

breast cancer

March 19, 2023

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

```
[2]: df=pd.read_csv('cancer.csv')
```

Attribute Information:

- 1) ID number
- 2) Diagnosis (M = malignant, B = benign) 3-32)

Ten real-valued features are computed for each cell nucleus:

- a) radius (mean of distances from center to points on the perimeter)
- b) texture (standard deviation of gray-scale values)
- c) perimeter
- d) area
- e) smoothness (local variation in radius lengths)
- f) compactness ($\text{perimeter}^2 / \text{area} - 1.0$)
- g) concavity (severity of concave portions of the contour)
- h) concave points (number of concave portions of the contour)
- i) symmetry
- j) fractal dimension ("coastline approximation" - 1)

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

```
[3]:      id diagnosis  radius_mean  texture_mean  perimeter_mean  area_mean  \
0    842302         M        17.99         10.38          122.80      1001.0
1    842517         M        20.57         17.77          132.90      1326.0
2  84300903         M        19.69         21.25          130.00      1203.0
3  84348301         M        11.42         20.38           77.58       386.1
4  84358402         M        20.29         14.34          135.10      1297.0

      smoothness_mean  compactness_mean  concavity_mean  concave points_mean  \
0          0.11840         0.27760         0.3001         0.14710
1          0.08474         0.07864         0.0869         0.07017
2          0.10960         0.15990         0.1974         0.12790
```

3	0.14250	0.28390	0.2414	0.10520
4	0.10030	0.13280	0.1980	0.10430

	texture_worst	perimeter_worst	area_worst	smoothness_worst	\
0	17.33	184.60	2019.0	0.1622	
1	23.41	158.80	1956.0	0.1238	
2	25.53	152.50	1709.0	0.1444	
3	26.50	98.87	567.7	0.2098	
4	16.67	152.20	1575.0	0.1374	

	compactness_worst	concavity_worst	concave points_worst	symmetry_worst	\
0	0.6656	0.7119	0.2654	0.4601	
1	0.1866	0.2416	0.1860	0.2750	
2	0.4245	0.4504	0.2430	0.3613	
3	0.8663	0.6869	0.2575	0.6638	
4	0.2050	0.4000	0.1625	0.2364	

	fractal_dimension_worst	Unnamed: 32
0	0.11890	NaN
1	0.08902	NaN
2	0.08758	NaN
3	0.17300	NaN
4	0.07678	NaN

[5 rows x 33 columns]

```
[4]: df.columns
```

```
[4]: Index(['id', 'diagnosis', 'radius_mean', 'texture_mean', 'perimeter_mean',
        'area_mean', 'smoothness_mean', 'compactness_mean', 'concavity_mean',
        'concave points_mean', 'symmetry_mean', 'fractal_dimension_mean',
        'radius_se', 'texture_se', 'perimeter_se', 'area_se', 'smoothness_se',
        'compactness_se', 'concavity_se', 'concave points_se', 'symmetry_se',
        'fractal_dimension_se', 'radius_worst', 'texture_worst',
        'perimeter_worst', 'area_worst', 'smoothness_worst',
        'compactness_worst', 'concavity_worst', 'concave points_worst',
        'symmetry_worst', 'fractal_dimension_worst', 'Unnamed: 32'],
        dtype='object')
```

```
[5]: df.drop(columns='Unnamed: 32', inplace=True, axis=1)
```

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

	id	diagnosis	radius_mean	texture_mean	perimeter_mean	area_mean	\
0	842302	M	17.99	10.38	122.80	1001.0	
1	842517	M	20.57	17.77	132.90	1326.0	
2	84300903	M	19.69	21.25	130.00	1203.0	

3	84348301	M	11.42	20.38	77.58	386.1
4	84358402	M	20.29	14.34	135.10	1297.0

	smoothness_mean	compactness_mean	concavity_mean	concave points_mean	\
0	0.11840	0.27760	0.3001	0.14710	
1	0.08474	0.07864	0.0869	0.07017	
2	0.10960	0.15990	0.1974	0.12790	
3	0.14250	0.28390	0.2414	0.10520	
4	0.10030	0.13280	0.1980	0.10430	

...	radius_worst	texture_worst	perimeter_worst	area_worst	\
0	25.38	17.33	184.60	2019.0	
1	24.99	23.41	158.80	1956.0	
2	23.57	25.53	152.50	1709.0	
3	14.91	26.50	98.87	567.7	
4	22.54	16.67	152.20	1575.0	

	smoothness_worst	compactness_worst	concavity_worst	concave points_worst	\
0	0.1622	0.6656	0.7119	0.2654	
1	0.1238	0.1866	0.2416	0.1860	
2	0.1444	0.4245	0.4504	0.2430	
3	0.2098	0.8663	0.6869	0.2575	
4	0.1374	0.2050	0.4000	0.1625	

	symmetry_worst	fractal_dimension_worst
0	0.4601	0.11890
1	0.2750	0.08902
2	0.3613	0.08758
3	0.6638	0.17300
4	0.2364	0.07678

[5 rows x 32 columns]

```
[7]: df.shape
```

```
[7]: (569, 32)
```

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

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 569 entries, 0 to 568
Data columns (total 32 columns):
#   Column                Non-Null Count  Dtype
---  -
0   id                    569 non-null    int64
1   diagnosis             569 non-null    object
2   radius_mean           569 non-null    float64
3   texture_mean          569 non-null    float64
```

```

4  perimeter_mean      569 non-null    float64
5  area_mean           569 non-null    float64
6  smoothness_mean     569 non-null    float64
7  compactness_mean    569 non-null    float64
8  concavity_mean      569 non-null    float64
9  concave points_mean  569 non-null    float64
10 symmetry_mean       569 non-null    float64
11 fractal_dimension_mean 569 non-null    float64
12 radius_se           569 non-null    float64
13 texture_se           569 non-null    float64
14 perimeter_se         569 non-null    float64
15 area_se              569 non-null    float64
16 smoothness_se        569 non-null    float64
17 compactness_se       569 non-null    float64
18 concavity_se         569 non-null    float64
19 concave points_se    569 non-null    float64
20 symmetry_se          569 non-null    float64
21 fractal_dimension_se  569 non-null    float64
22 radius_worst         569 non-null    float64
23 texture_worst        569 non-null    float64
24 perimeter_worst      569 non-null    float64
25 area_worst           569 non-null    float64
26 smoothness_worst     569 non-null    float64
27 compactness_worst    569 non-null    float64
28 concavity_worst      569 non-null    float64
29 concave points_worst  569 non-null    float64
30 symmetry_worst       569 non-null    float64
31 fractal_dimension_worst 569 non-null    float64

```

dtypes: float64(30), int64(1), object(1)

memory usage: 142.4+ KB

```
[9]: df.drop(columns= 'id', inplace=True, axis=1)
```

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

```

[10]:  diagnosis  radius_mean  texture_mean  perimeter_mean  area_mean  \
0         M         17.99         10.38         122.80        1001.0
1         M         20.57         17.77         132.90        1326.0
2         M         19.69         21.25         130.00        1203.0
3         M         11.42         20.38          77.58         386.1
4         M         20.29         14.34         135.10        1297.0

      smoothness_mean  compactness_mean  concavity_mean  concave points_mean  \
0          0.11840         0.27760         0.3001         0.14710
1          0.08474         0.07864         0.0869         0.07017
2          0.10960         0.15990         0.1974         0.12790
3          0.14250         0.28390         0.2414         0.10520

```

4	0.10030	0.13280	0.1980	0.10430
---	---------	---------	--------	---------

	symmetry_mean	...	radius_worst	texture_worst	perimeter_worst	\
0	0.2419	...	25.38	17.33	184.60	
1	0.1812	...	24.99	23.41	158.80	
2	0.2069	...	23.57	25.53	152.50	
3	0.2597	...	14.91	26.50	98.87	
4	0.1809	...	22.54	16.67	152.20	

	area_worst	smoothness_worst	compactness_worst	concavity_worst	\
0	2019.0	0.1622	0.6656	0.7119	
1	1956.0	0.1238	0.1866	0.2416	
2	1709.0	0.1444	0.4245	0.4504	
3	567.7	0.2098	0.8663	0.6869	
4	1575.0	0.1374	0.2050	0.4000	

	concave points_worst	symmetry_worst	fractal_dimension_worst
0	0.2654	0.4601	0.11890
1	0.1860	0.2750	0.08902
2	0.2430	0.3613	0.08758
3	0.2575	0.6638	0.17300
4	0.1625	0.2364	0.07678

[5 rows x 31 columns]

```
[11]: df.isnull().sum()
```

```
[11]: diagnosis      0
radius_mean        0
texture_mean       0
perimeter_mean     0
area_mean          0
smoothness_mean    0
compactness_mean   0
concavity_mean     0
concave points_mean 0
symmetry_mean      0
fractal_dimension_mean 0
radius_se          0
texture_se         0
perimeter_se       0
area_se            0
smoothness_se      0
compactness_se     0
concavity_se       0
concave points_se  0
symmetry_se        0
```

```

fractal_dimension_se      0
radius_worst              0
texture_worst             0
perimeter_worst          0
area_worst               0
smoothness_worst         0
compactness_worst        0
concavity_worst          0
concave points_worst     0
symmetry_worst           0
fractal_dimension_worst  0
dtype: int64

```

```
[12]: df.describe()
```

```

[12]:      radius_mean  texture_mean  perimeter_mean  area_mean  \
count    569.000000    569.000000    569.000000    569.000000
mean      14.127292     19.289649     91.969033    654.889104
std        3.524049     4.301036     24.298981    351.914129
min        6.981000     9.710000     43.790000    143.500000
25%       11.700000    16.170000     75.170000    420.300000
50%       13.370000    18.840000     86.240000    551.100000
75%       15.780000    21.800000    104.100000    782.700000
max       28.110000    39.280000    188.500000   2501.000000

      smoothness_mean  compactness_mean  concavity_mean  concave points_mean  \
count    569.000000    569.000000    569.000000    569.000000
mean        0.096360     0.104341     0.088799     0.048919
std         0.014064     0.052813     0.079720     0.038803
min         0.052630     0.019380     0.000000     0.000000
25%         0.086370     0.064920     0.029560     0.020310
50%         0.095870     0.092630     0.061540     0.033500
75%         0.105300     0.130400     0.130700     0.074000
max         0.163400     0.345400     0.426800     0.201200

      symmetry_mean  fractal_dimension_mean  ...  radius_worst  \
count    569.000000    569.000000  ...    569.000000
mean        0.181162         0.062798  ...     16.269190
std         0.027414         0.007060  ...     4.833242
min         0.106000         0.049960  ...     7.930000
25%         0.161900         0.057700  ...    13.010000
50%         0.179200         0.061540  ...    14.970000
75%         0.195700         0.066120  ...    18.790000
max         0.304000         0.097440  ...    36.040000

      texture_worst  perimeter_worst  area_worst  smoothness_worst  \
count    569.000000    569.000000    569.000000    569.000000

```

mean	25.677223	107.261213	880.583128	0.132369
std	6.146258	33.602542	569.356993	0.022832
min	12.020000	50.410000	185.200000	0.071170
25%	21.080000	84.110000	515.300000	0.116600
50%	25.410000	97.660000	686.500000	0.131300
75%	29.720000	125.400000	1084.000000	0.146000
max	49.540000	251.200000	4254.000000	0.222600

	compactness_worst	concavity_worst	concave points_worst	\
count	569.000000	569.000000	569.000000	
mean	0.254265	0.272188	0.114606	
std	0.157336	0.208624	0.065732	
min	0.027290	0.000000	0.000000	
25%	0.147200	0.114500	0.064930	
50%	0.211900	0.226700	0.099930	
75%	0.339100	0.382900	0.161400	
max	1.058000	1.252000	0.291000	

	symmetry_worst	fractal_dimension_worst
count	569.000000	569.000000
mean	0.290076	0.083946
std	0.061867	0.018061
min	0.156500	0.055040
25%	0.250400	0.071460
50%	0.282200	0.080040
75%	0.317900	0.092080
max	0.663800	0.207500

[8 rows x 30 columns]

as we see median < mean so it is positive skewed data

```
[13]: df['diagnosis'].value_counts()
```

```
[13]: B    357
      M    212
      Name: diagnosis, dtype: int64
```

```
[14]: from sklearn import preprocessing
      le = preprocessing.LabelEncoder()
      labels= le.fit_transform(df['diagnosis'])
      df['Target']= labels
      df.drop(columns='diagnosis', inplace= True, axis=0)
```

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

```
[15]:   radius_mean  texture_mean  perimeter_mean  area_mean  smoothness_mean  \
0         17.99         10.38          122.80        1001.0         0.11840
```

1	20.57	17.77	132.90	1326.0	0.08474
2	19.69	21.25	130.00	1203.0	0.10960
3	11.42	20.38	77.58	386.1	0.14250
4	20.29	14.34	135.10	1297.0	0.10030

	compactness_mean	concavity_mean	concave	points_mean	symmetry_mean	\
0	0.27760	0.3001		0.14710	0.2419	
1	0.07864	0.0869		0.07017	0.1812	
2	0.15990	0.1974		0.12790	0.2069	
3	0.28390	0.2414		0.10520	0.2597	
4	0.13280	0.1980		0.10430	0.1809	

	fractal_dimension_mean	...	texture_worst	perimeter_worst	area_worst	\
0	0.07871	...	17.33	184.60	2019.0	
1	0.05667	...	23.41	158.80	1956.0	
2	0.05999	...	25.53	152.50	1709.0	
3	0.09744	...	26.50	98.87	567.7	
4	0.05883	...	16.67	152.20	1575.0	

	smoothness_worst	compactness_worst	concavity_worst	concave	points_worst	\
0	0.1622	0.6656	0.7119		0.2654	
1	0.1238	0.1866	0.2416		0.1860	
2	0.1444	0.4245	0.4504		0.2430	
3	0.2098	0.8663	0.6869		0.2575	
4	0.1374	0.2050	0.4000		0.1625	

	symmetry_worst	fractal_dimension_worst	Target
0	0.4601	0.11890	1
1	0.2750	0.08902	1
2	0.3613	0.08758	1
3	0.6638	0.17300	1
4	0.2364	0.07678	1

[5 rows x 31 columns]

```
[16]: df['Target'].value_counts()
```

```
[16]: 0    357
      1    212
      Name: Target, dtype: int64
```

```
[17]: df.groupby('Target').mean()
```

```
[17]:      radius_mean  texture_mean  perimeter_mean  area_mean  \
Target
0      12.146524    17.914762      78.075406  462.790196
1      17.462830    21.604906     115.365377  978.376415
```


	smoothness_mean	compactness_mean	concavity_mean	\
Target				
0	0.092478	0.080085	0.046058	
1	0.102898	0.145188	0.160775	

	concave points_mean	symmetry_mean	fractal_dimension_mean	...	\
Target				...	
0	0.025717	0.174186		0.062867	...
1	0.087990	0.192909		0.062680	...

	radius_worst	texture_worst	perimeter_worst	area_worst	\
Target					
0	13.379801	23.515070	87.005938	558.899440	
1	21.134811	29.318208	141.370330	1422.286321	

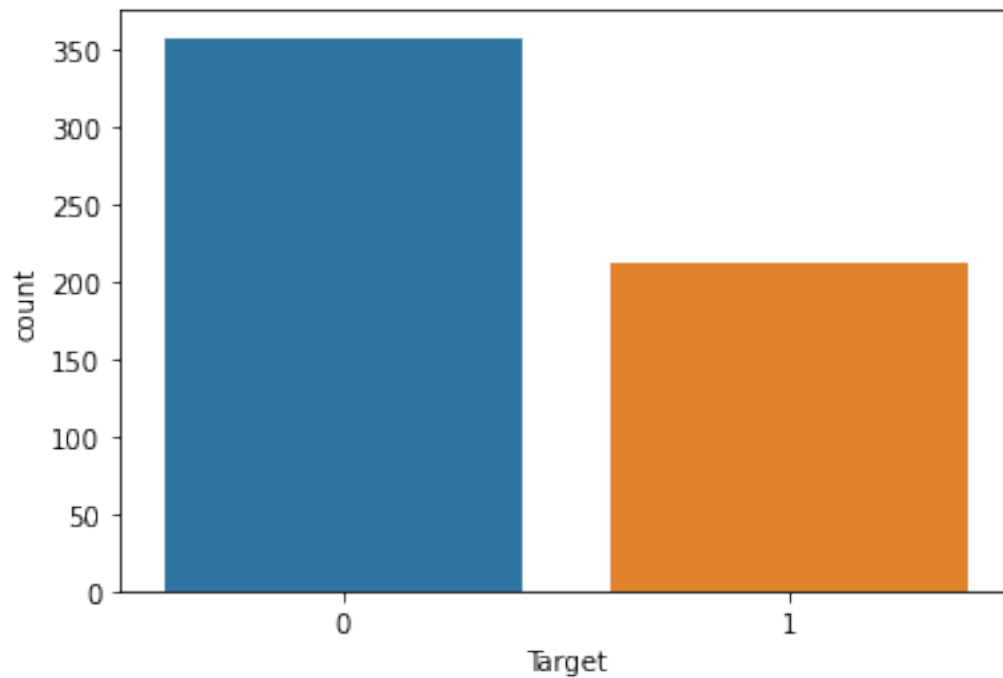
	smoothness_worst	compactness_worst	concavity_worst	\
Target				
0	0.124959	0.182673	0.166238	
1	0.144845	0.374824	0.450606	

	concave points_worst	symmetry_worst	fractal_dimension_worst
Target			
0	0.074444	0.270246	0.079442
1	0.182237	0.323468	0.091530

[2 rows x 30 columns]

