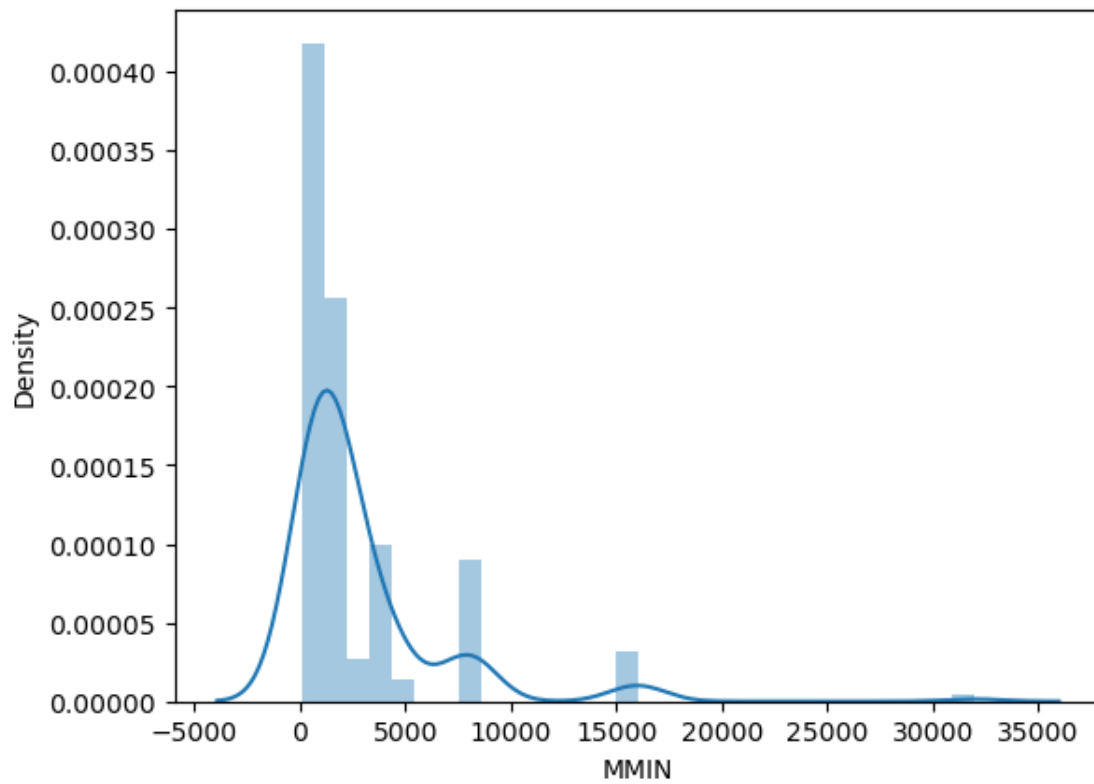
```
[168]: for i in df[colname]:
        print(i)
        print(skew(df[i]))
        sns.distplot(df[i])
        plt.show()
```