```
[18]: sns.countplot(x='Target', data=df)
```

```
[18]: <AxesSubplot:xlabel='Target', ylabel='count'>
```

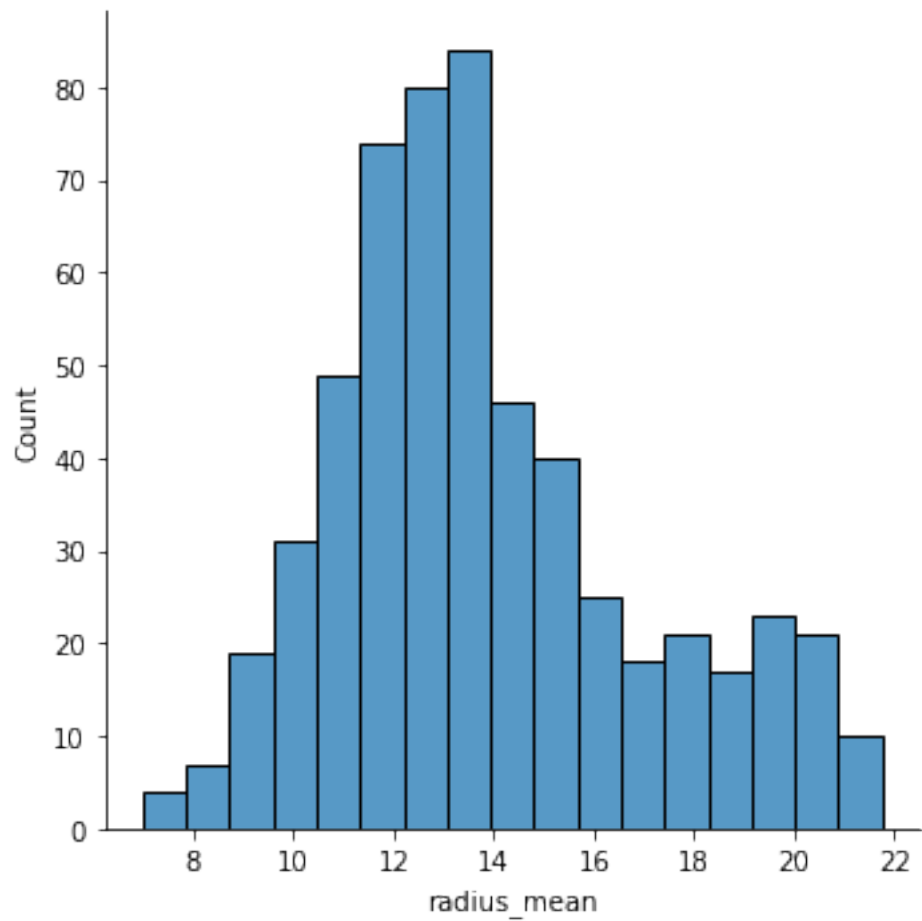


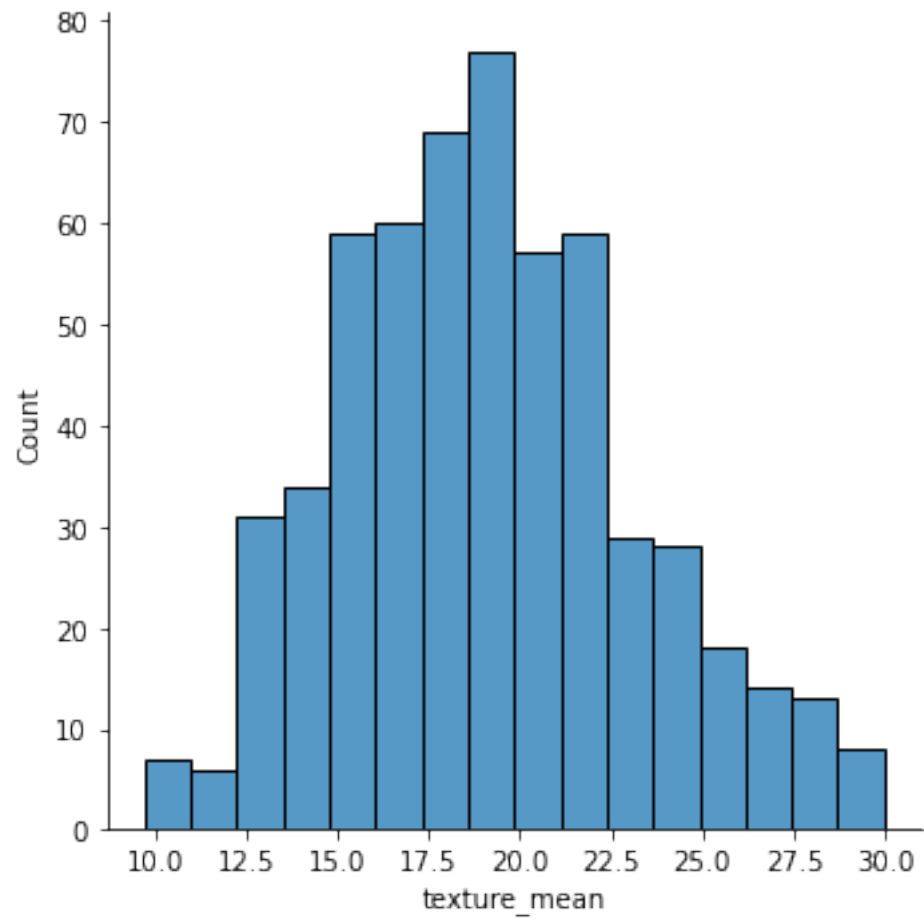
```
[19]: for column in df :  
       print(column)
```

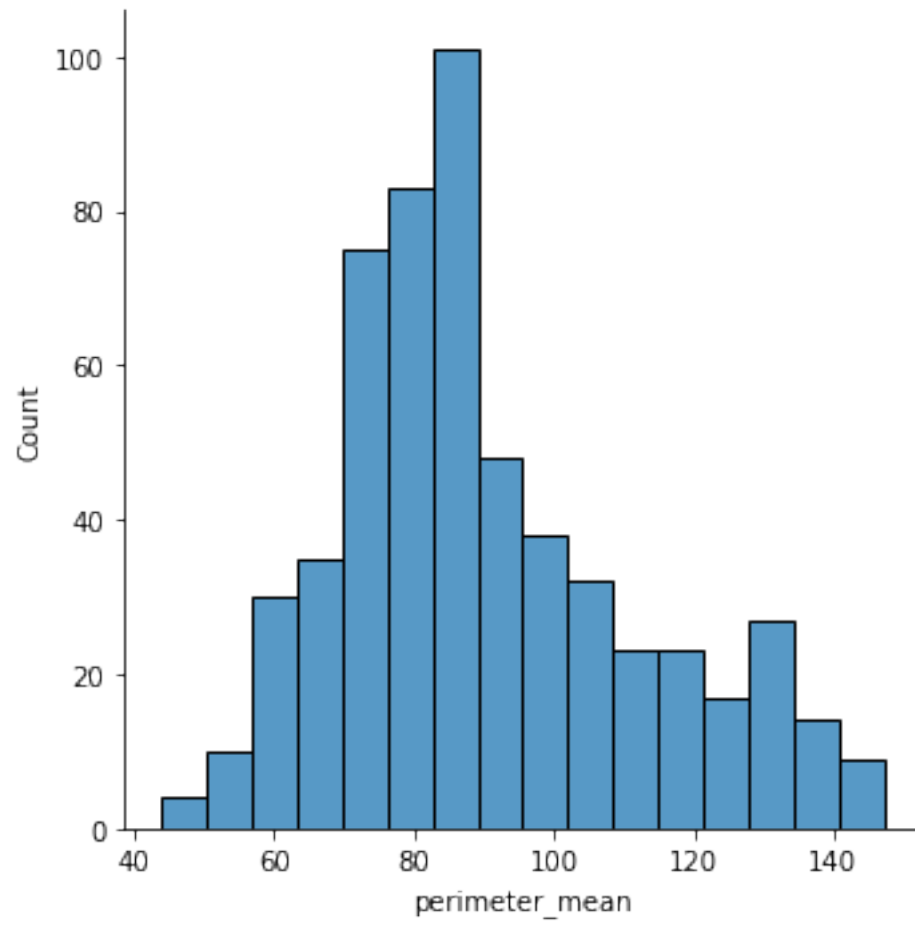
```
radius_mean  
texture_mean  
perimeter_mean  
area_mean  
smoothness_mean  
compactness_mean  
concavity_mean  
concave points_mean  
symmetry_mean  
fractal_dimension_mean  
radius_se  
texture_se  
perimeter_se  
area_se  
smoothness_se  
compactness_se  
concavity_se  
concave points_se  
symmetry_se  
fractal_dimension_se  
radius_worst  
texture_worst
```

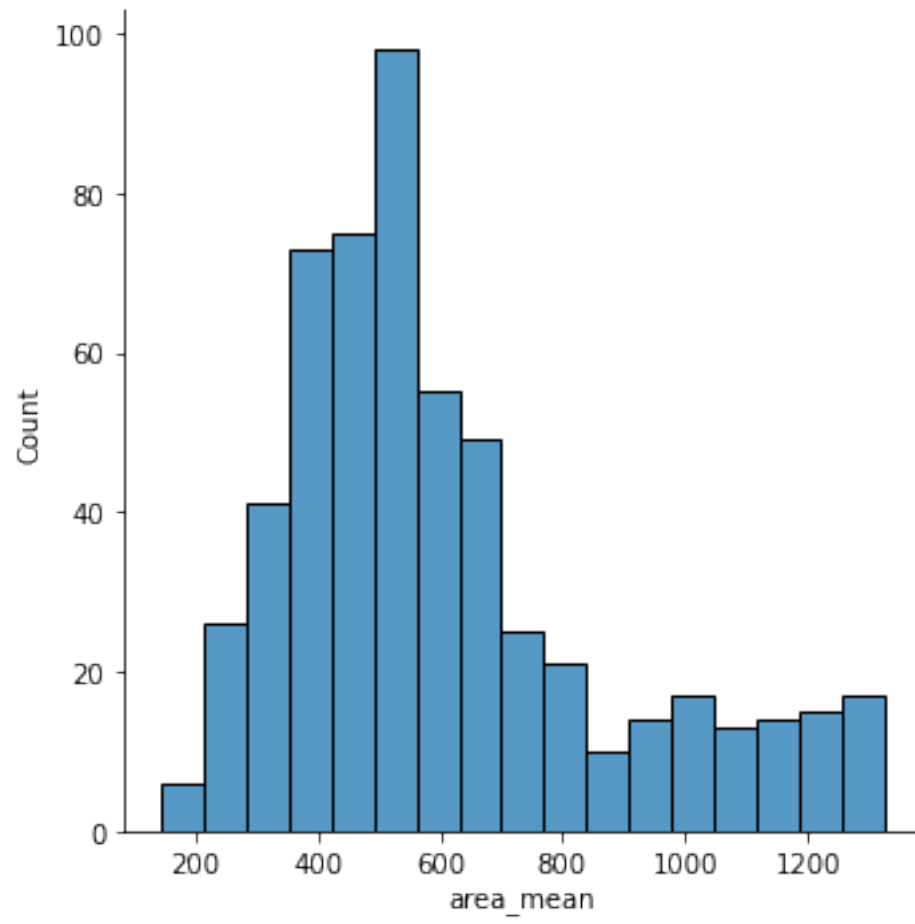
```
perimeter_worst  
area_worst  
smoothness_worst  
compactness_worst  
concavity_worst  
concave points_worst  
symmetry_worst  
fractal_dimension_worst  
Target
```

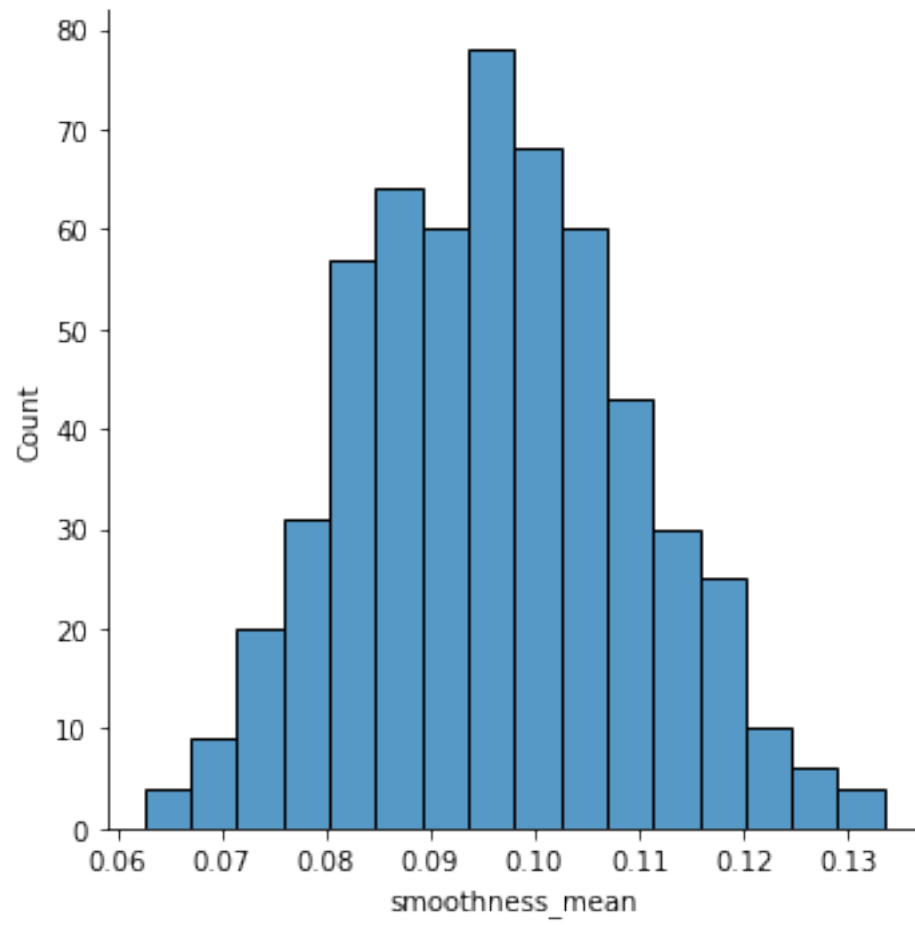
```
[44]: for column in df:  
      sns.displot(x=column, data=df)
```

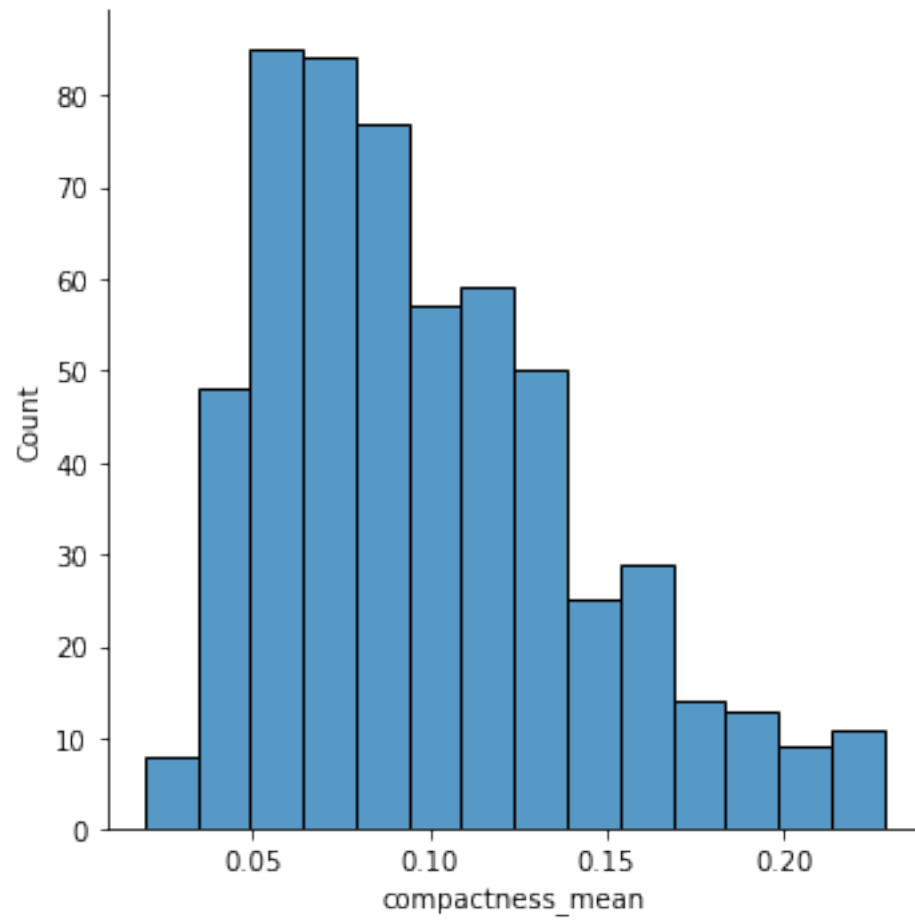


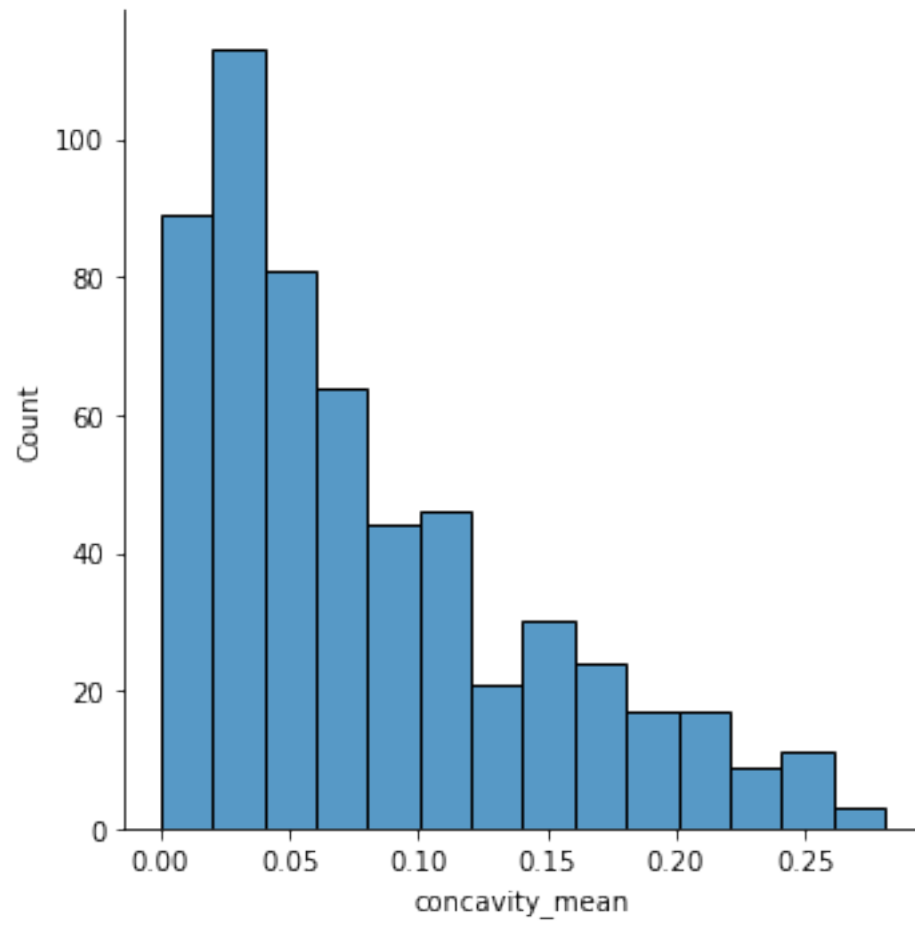


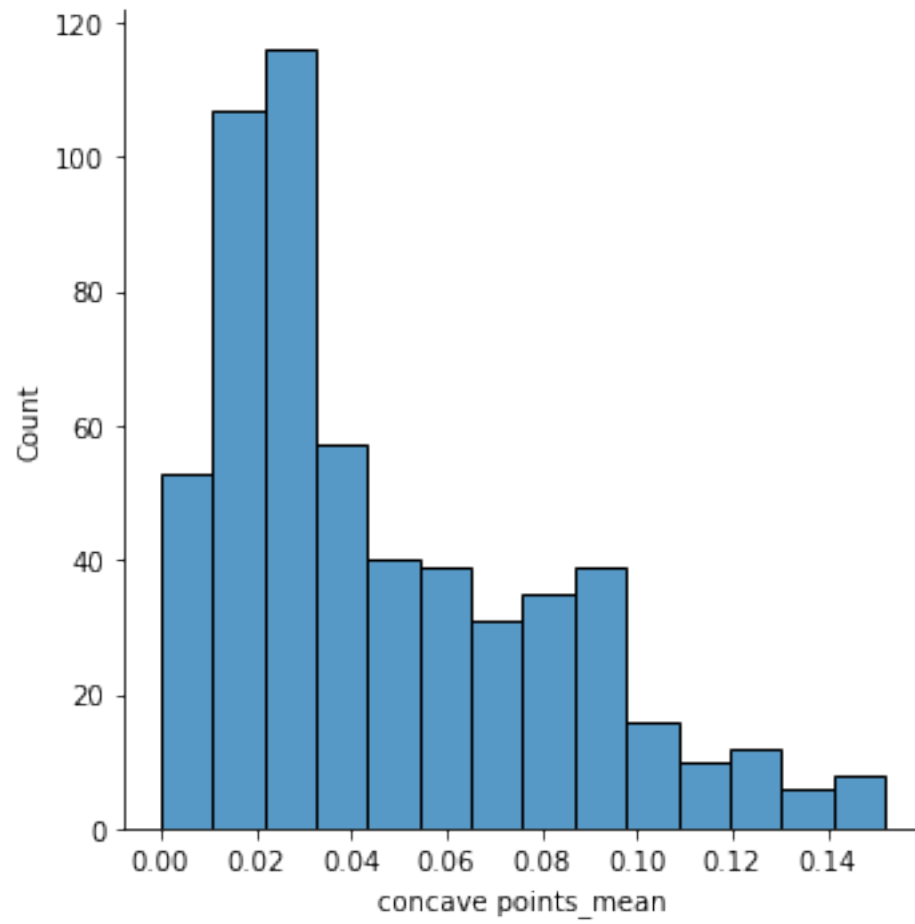


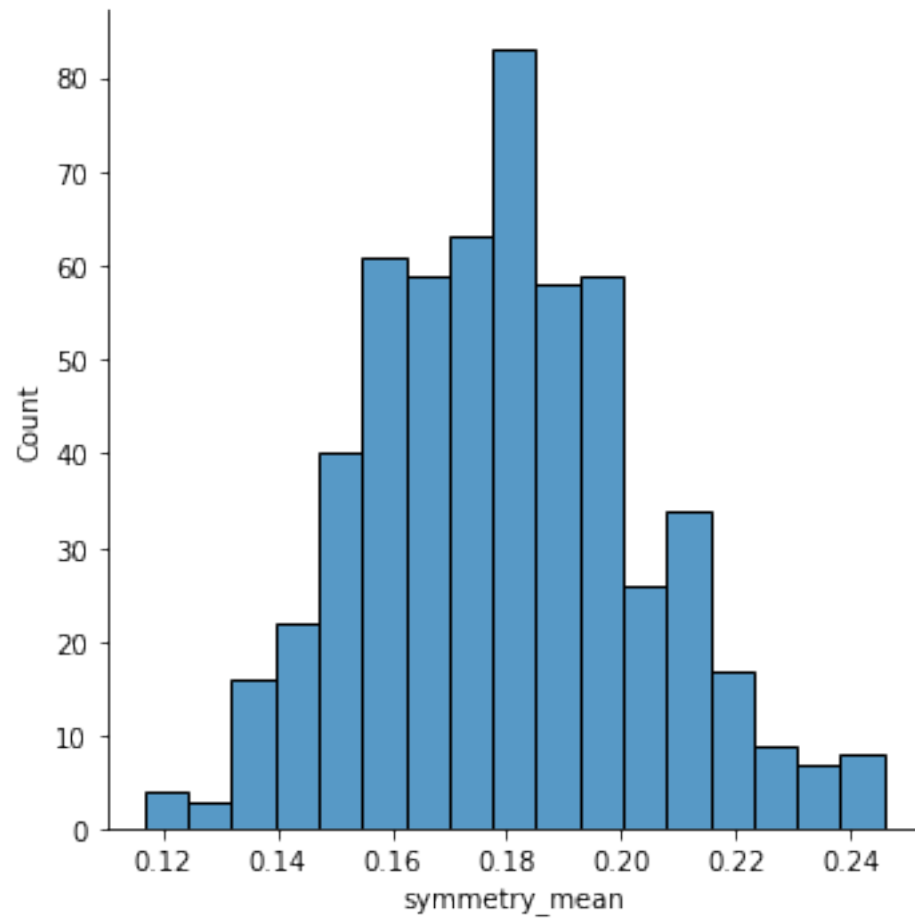


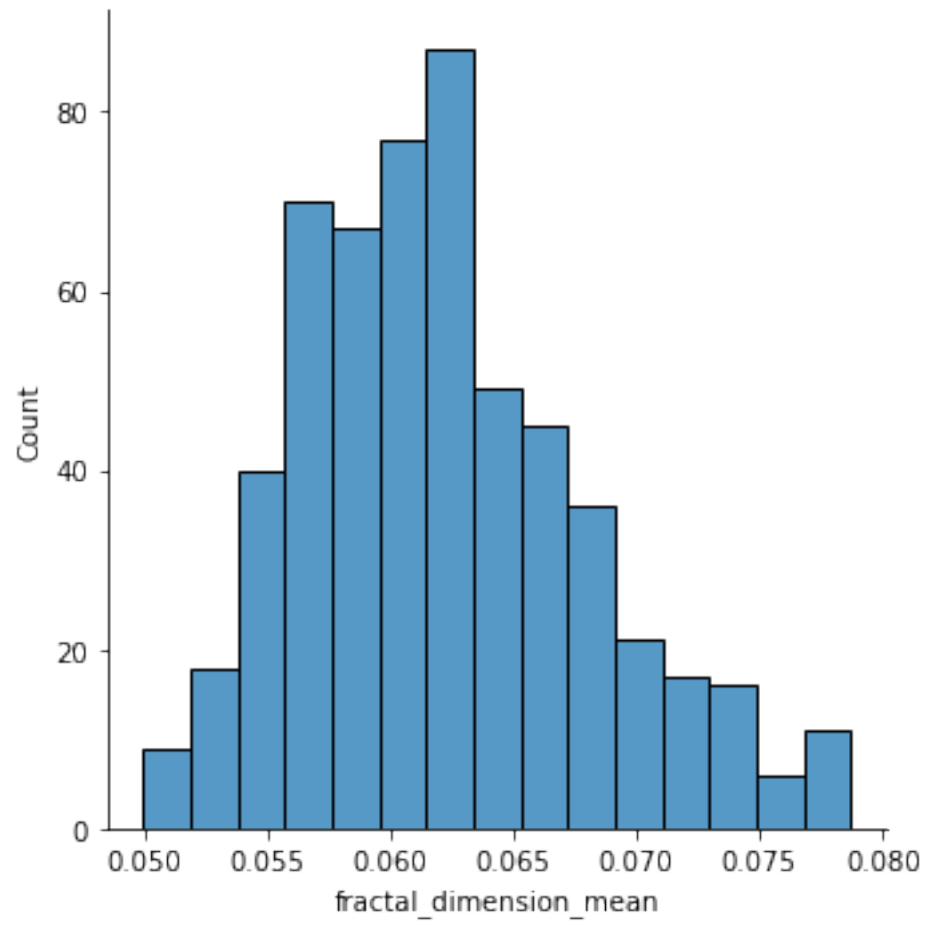


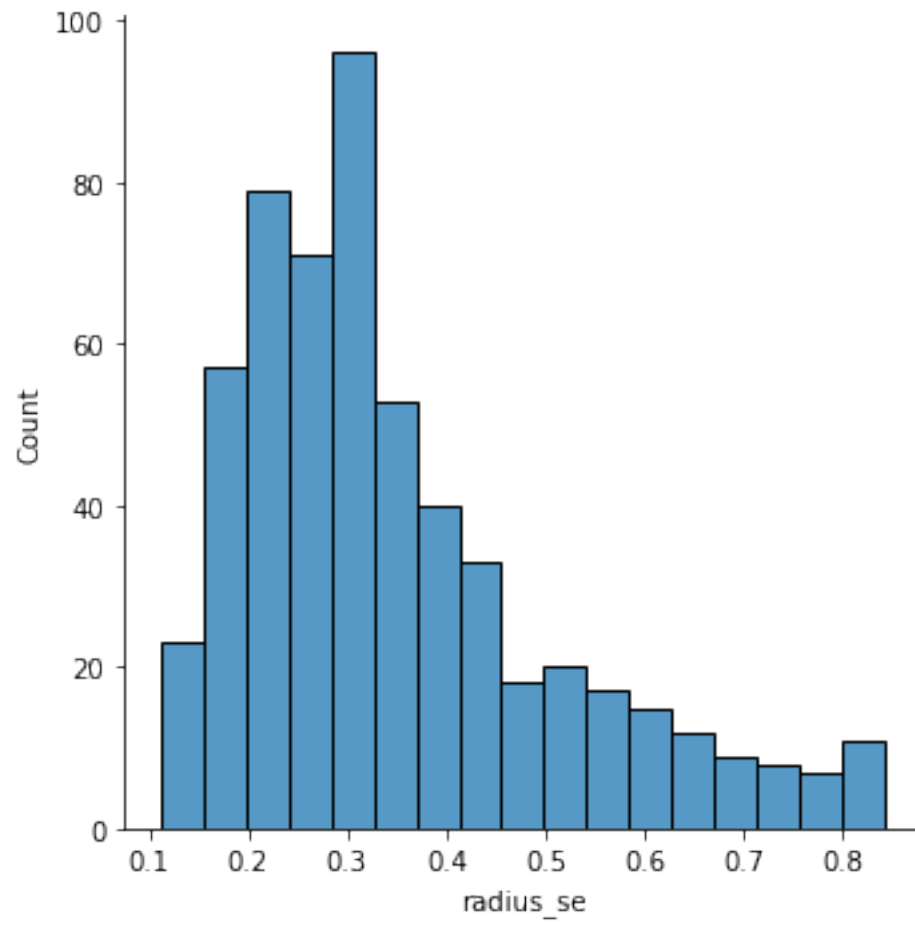


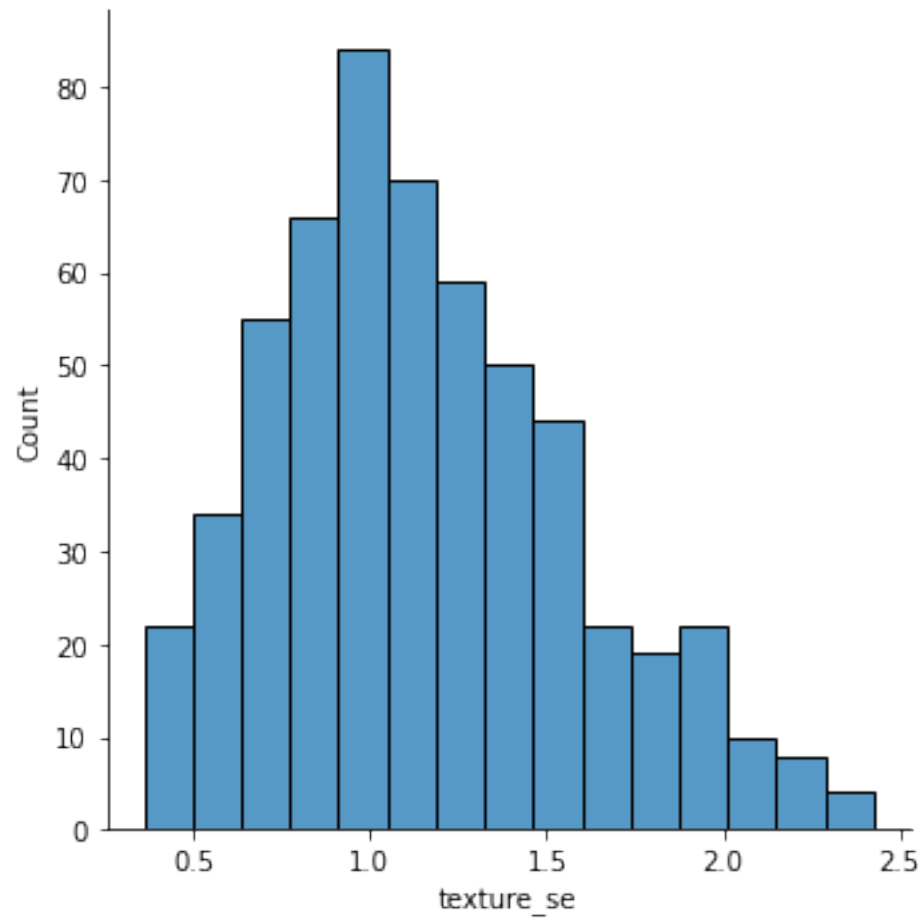


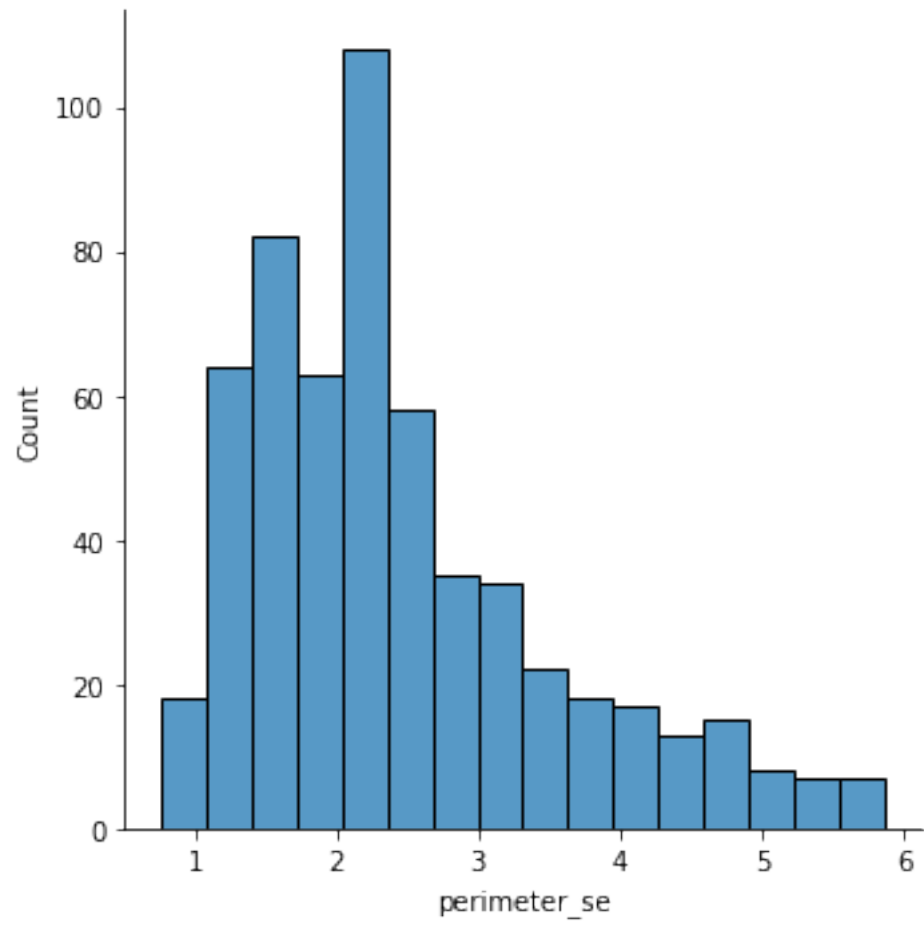