MYCT

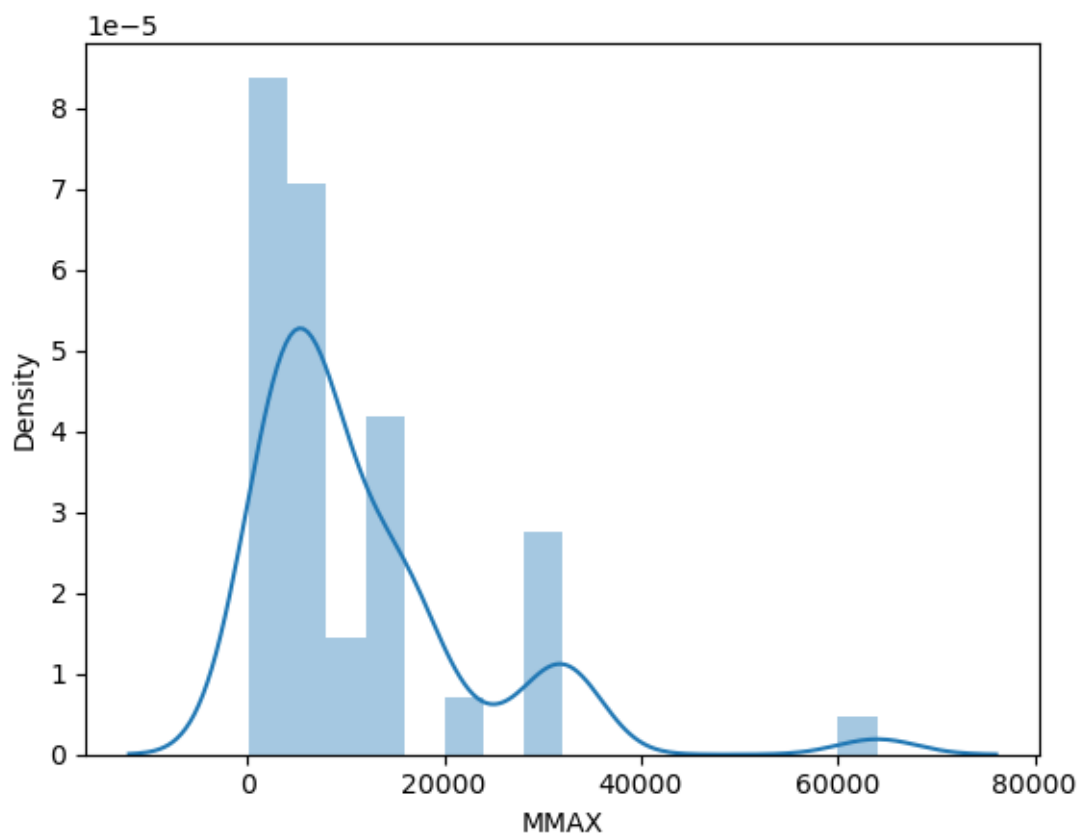
0.39497709484613314



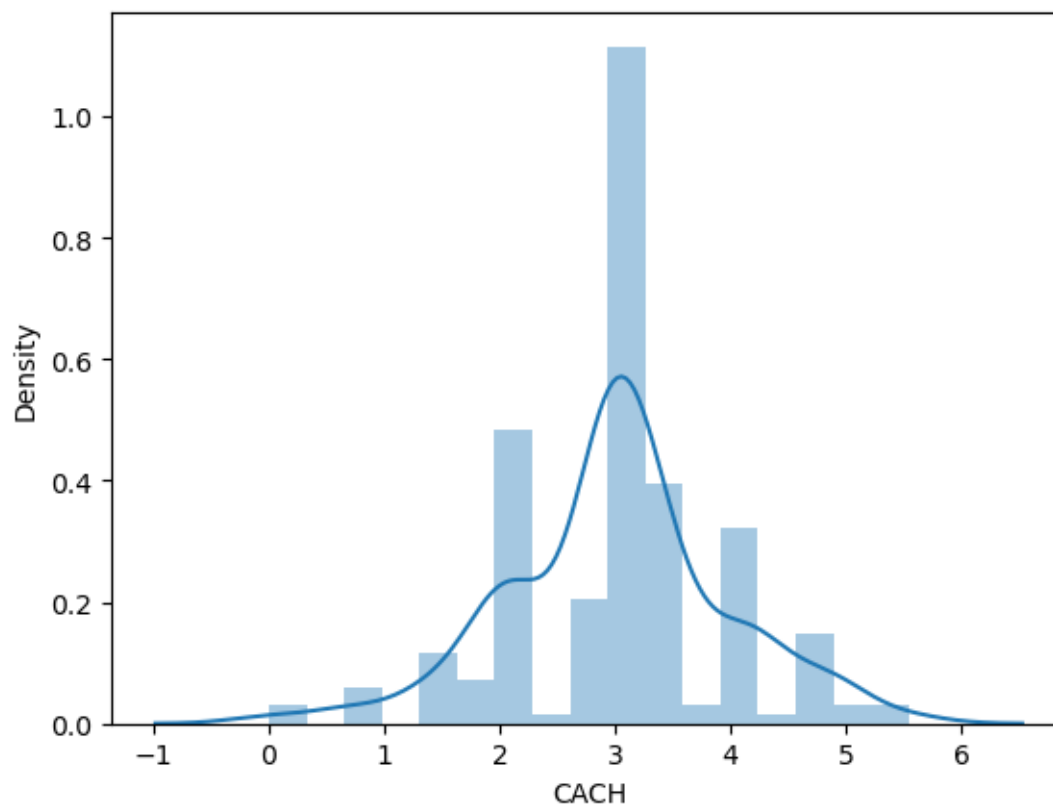
MMIN
3.4906489998203925



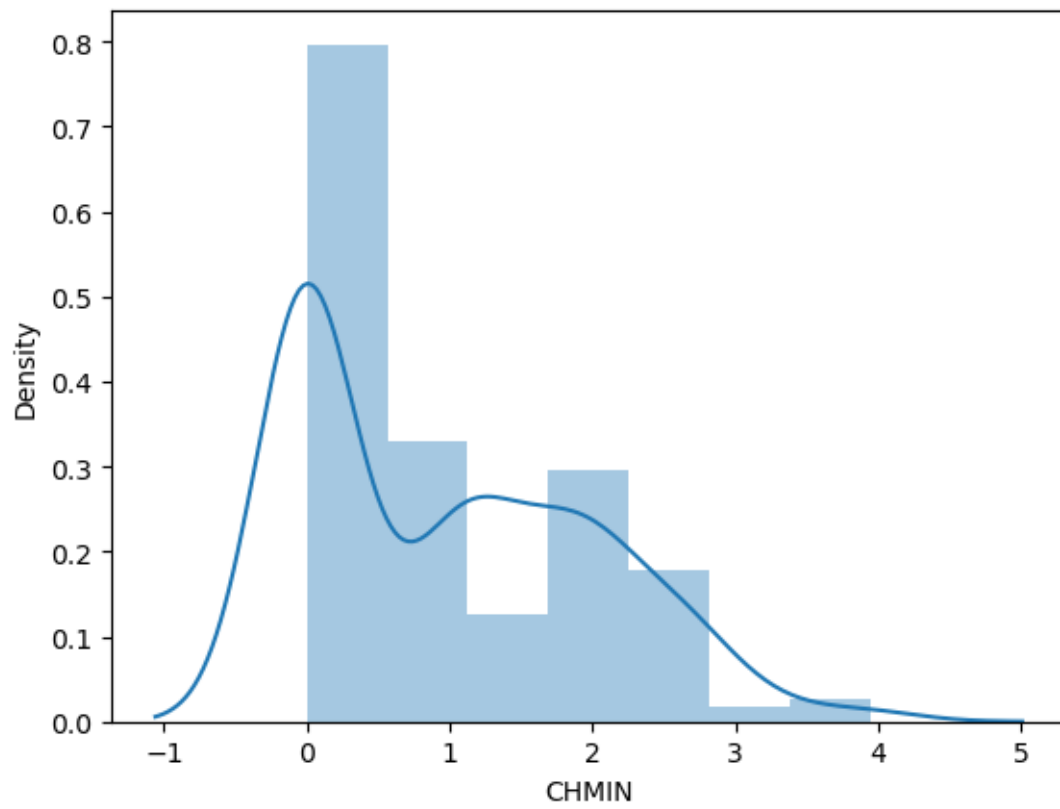
MMAX
2.1252682972972194



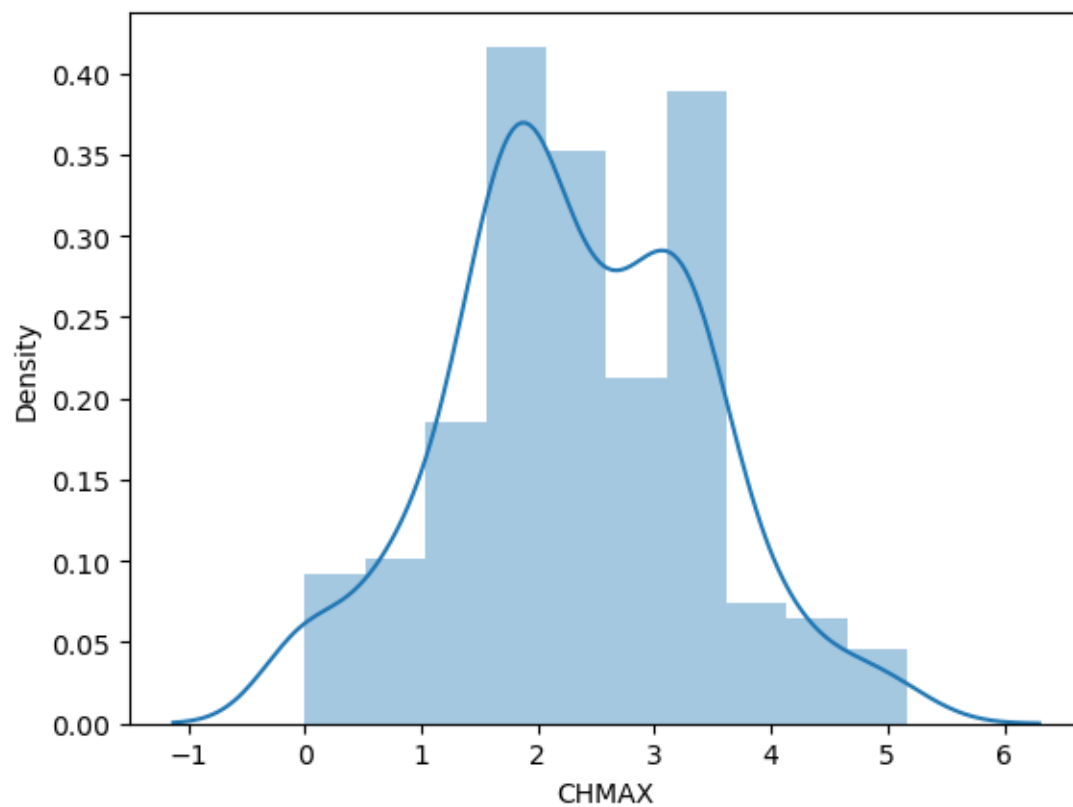
CACH
-0.1401785144584389



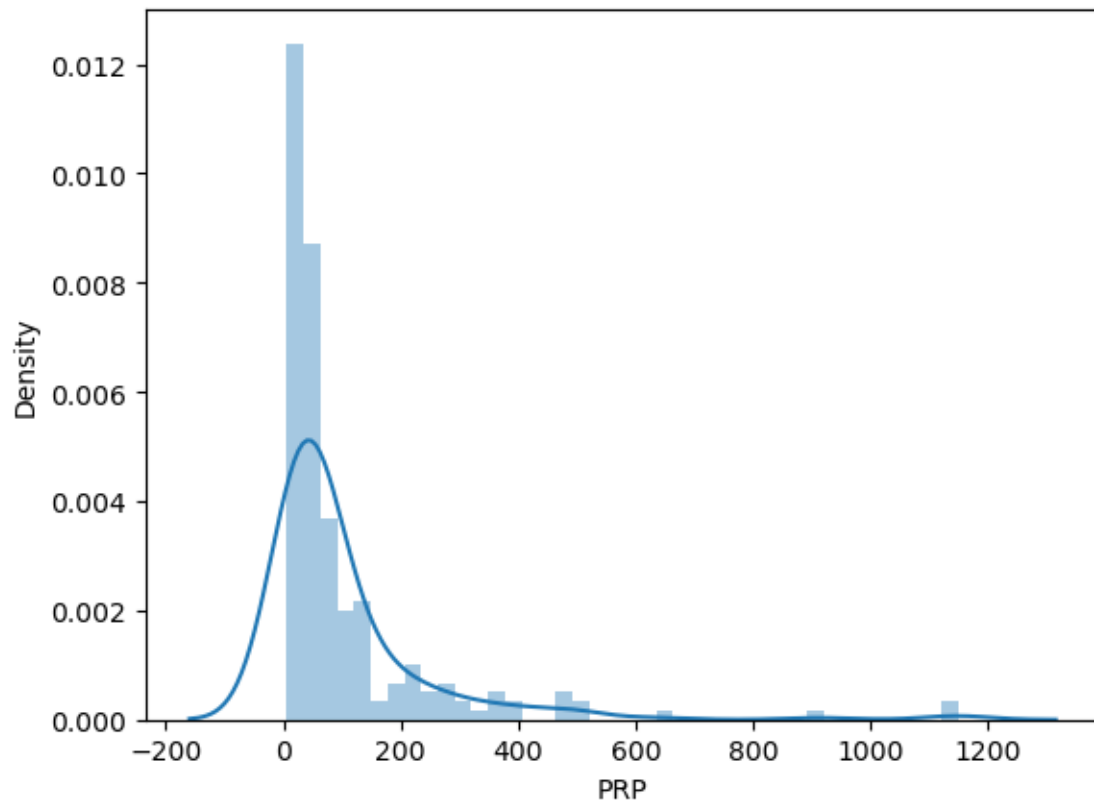
CHMIN
0.6142090540769856



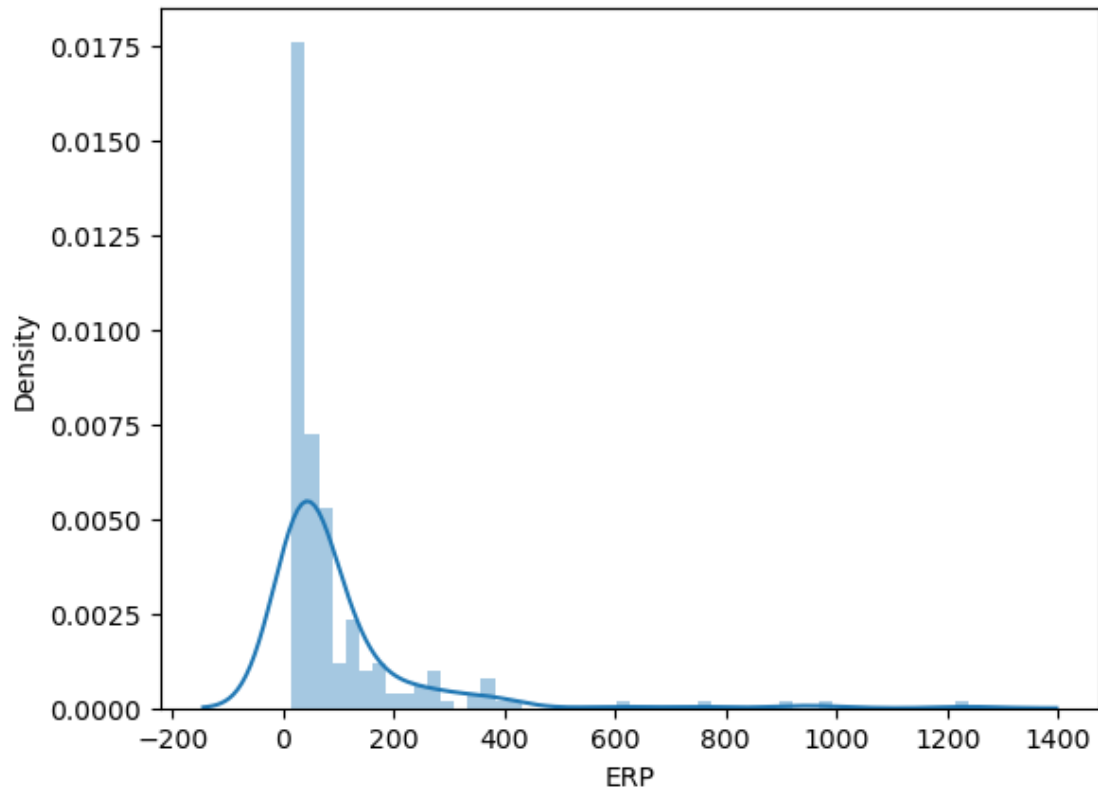
CHMAX
0.059470997616673085



PRP
3.864819547035671



ERP
4.273072105948175



```
[169]: df.head()
```

```
[169]:
```