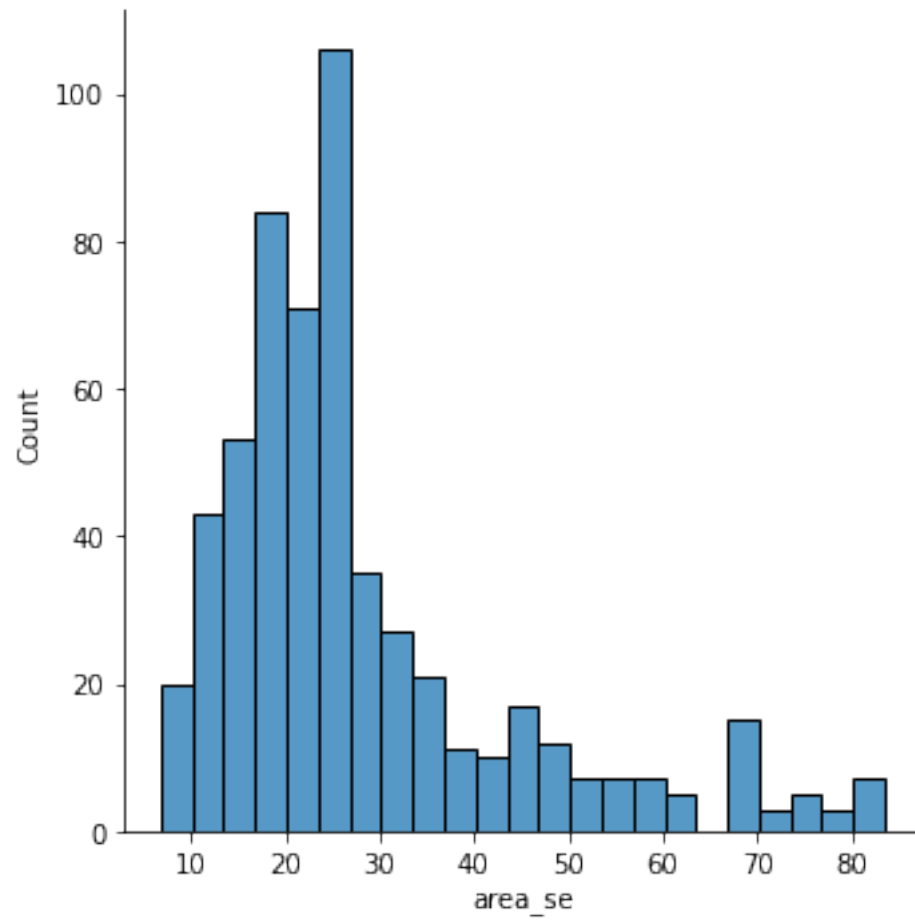


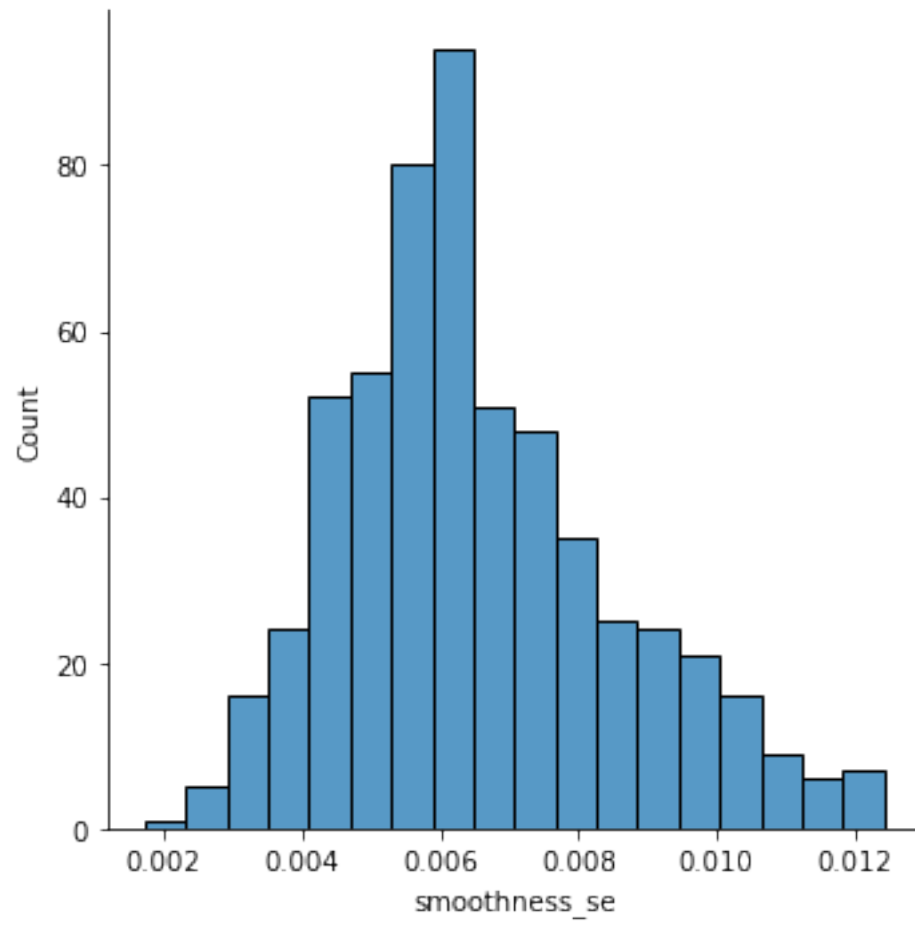


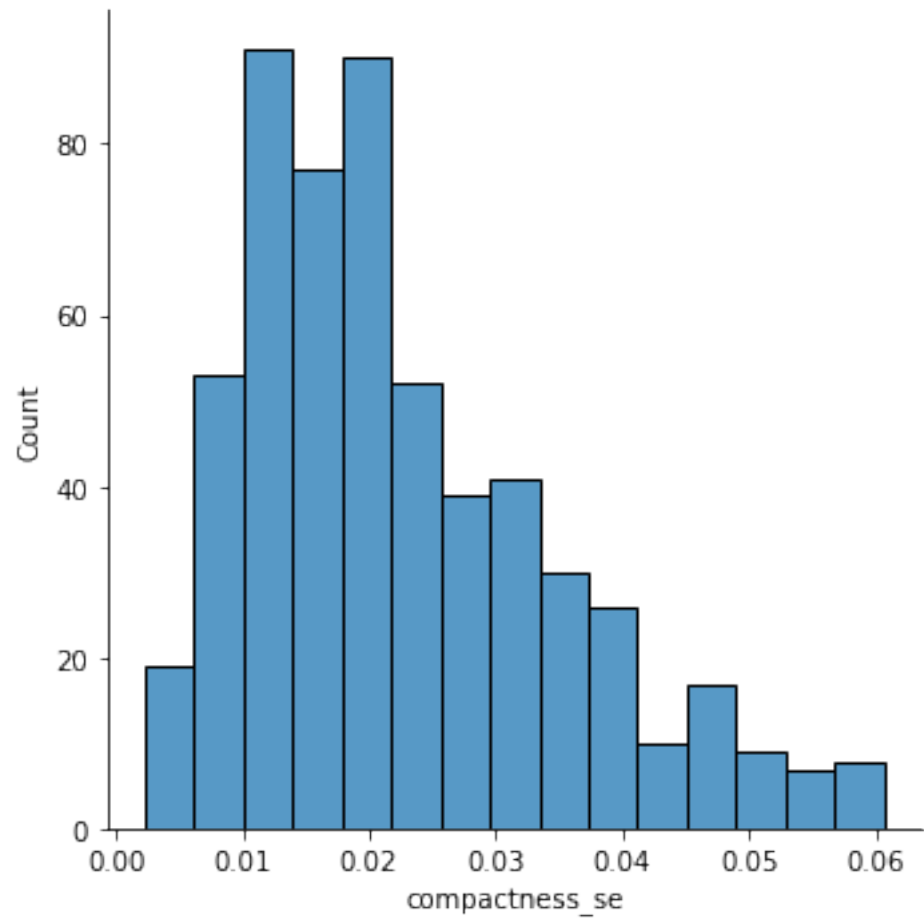


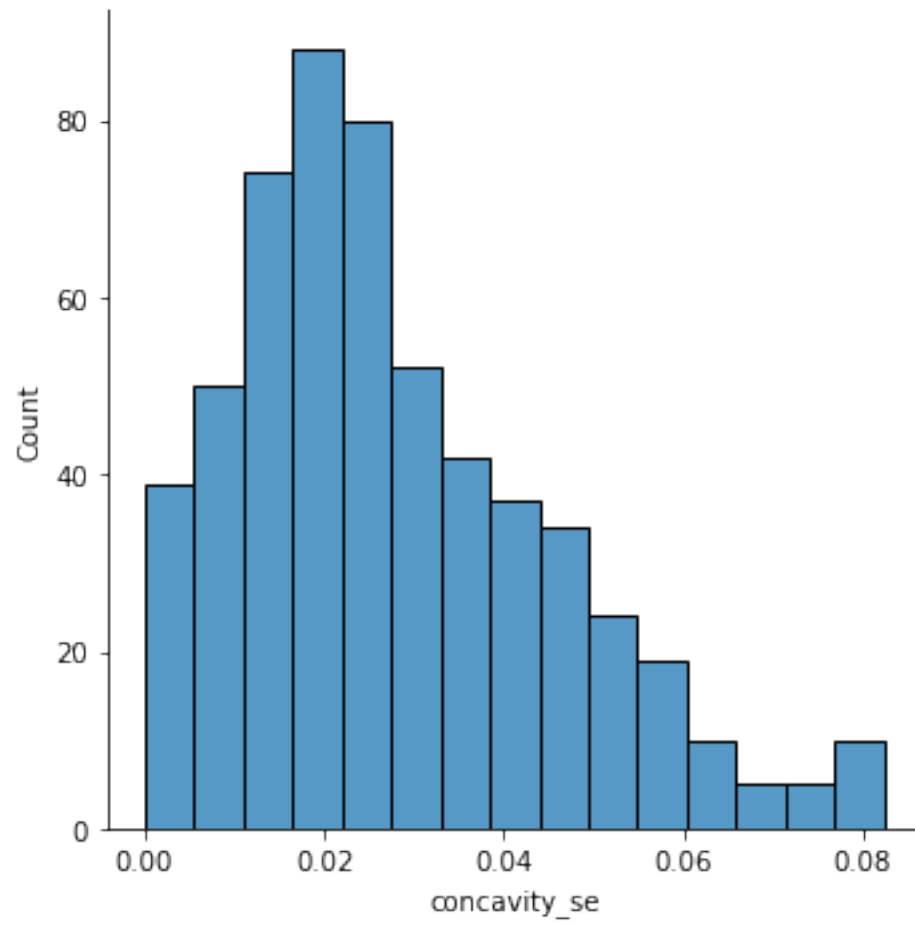


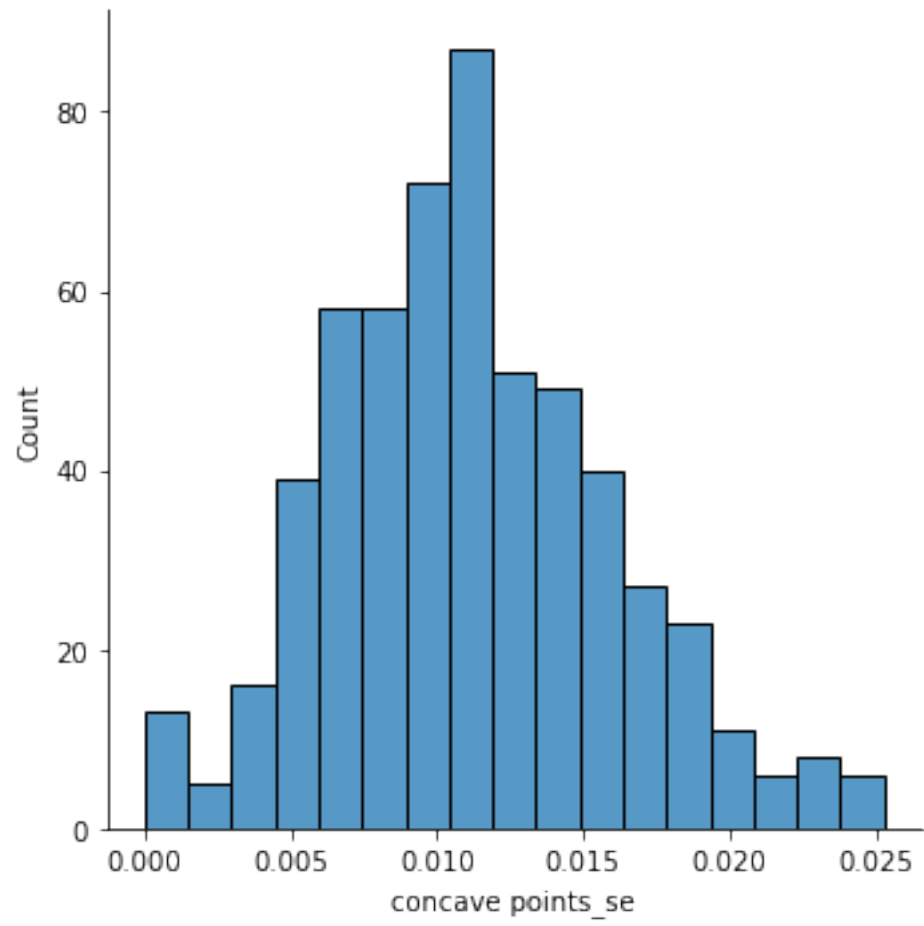


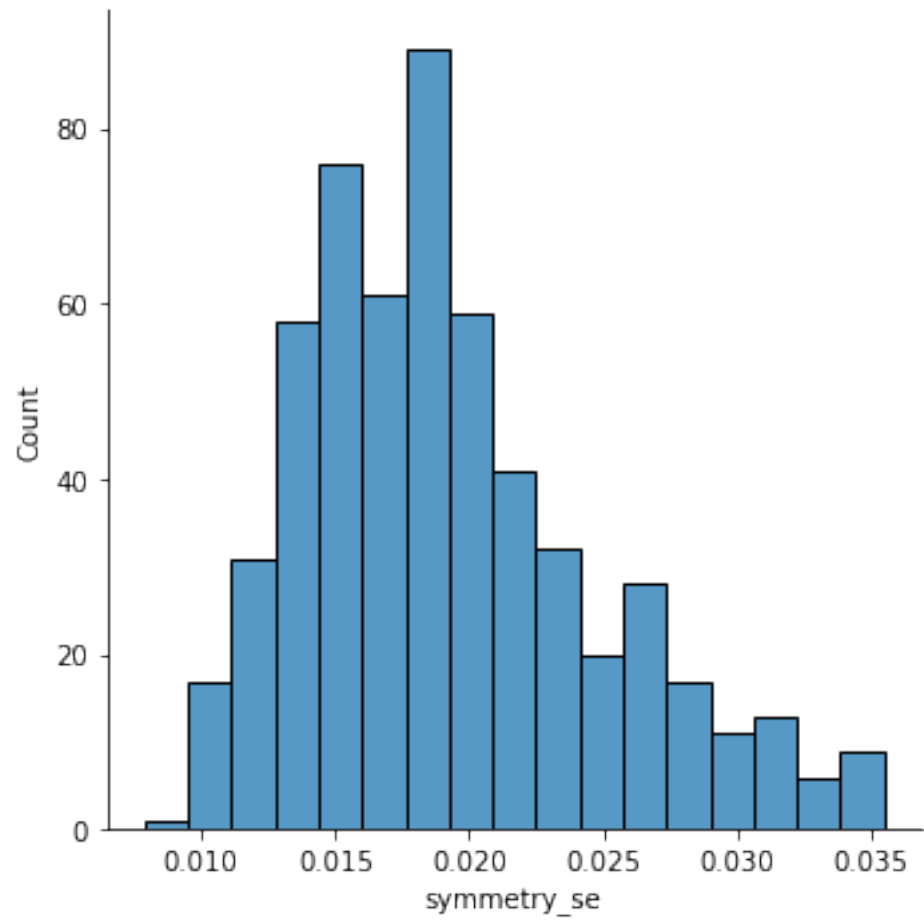


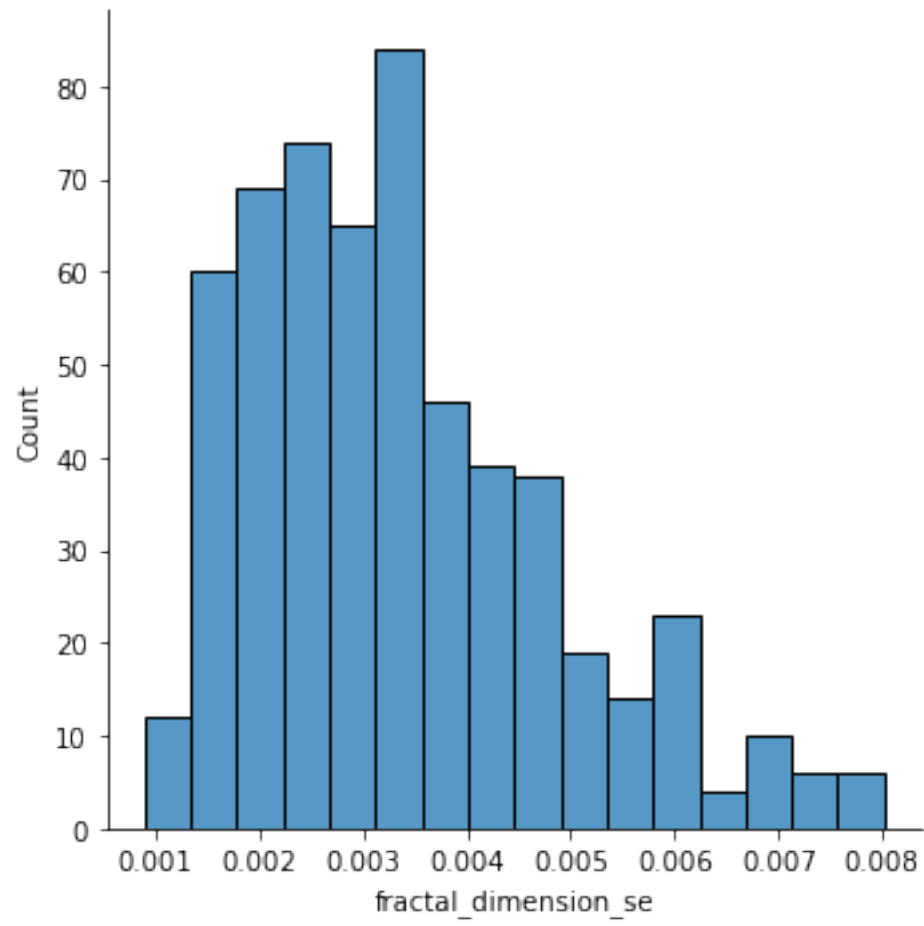


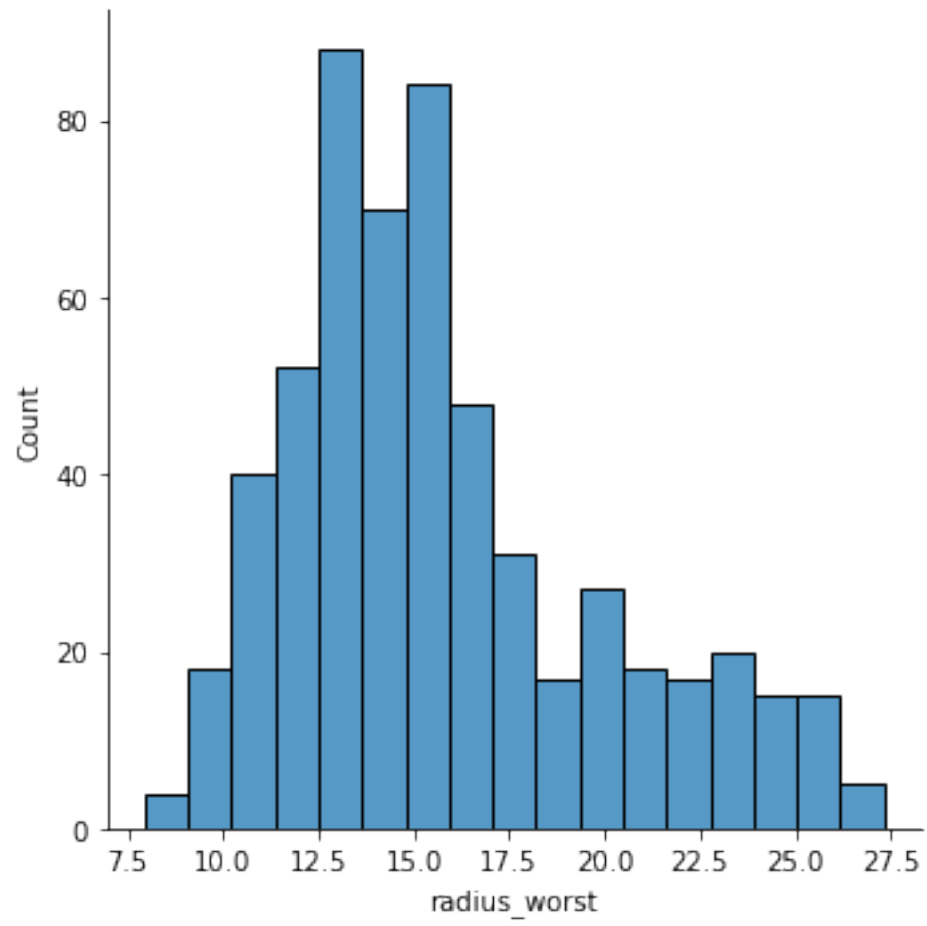


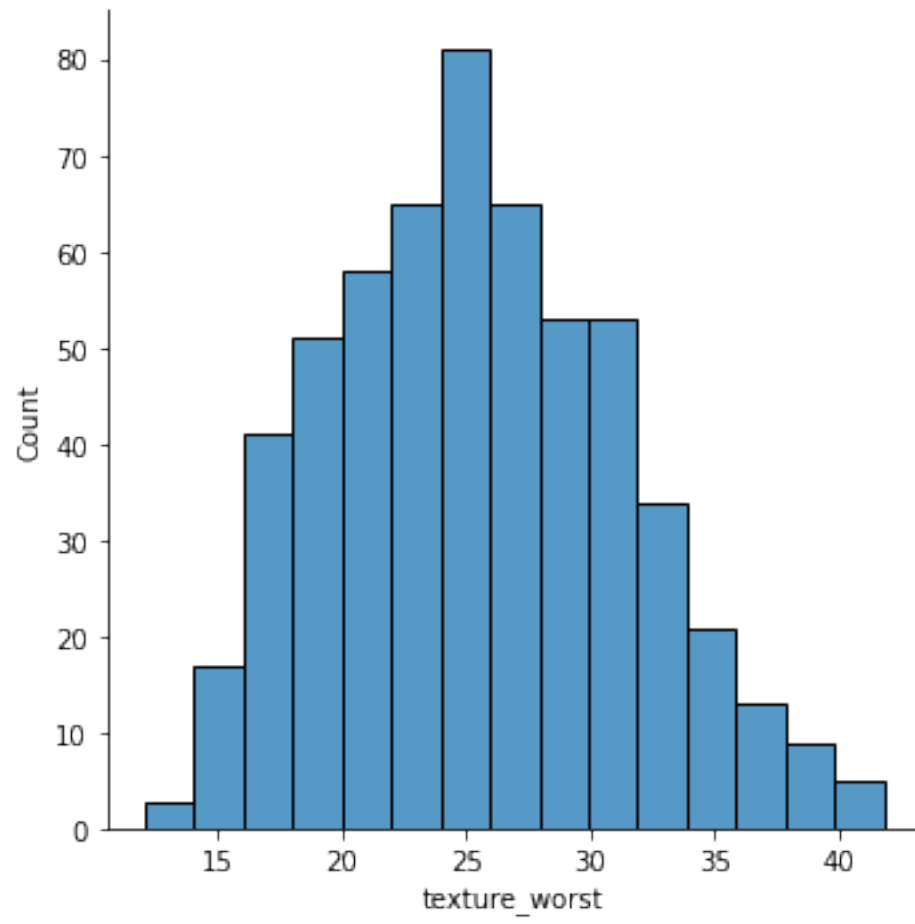


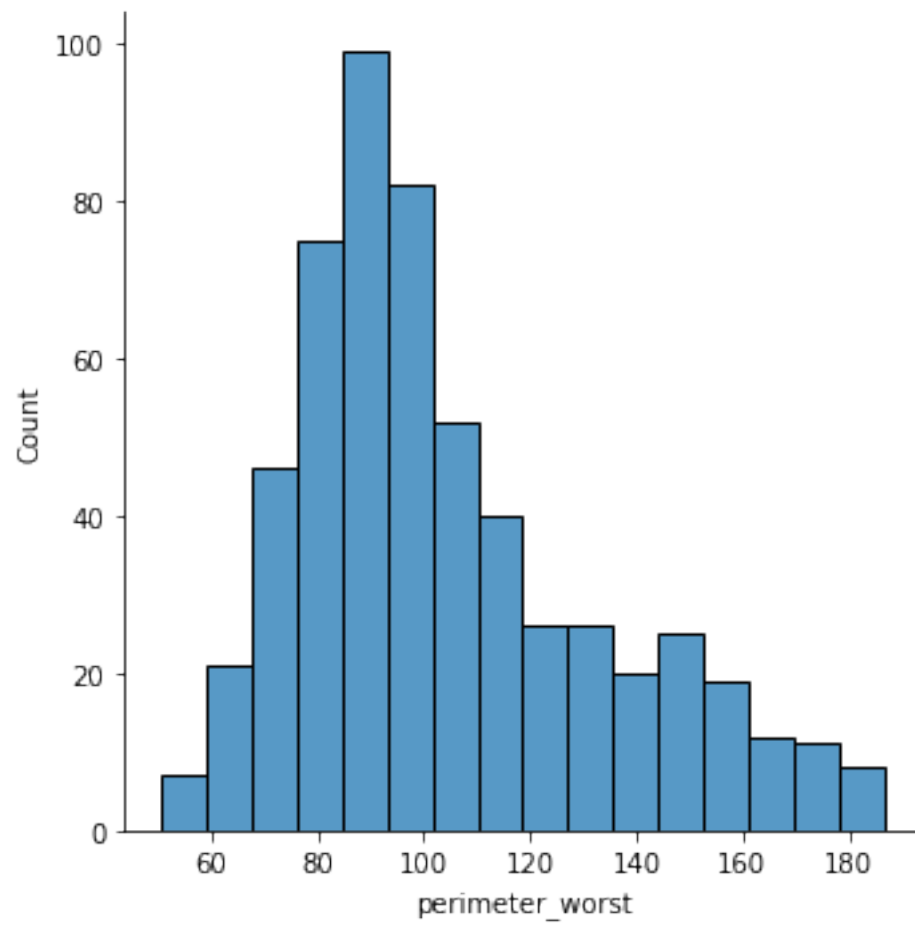


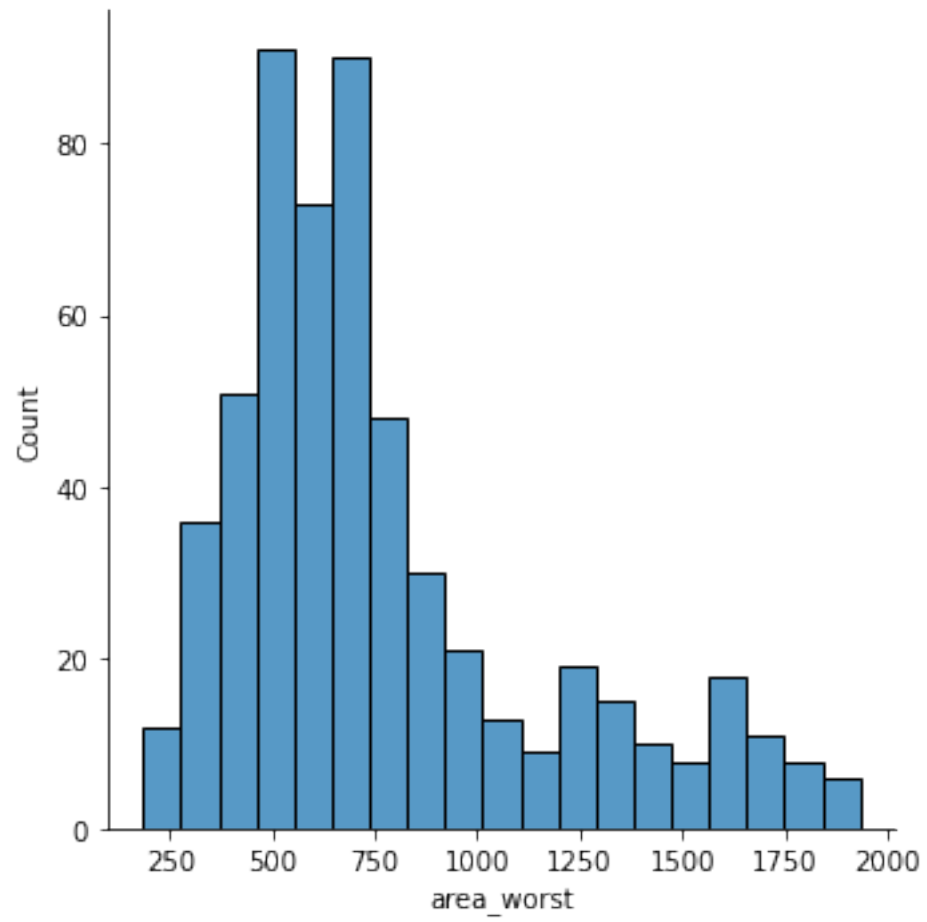


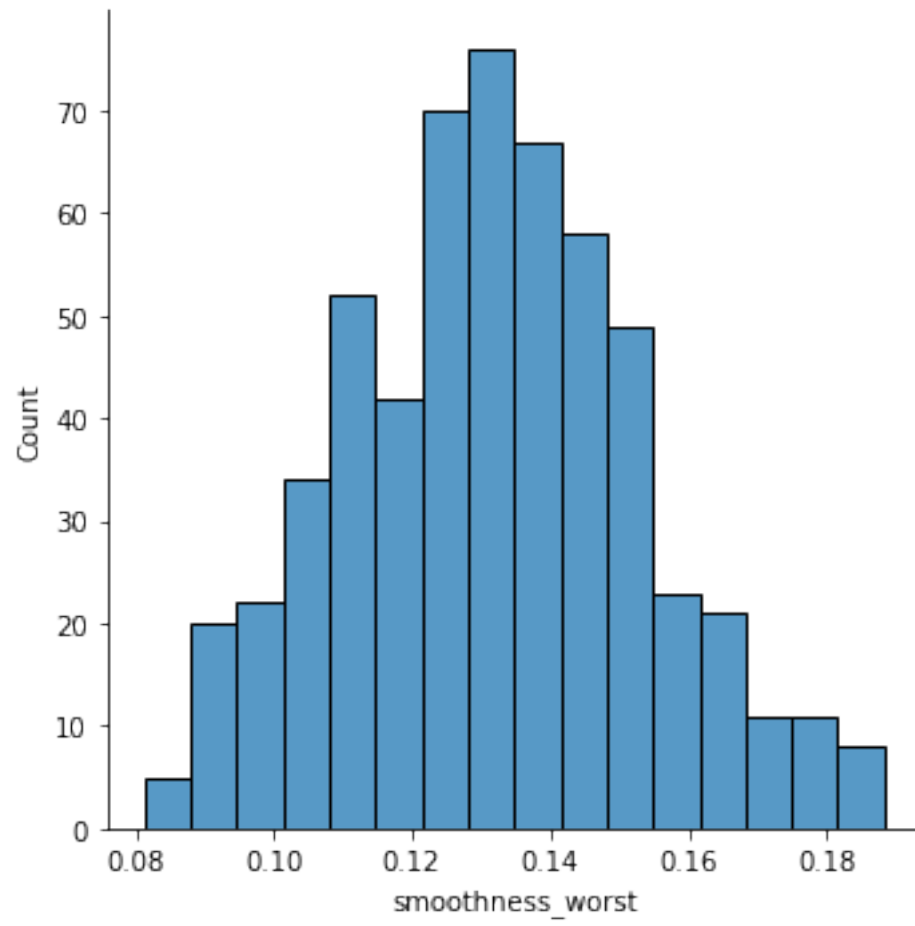


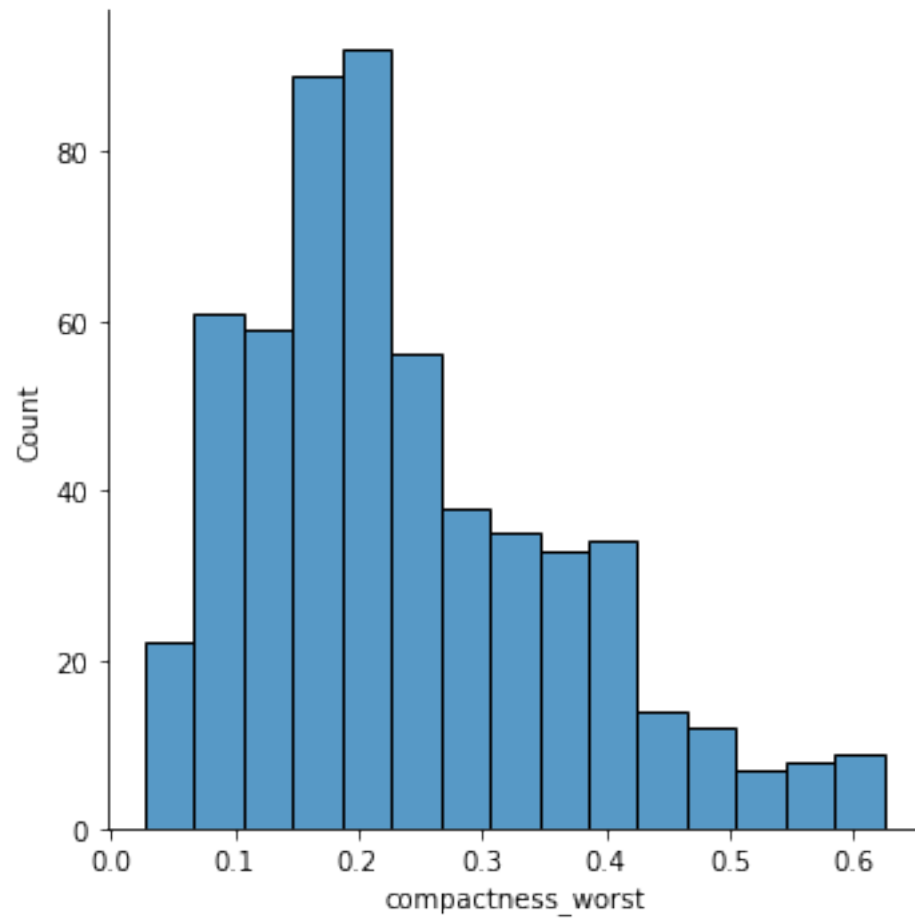


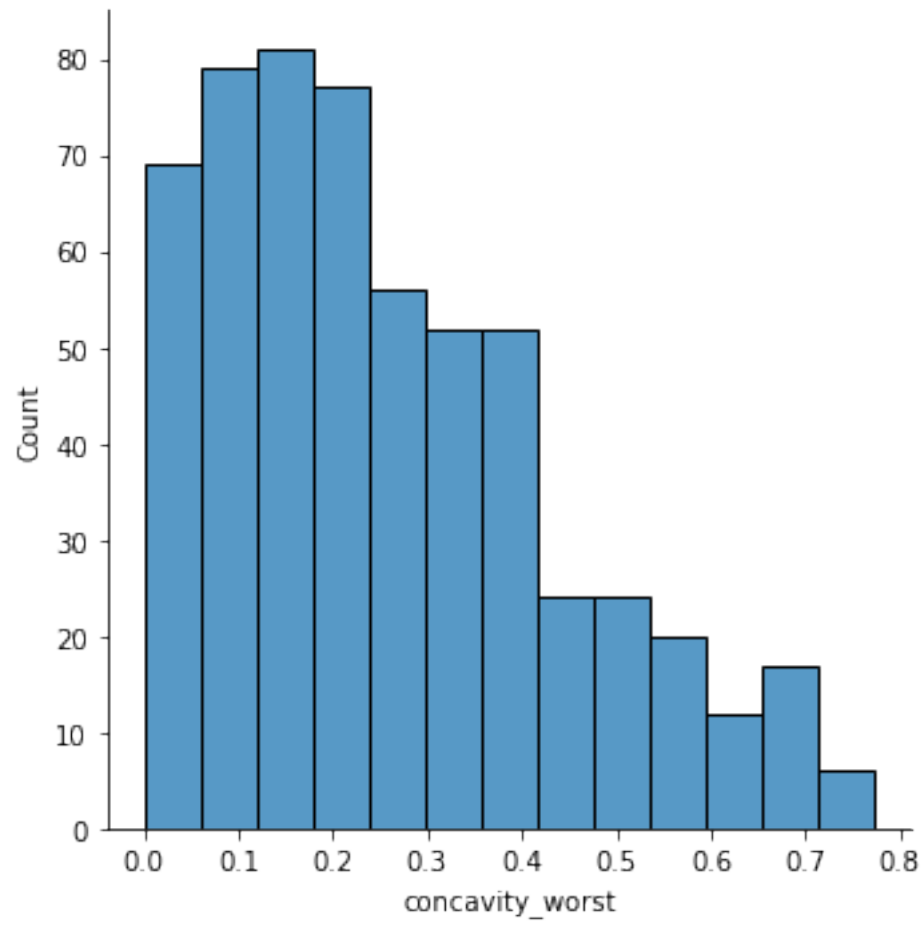


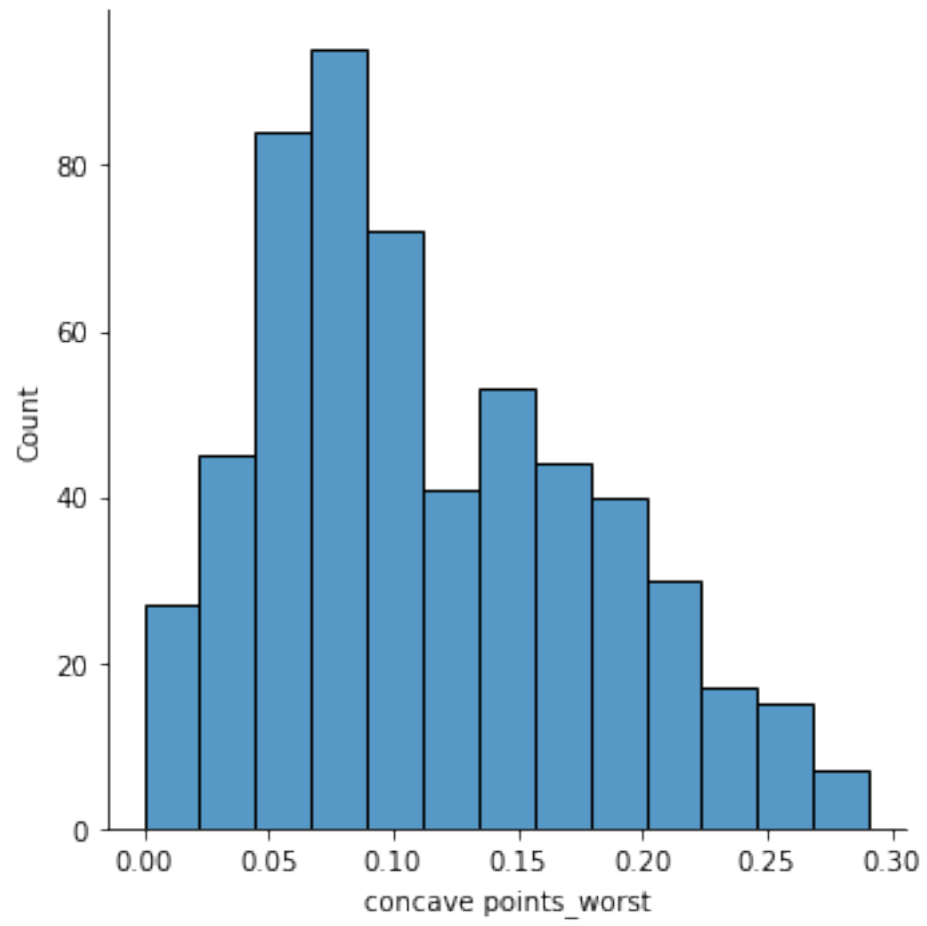


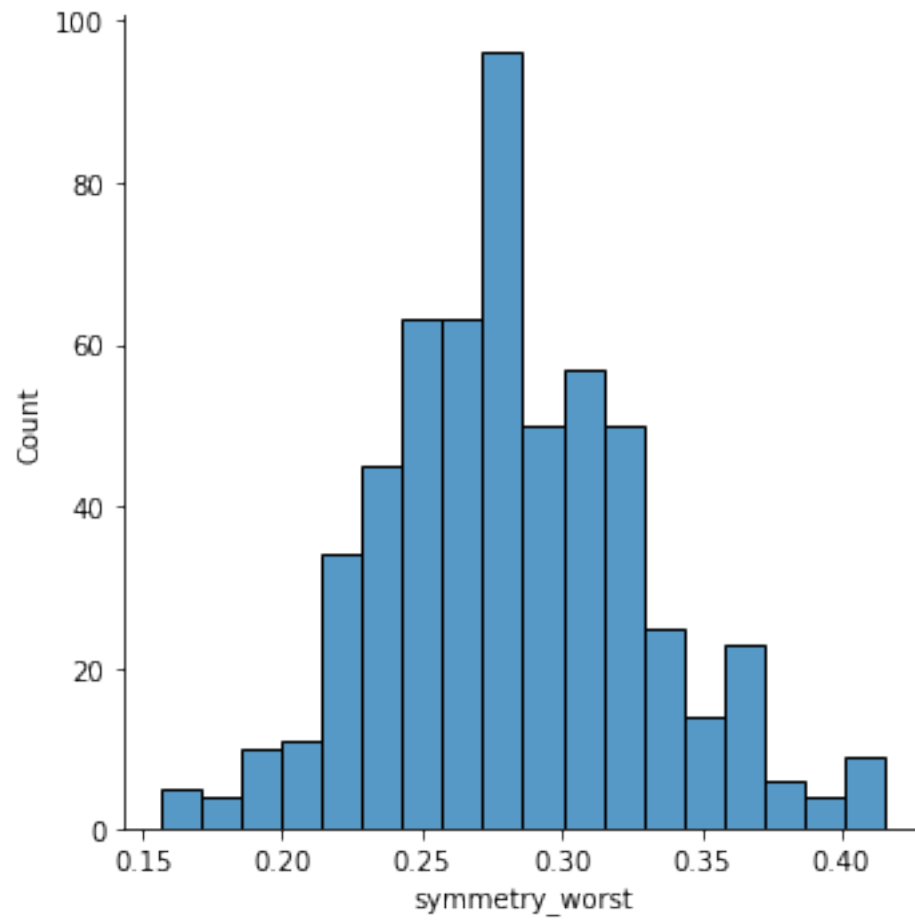