	Vendors	Model	MYCT	MMIN	MMAx	CACH	CHMIN	CHMAX	PRP	\
0	adviser	32/60	4.828314	256	6000	5.545177	2.772589	4.852030	198	
1	amdahl	470v/7	3.367296	8000	32000	3.465736	2.079442	3.465736	269	
2	amdahl	470v/7a	3.367296	8000	32000	3.465736	2.079442	3.465736	220	
3	amdahl	470v/7b	3.367296	8000	32000	3.465736	2.079442	3.465736	172	
4	amdahl	470v/7c	3.367296	8000	16000	3.465736	2.079442	2.772589	132	

```

ERP
0 199
1 253
2 253
3 253
4 132

```

```
[170]: catcol=df.select_dtypes(['object']).columns
```

```
[171]: df[catcol]
```

```
[171]:
```

	Vendors	Model
0	adviser	32/60
1	amdahl	470v/7
2	amdahl	470v/7a
3	amdahl	470v/7b
4	amdahl	470v/7c
..
204	sperry	80/8
205	sperry	90/80-model-3
206	sratus	32
207	wang	vs-100
208	wang	vs-90

[209 rows x 2 columns]

```
[172]: x=df.iloc[:, :-1]
       y=df.iloc[:, -1]
```

```
[173]: from sklearn.preprocessing import OrdinalEncoder #encoding the categorical value
```

```
[174]: oe=OrdinalEncoder()
```

```
[175]: x[catcol]=oe.fit_transform(x[catcol])
```

```
[176]: x
```

```
[176]:
```

	Vendors	Model	MYCT	MMIN	MMAX	CACH	CHMIN	CHMAX	PRP
0	0.0	29.0	4.828314	256	6000	5.545177	2.772589	4.852030	198
1	1.0	62.0	3.367296	8000	32000	3.465736	2.079442	3.465736	269
2	1.0	63.0	3.367296	8000	32000	3.465736	2.079442	3.465736	220
3	1.0	64.0	3.367296	8000	32000	3.465736	2.079442	3.465736	172
4	1.0	65.0	3.367296	8000	16000	3.465736	2.079442	2.772589	132
..
204	27.0	100.0	4.820282	1000	8000	3.012837	0.000000	2.079442	42
205	27.0	109.0	4.584967	1000	8000	3.465736	0.693147	2.079442	46
206	28.0	28.0	4.828314	2000	8000	3.012837	0.693147	2.639057	52
207	29.0	207.0	6.173786	512	8000	3.465736	0.971494	2.317956	67
208	29.0	208.0	6.173786	1000	4000	3.012837	0.971494	2.317956	45

[209 rows x 9 columns]

```
[177]: y
```

```
[177]:
```

0	199
1	253
2	253
3	253
4	132

```

204     37
205     50
206     41
207     47
208     25
Name: ERP, Length: 209, dtype: int64

```

2 training the model

```
[178]: from sklearn.model_selection import train_test_split
```

```
[179]: xtrain,xtest,ytrain,ytest=train_test_split(x,y,test_size=0.3,random_state=1)
```

```
[180]: xtrain
```

```
[180]:
```