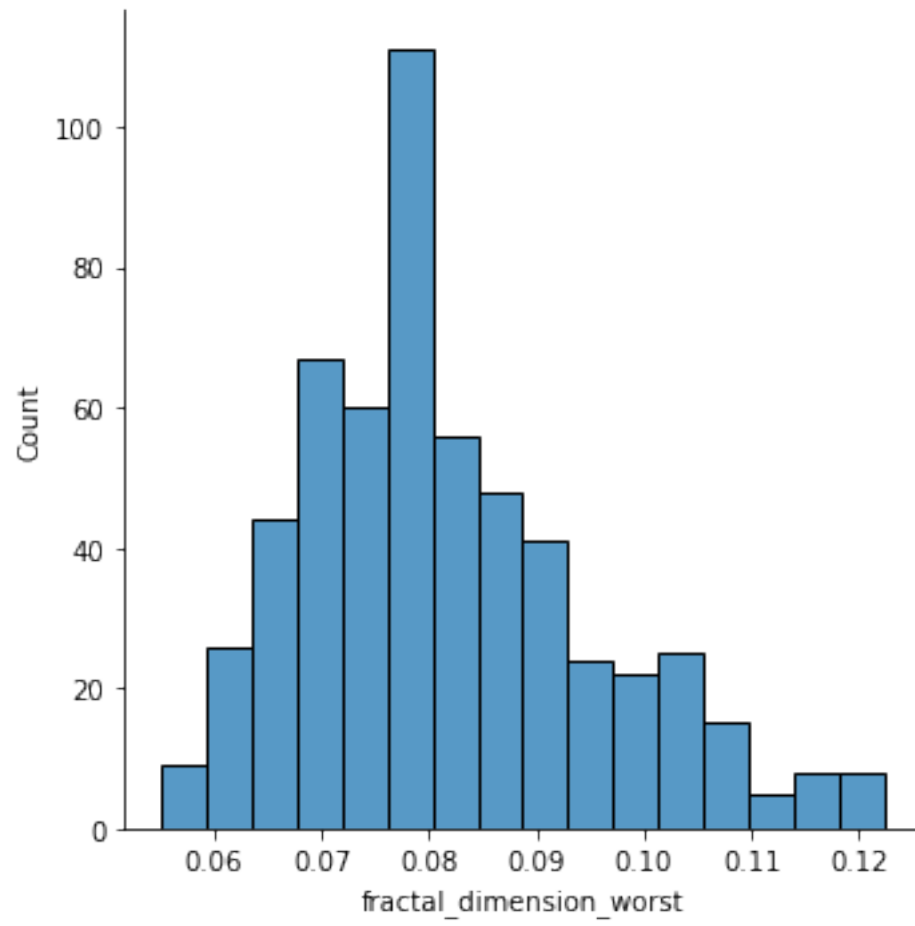


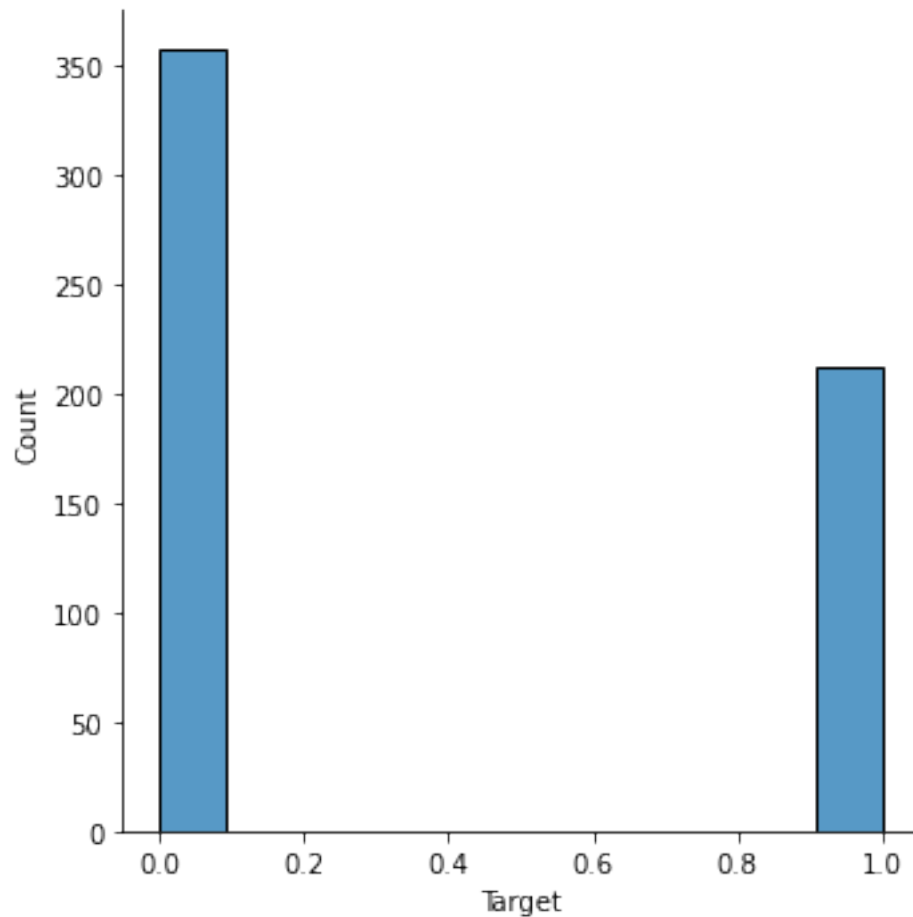












As we see data is right skewed , most of the value are left side

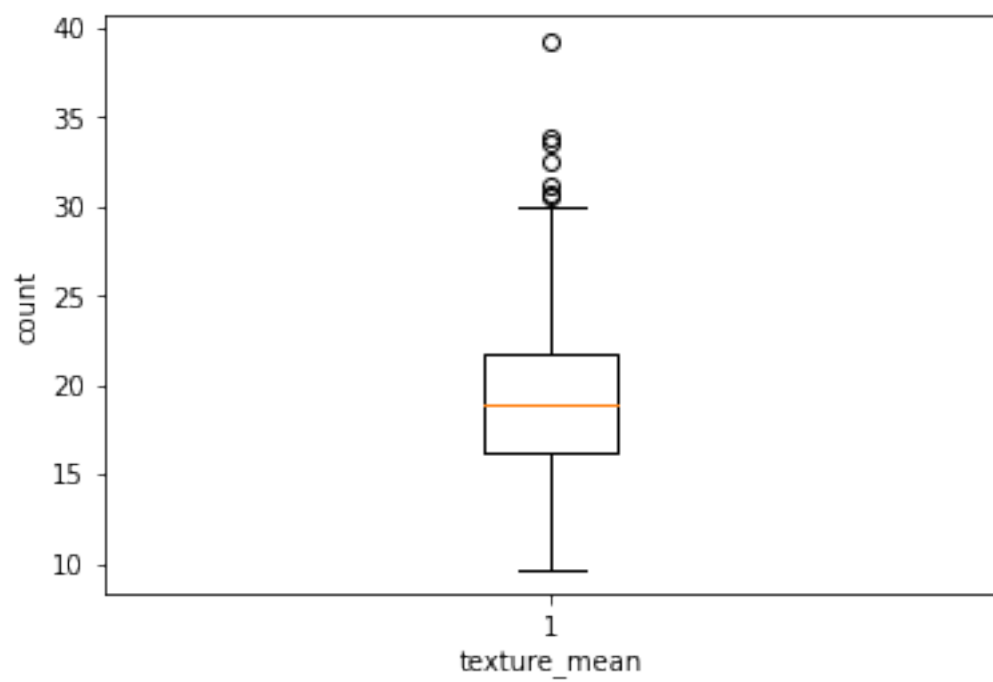
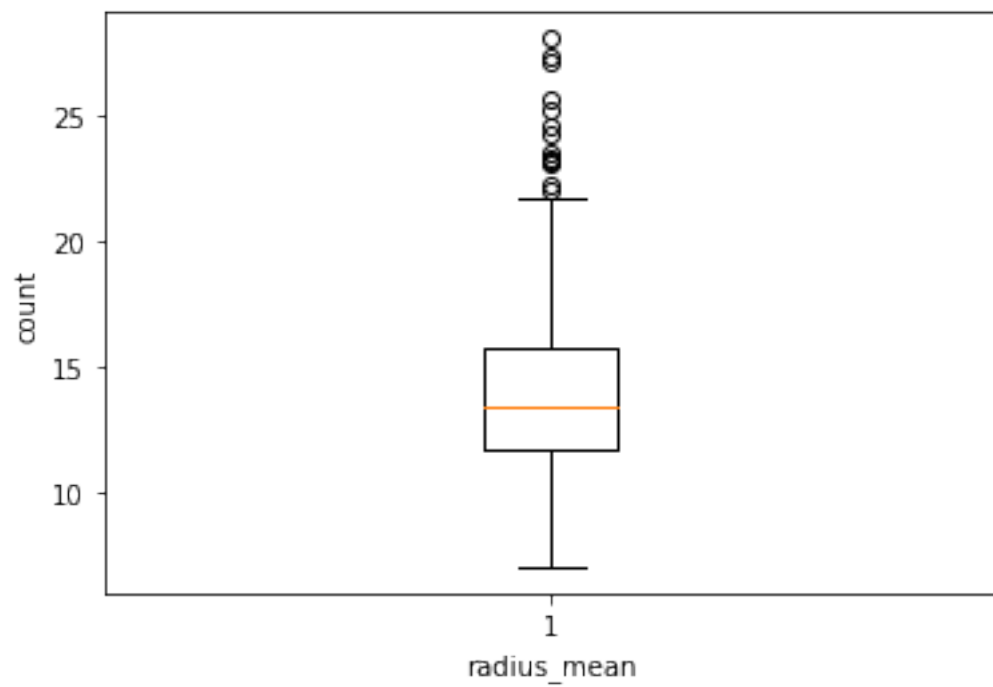
```
[21]: df.skew()
```

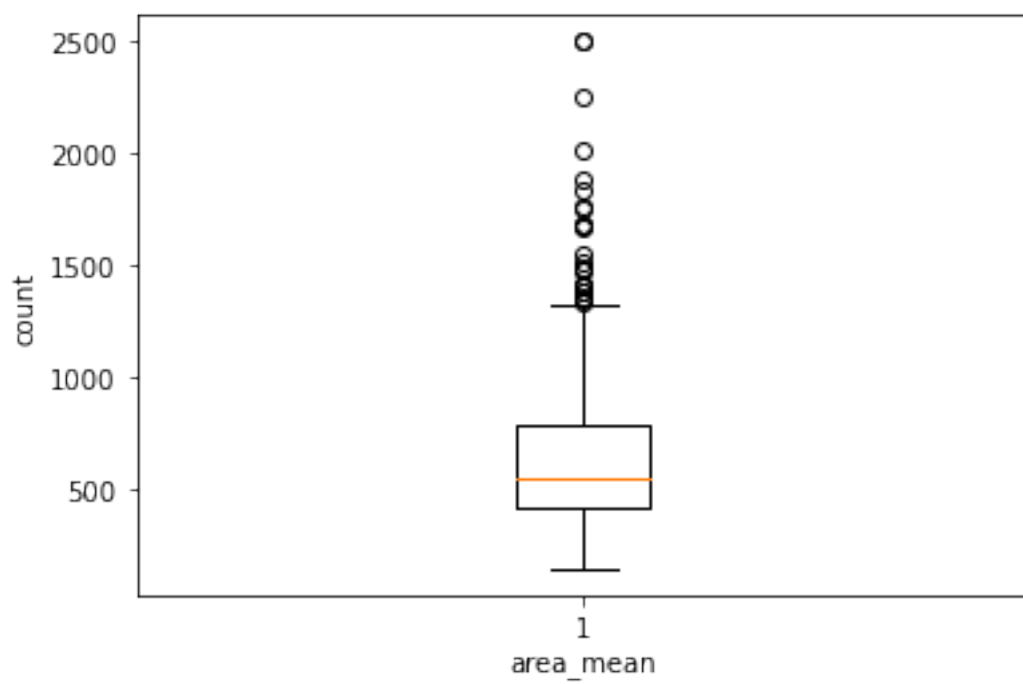
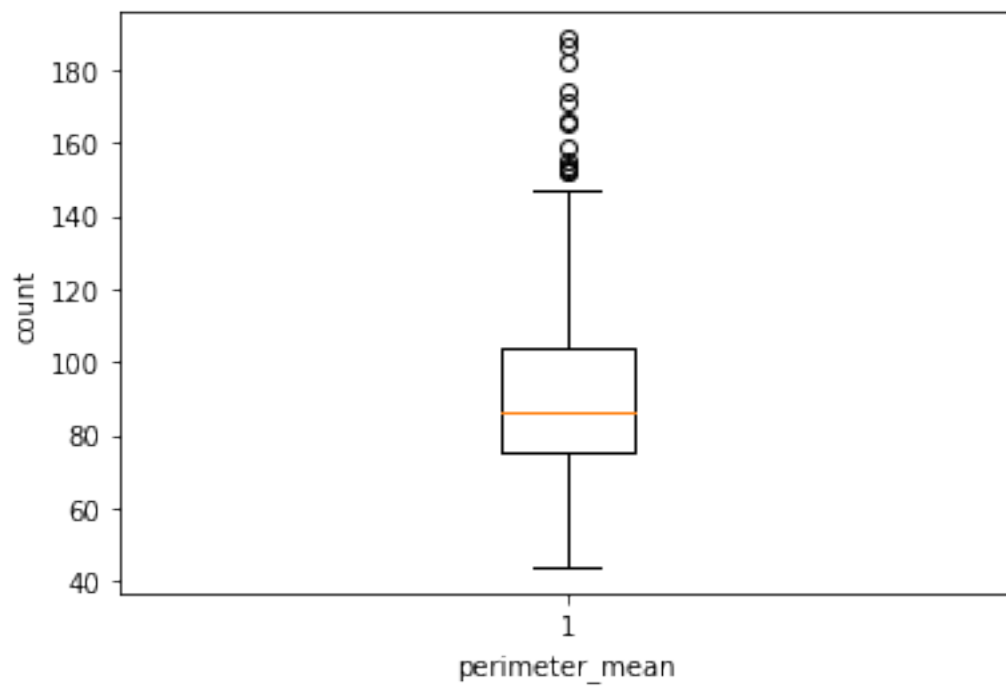
```
[21]: radius_mean      0.942380
      texture_mean    0.650450
      perimeter_mean  0.990650
      area_mean       1.645732
      smoothness_mean 0.456324
      compactness_mean 1.190123
      concavity_mean  1.401180
      concave points_mean 1.171180
      symmetry_mean   0.725609
      fractal_dimension_mean 1.304489
      radius_se       3.088612
      texture_se      1.646444
      perimeter_se    3.443615
      area_se         5.447186
```

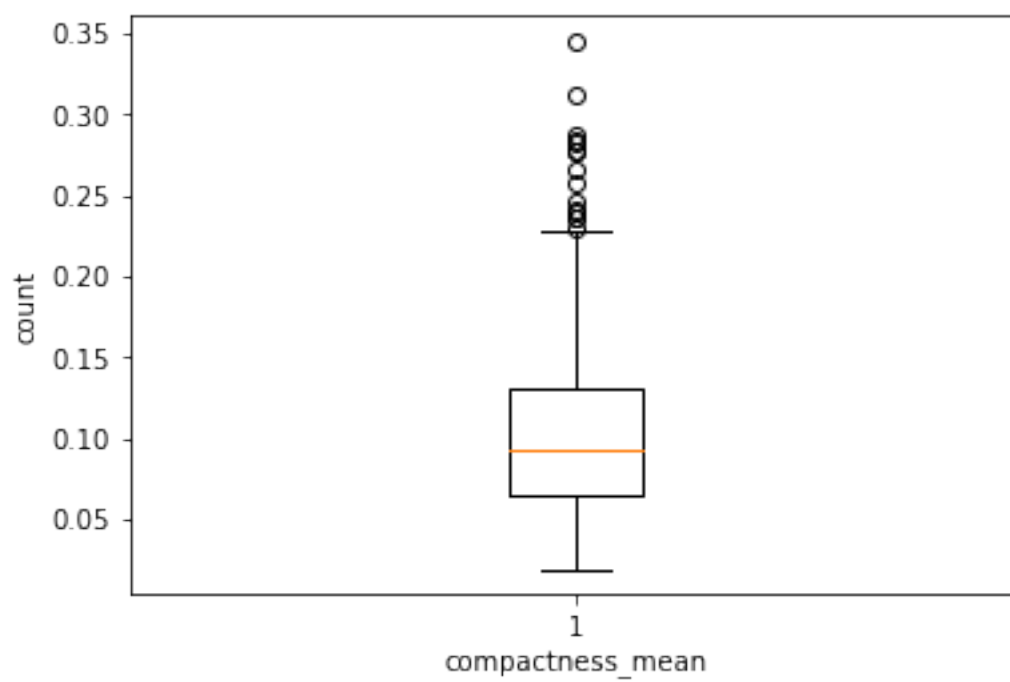
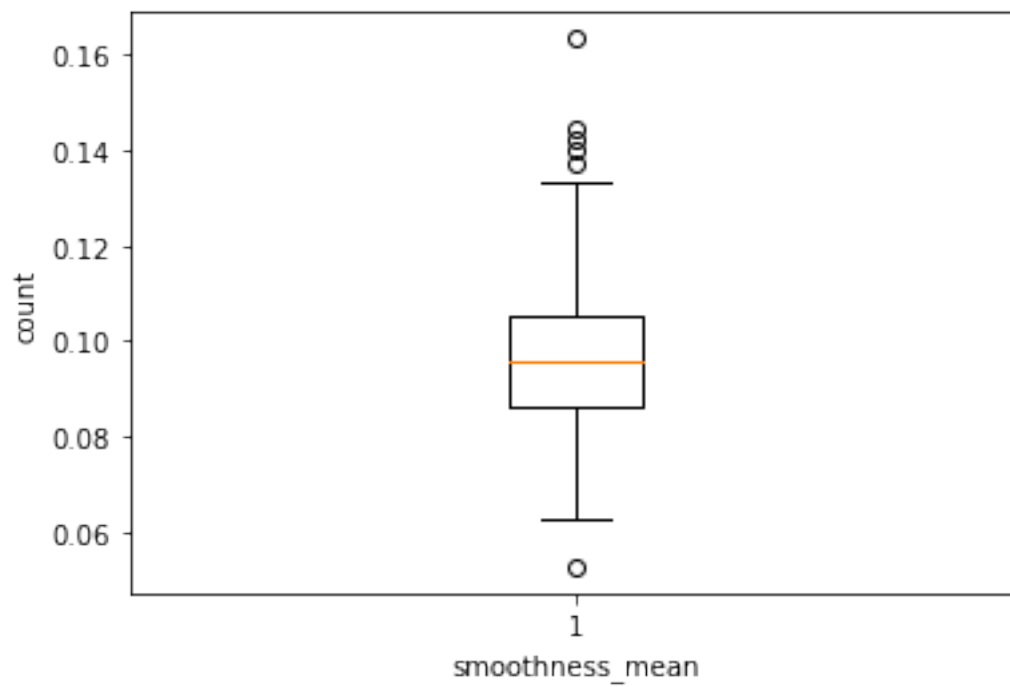
smoothness_se	2.314450
compactness_se	1.902221
concavity_se	5.110463
concave points_se	1.444678
symmetry_se	2.195133
fractal_dimension_se	3.923969
radius_worst	1.103115
texture_worst	0.498321
perimeter_worst	1.128164
area_worst	1.859373
smoothness_worst	0.415426
compactness_worst	1.473555
concavity_worst	1.150237
concave points_worst	0.492616
symmetry_worst	1.433928
fractal_dimension_worst	1.662579
Target	0.528461
dtype:	float64