	Vendors	Model	MYCT	MMIN	MMAx	CACH	CHMIN	CHMAX	PRP
56	10.0	169.0	5.393628	1000	8000	2.772589	0.000000	0.693147	71
179	25.0	70.0	5.075174	1000	8000	2.772589	0.000000	2.639057	60
55	10.0	168.0	4.700480	1000	12000	2.772589	0.000000	0.693147	60
84	15.0	154.0	5.799093	1000	4000	3.012837	1.098612	1.791759	22
53	10.0	166.0	5.298317	1000	8000	3.012837	0.000000	0.693147	36
..
203	27.0	99.0	5.192957	512	4000	3.012837	0.000000	1.098612	21
137	20.0	185.0	5.010635	512	4000	3.012837	2.079442	4.852030	30
72	16.0	21.0	5.164786	256	2000	3.012837	1.098612	3.178054	22
140	21.0	112.0	4.521789	2000	8000	3.465736	0.000000	1.791759	62
37	8.0	183.0	3.912023	1000	4000	2.079442	0.000000	1.609438	29

```
[146 rows x 9 columns]
```

```
[181]: ytrain
```

```
[181]:
```

56	42
179	43
55	56
84	25
53	36
..	..
203	24
137	33
72	20
140	53
37	29

```

Name: ERP, Length: 146, dtype: int64

```

```
[182]: from sklearn.linear_model import LinearRegression
```

```
[183]: linreg=LinearRegression()
```

```
[184]: linreg.fit(xtrain,ytrain)
```

```
[184]: LinearRegression()
```

```
[185]: ypred=linreg.predict(xtest)
```

```
[186]: ypred
```

```
[186]: array([ 71.09881595,  89.76422957,  25.41421396,  18.15197405,
          211.85664161, 187.33709124,  59.8855257 , 150.28876567,
          199.72471908, 150.22897457,  12.52301752, 139.67596735,
           40.37717789,   7.83430363,  89.25287228, 220.45400742,
           14.56873135,  13.07623384,  50.77784574, 140.56686816,
           36.86734133, 227.31498795,  60.77895385,  14.87916165,
           15.83328941,  41.02384623,  24.10778208, 113.41823494,
           20.79010846, 211.96629983,  14.62318372,  15.87276436,
            8.86369055,  56.86932004, -2.47867215, 469.47775874,
          207.47294864,  32.4151343 ,  16.22207988, 372.94130696,
          372.12033212,  52.95061996,  36.60098724,  86.52828754,
          129.46335762, 288.8646765 ,  91.41897074,  22.13620622,
           30.99243953, -4.46405754,  20.24559101,  11.12082273,
           11.81853063,  59.47151299, -12.79083495,  56.71994703,
            1.45008137,  94.76459068,  50.48813859,  53.43457939,
           -5.03319843, 343.15805823,  39.46865836])
```

3 evaluating the model

```
[187]: from sklearn.metrics import r2_score
```

```
[216]: r2=r2_score(ytest,ypred)
       print(f'accuracy:{r2}')
```

```
accuracy:0.8022673023063933
```

```
[217]: train=linreg.score(xtrain,ytrain)
       test=linreg.score(xtest,ytest)
       print(train)
       print(test)
```

```
0.9704973330876837
0.8022673023063933
```

```
[190]: from sklearn.linear_model import Ridge,Lasso
```

```
[215]: l2=Ridge(alpha=55)
       l2.fit(xtrain,ytrain)
```

```
train=l2.score(xtrain,ytrain)
test=l2.score(xtest,ytest)
print(train)
print(test)
```

0.9700338483945639

0.801244795002553

```
[211]: l1=Lasso(alpha=1100)
l1.fit(xtrain,ytrain)
train=l1.score(xtrain,ytrain)
test=l1.score(xtest,ytest)
print(train)
print(test)
```

0.9609286292482392

0.8031233558099401