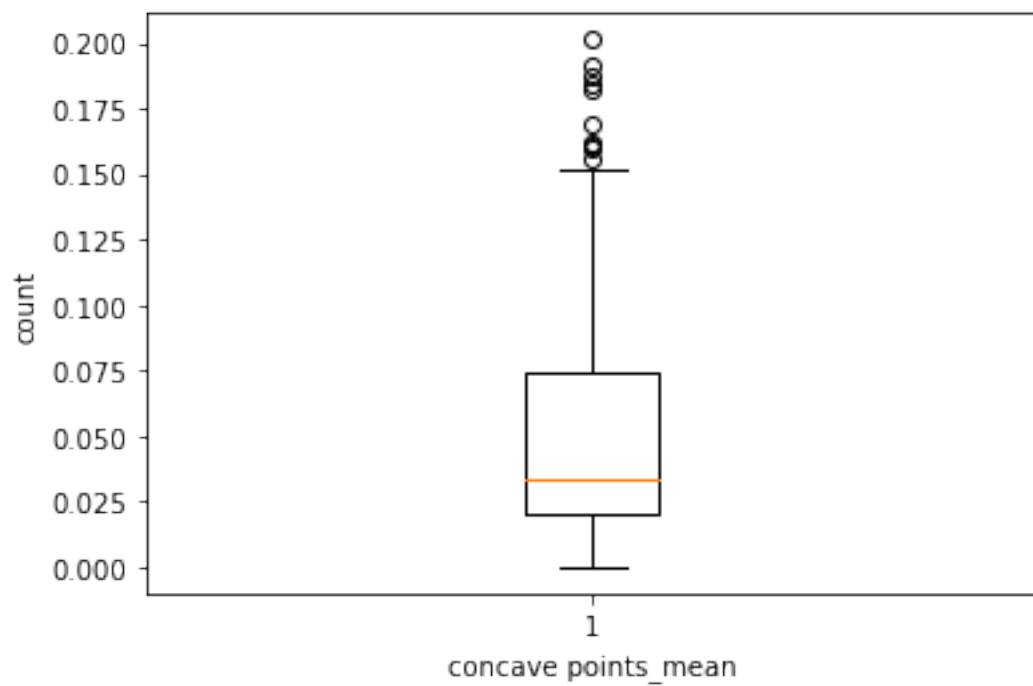
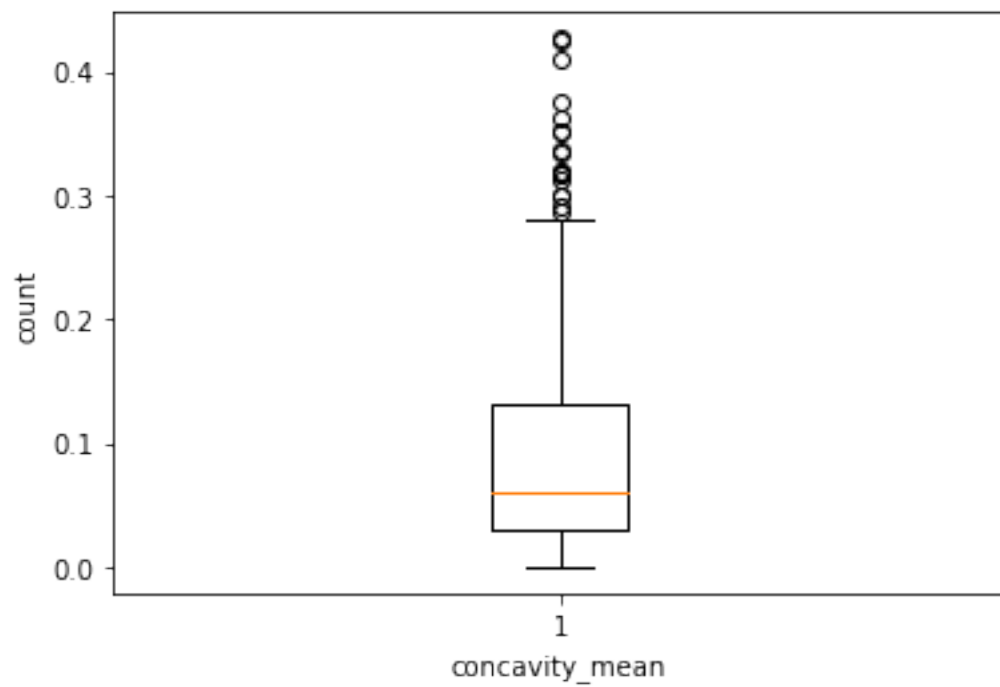
If the skewness is between -0.5 and 0.5, the data is approximately symmetric. If the skewness is between -1 and -0.5 or between 0.5 and 1, the data is moderately skewed. If the skewness is less than -1 or greater than 1, the data is highly skewed.

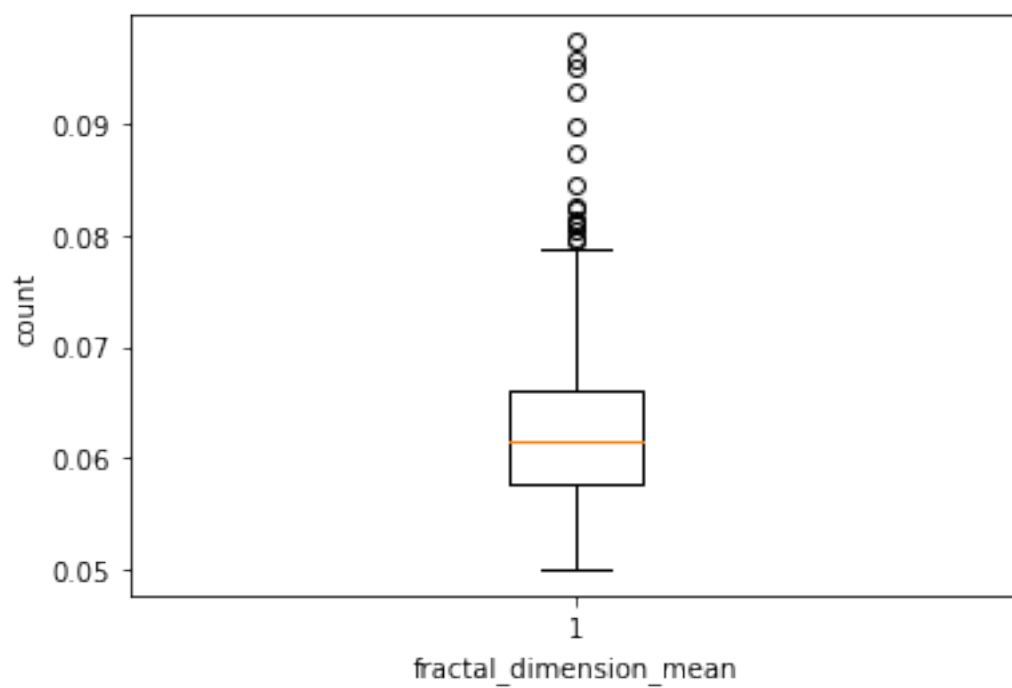
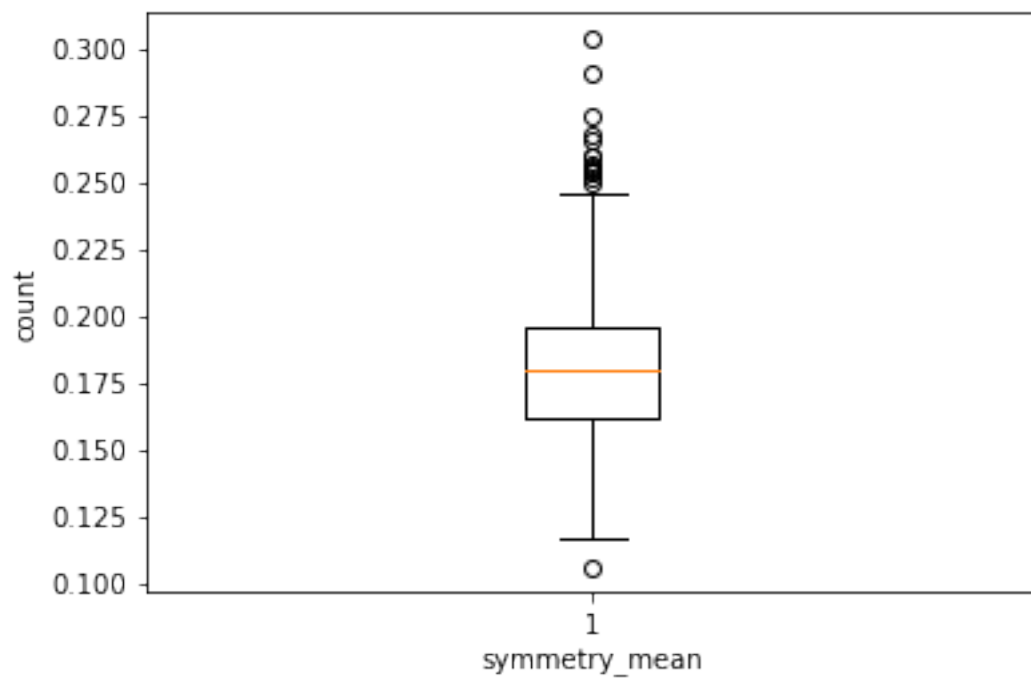
```
[22]: for col_name in df.columns:
        if(df[col_name].dtypes=='int64' or df[col_name].dtypes=='float64'):
            plt.boxplot(df[col_name])
            plt.xlabel(col_name)
            plt.ylabel('count')
            plt.show()
```

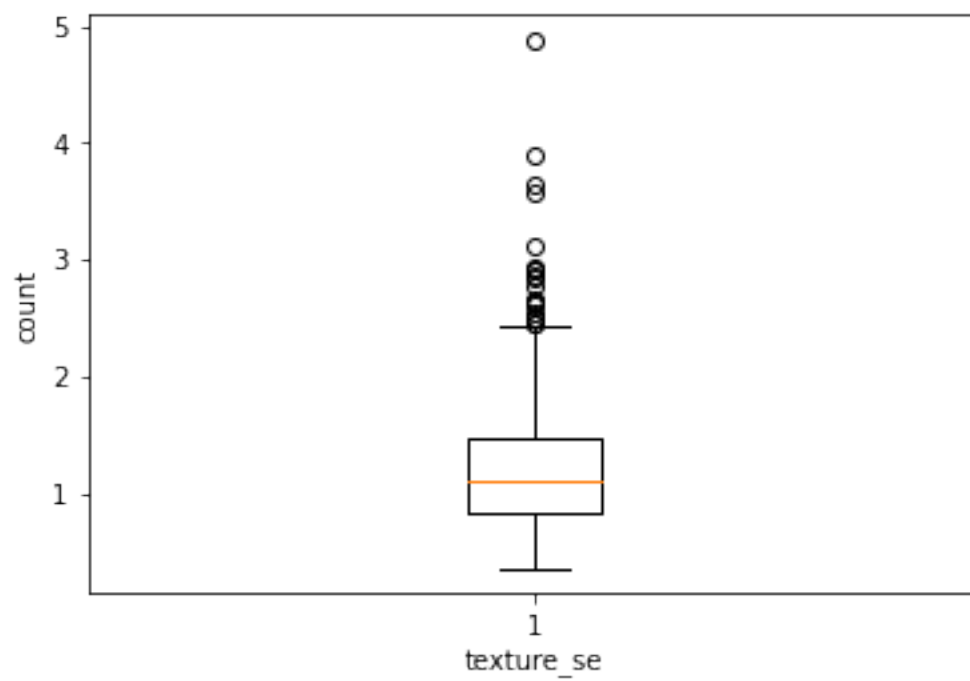
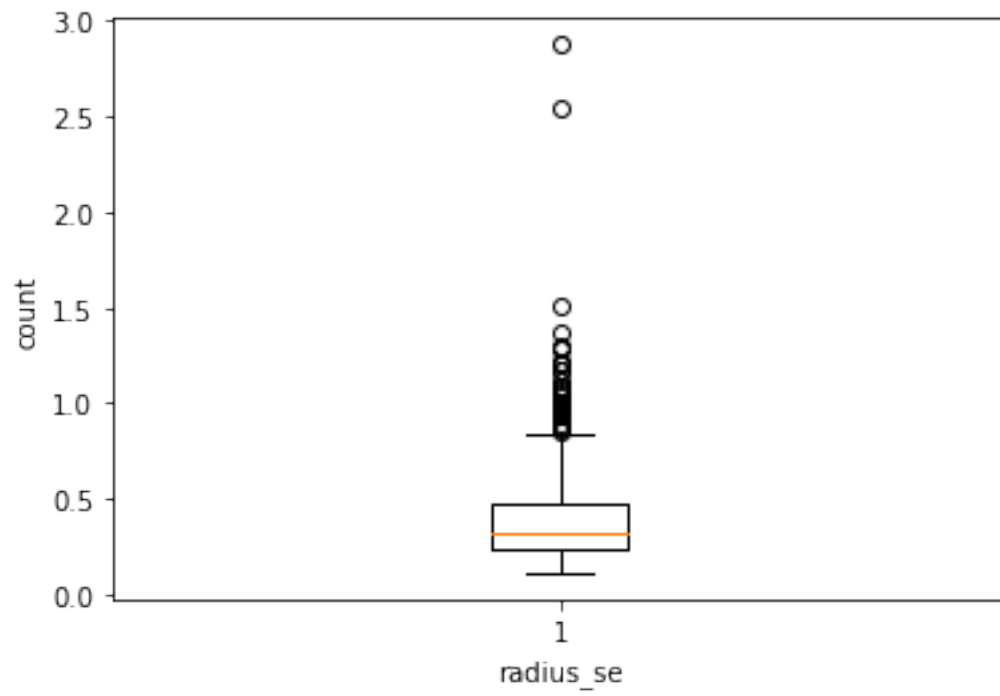


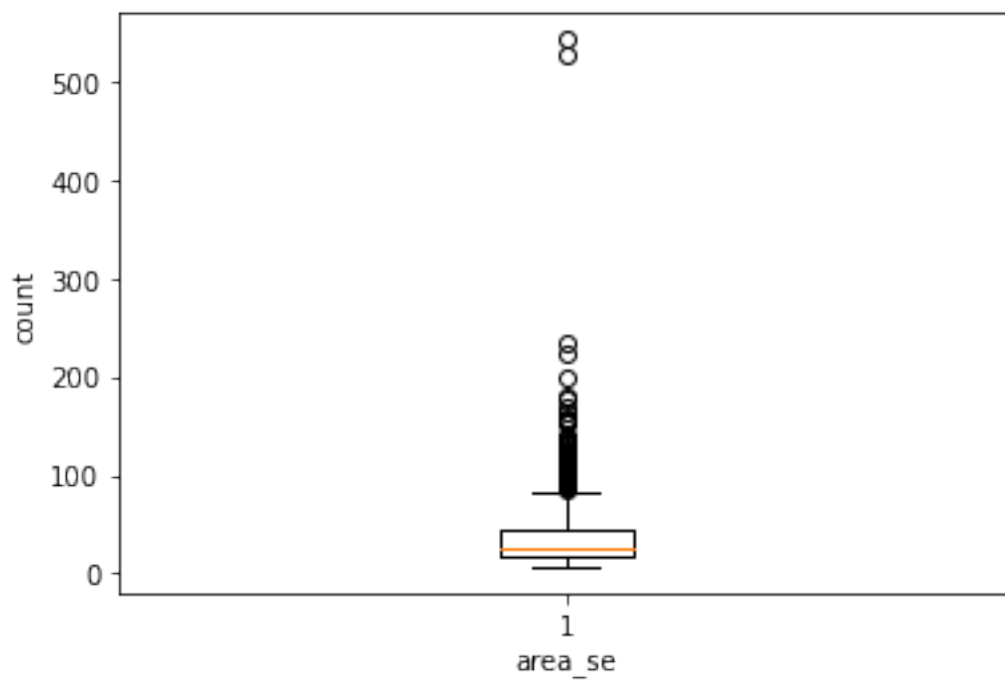
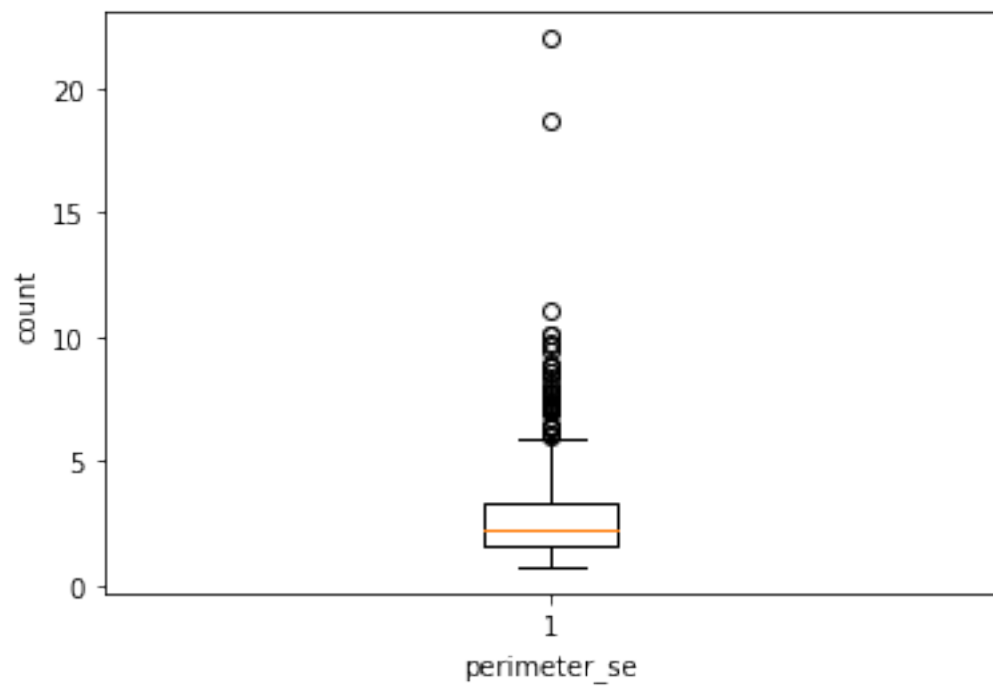


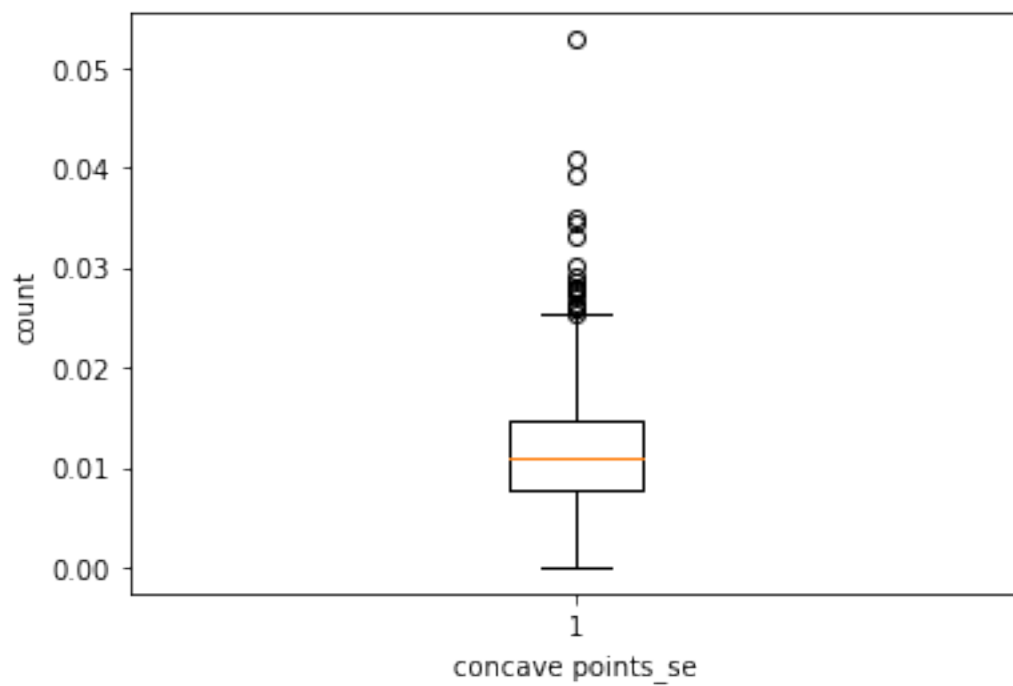
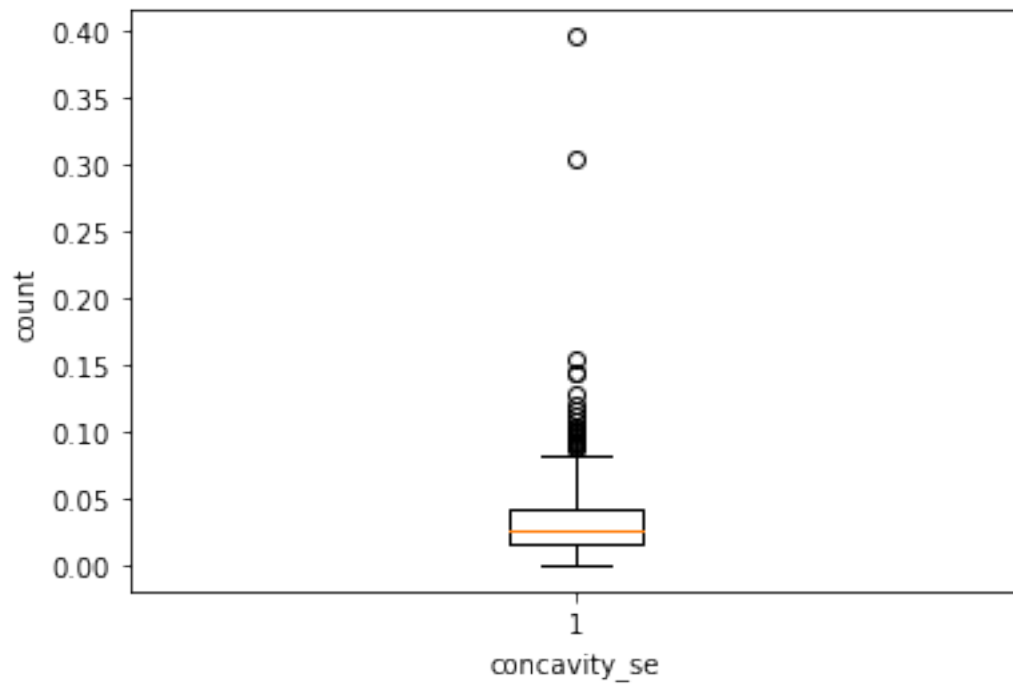


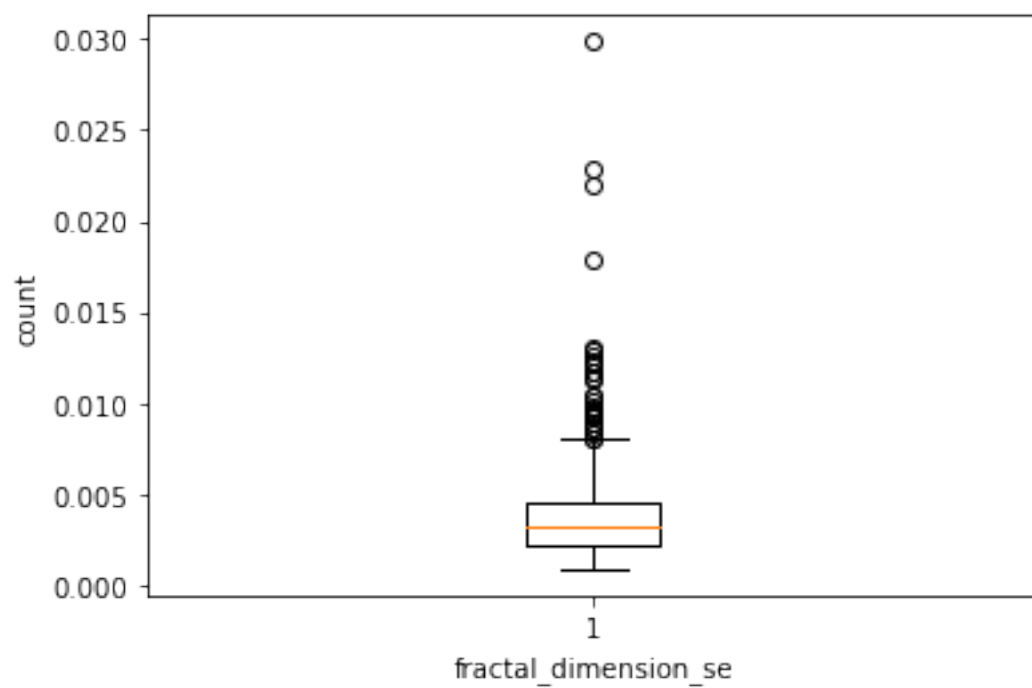
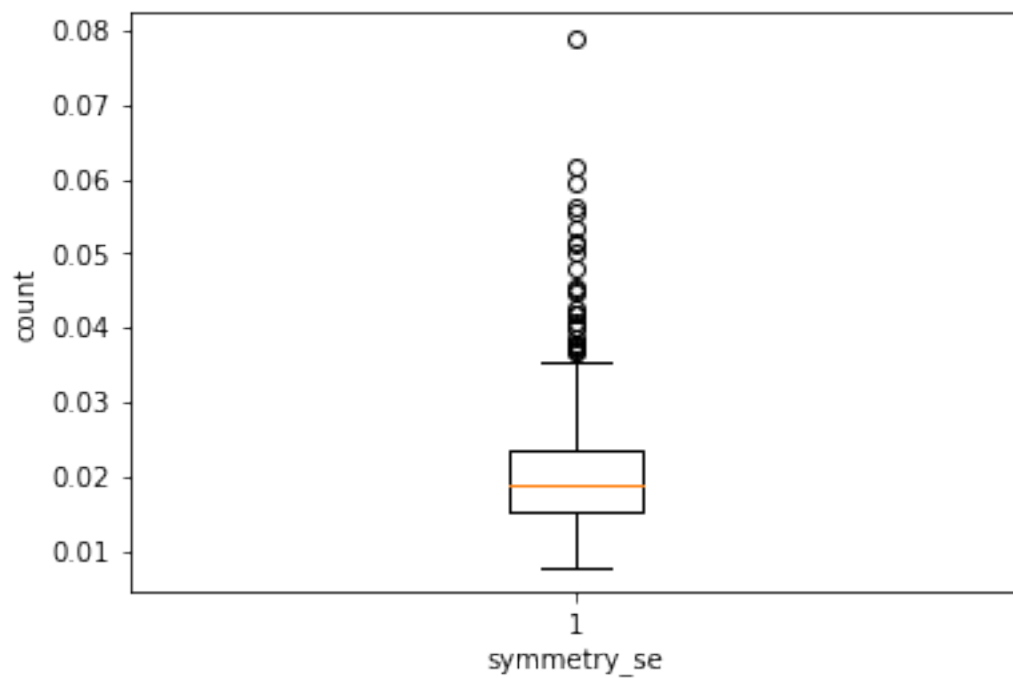


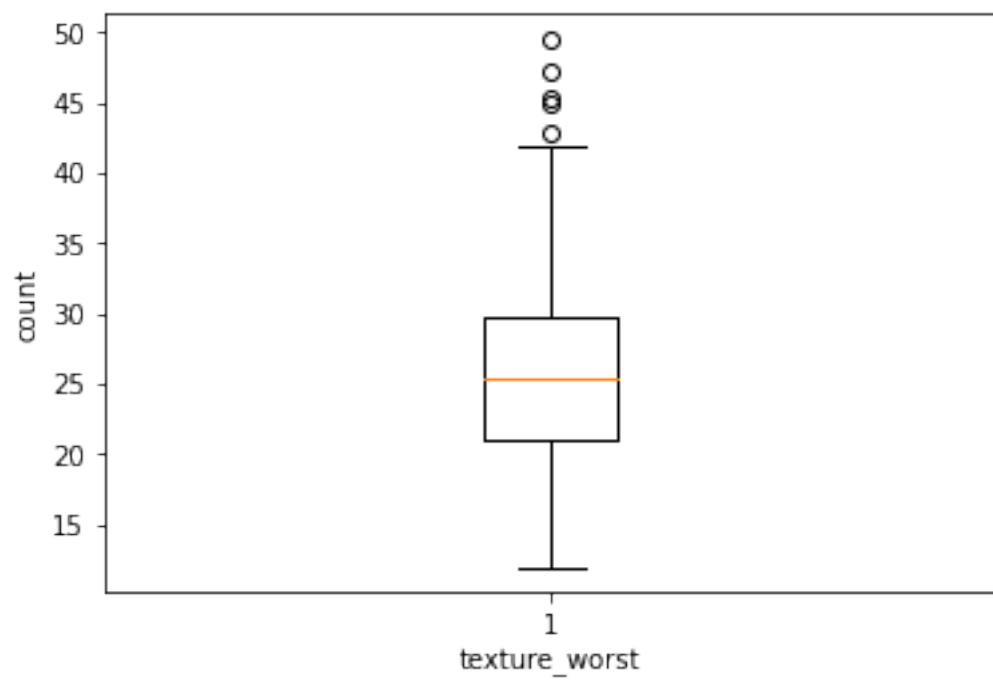
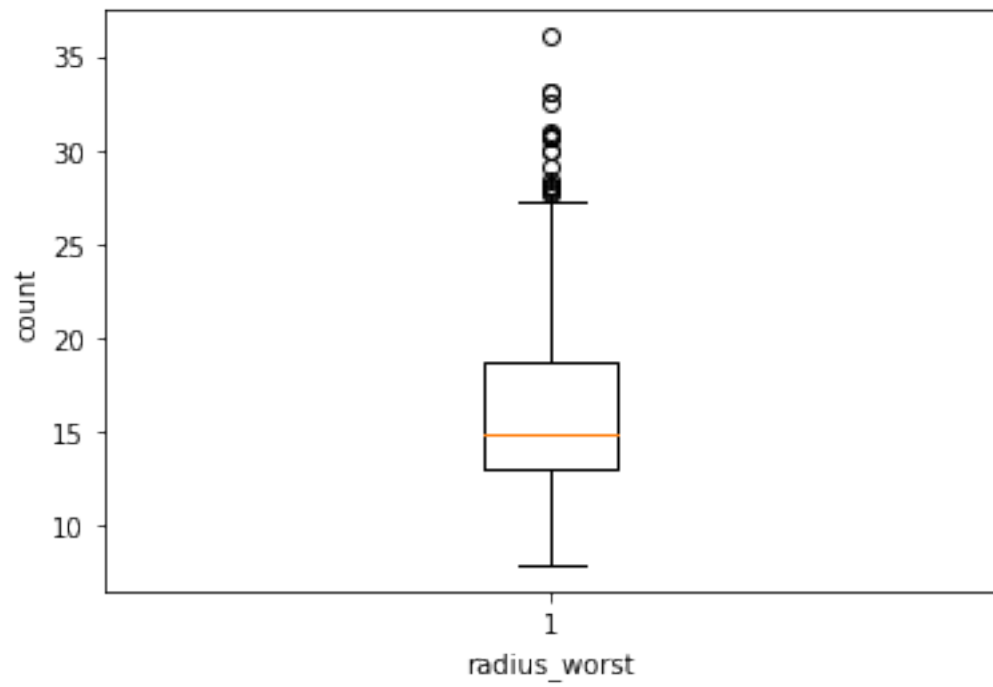


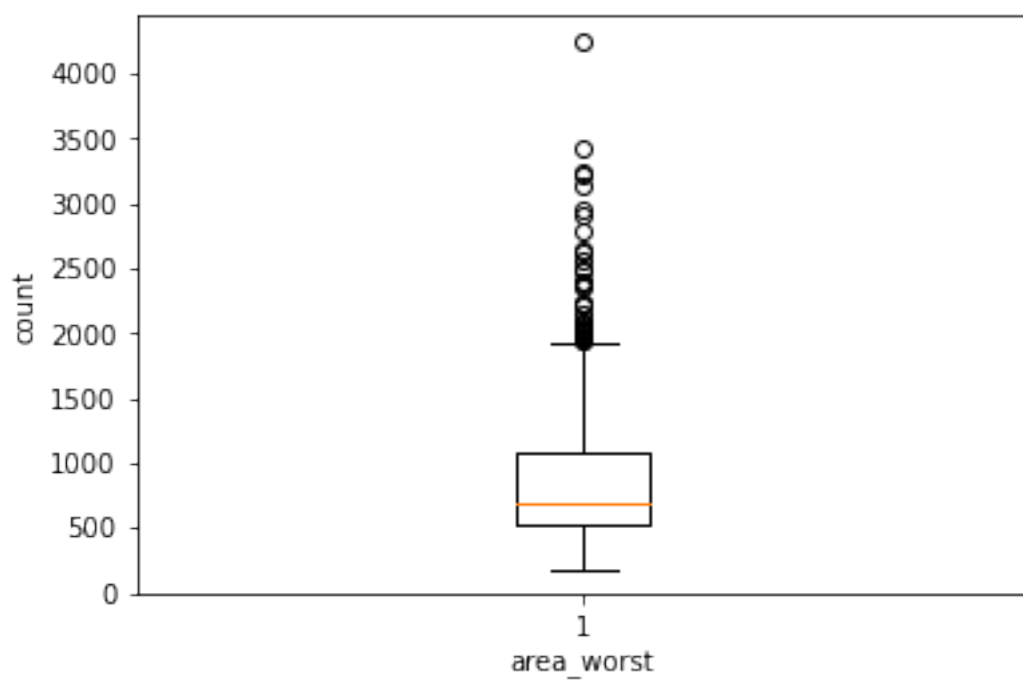
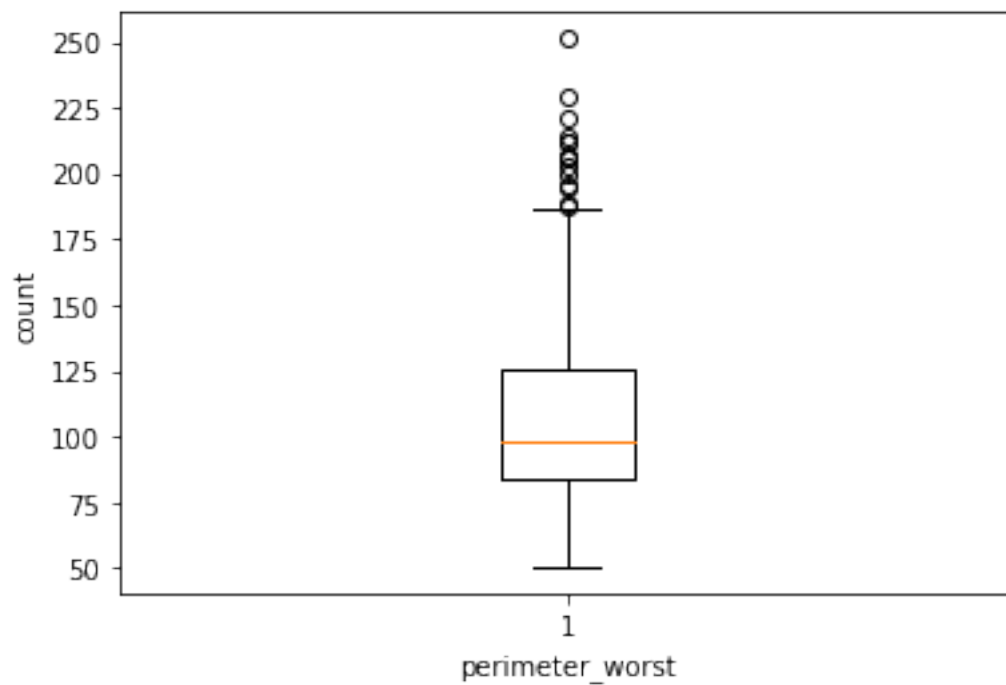


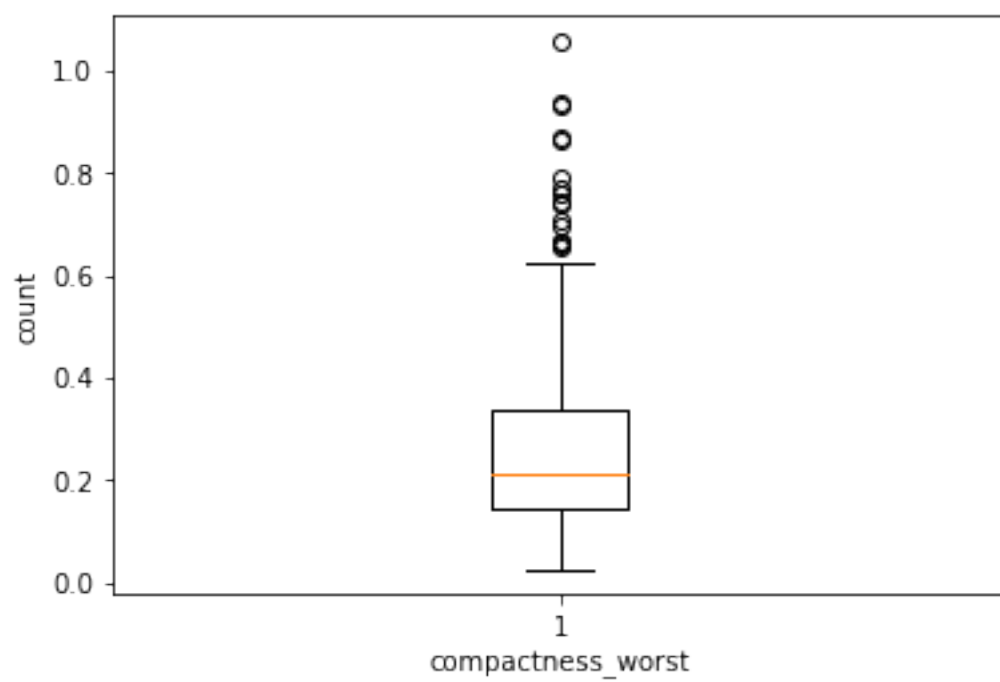


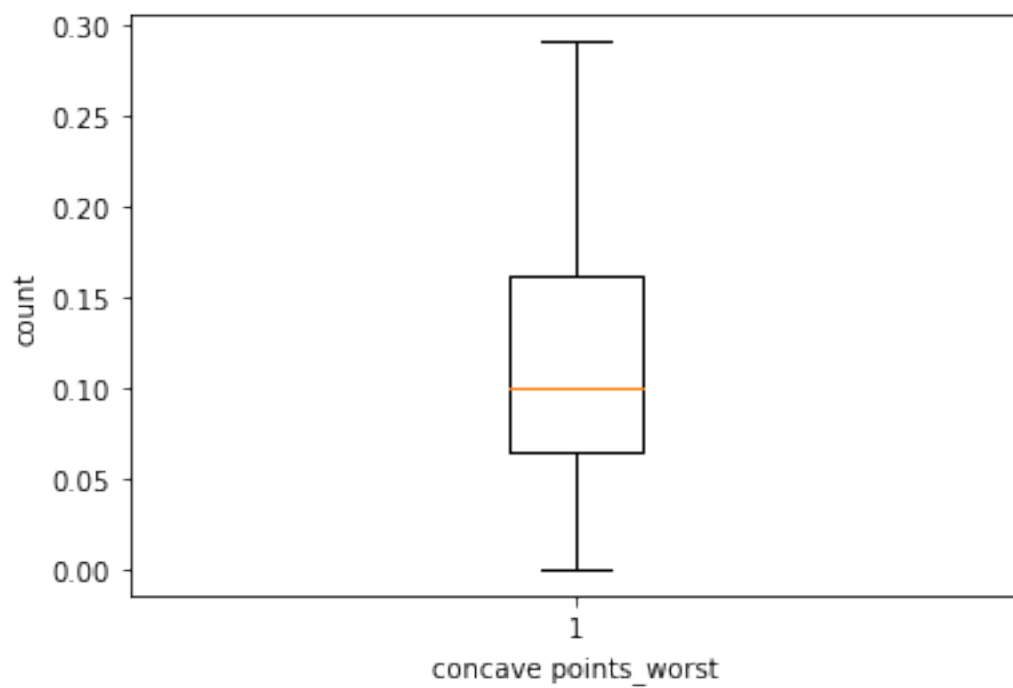
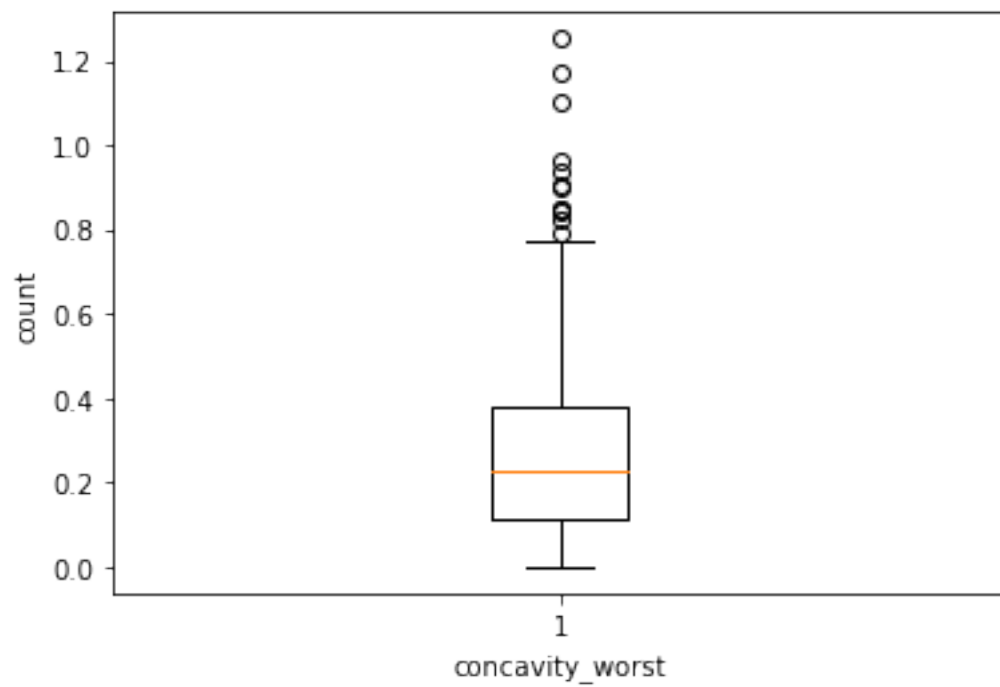


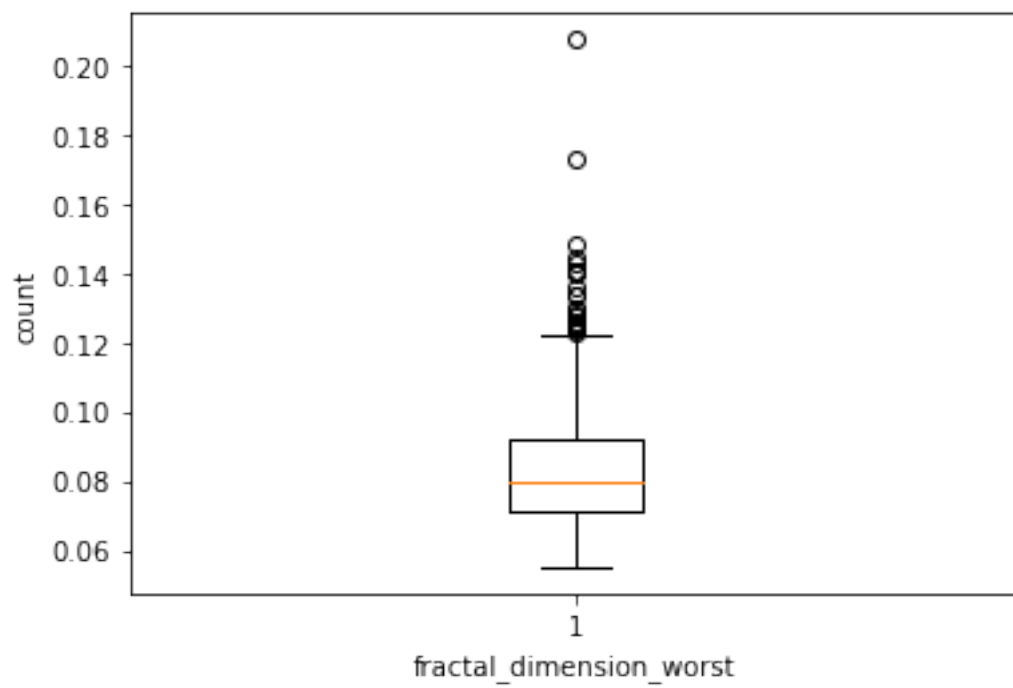
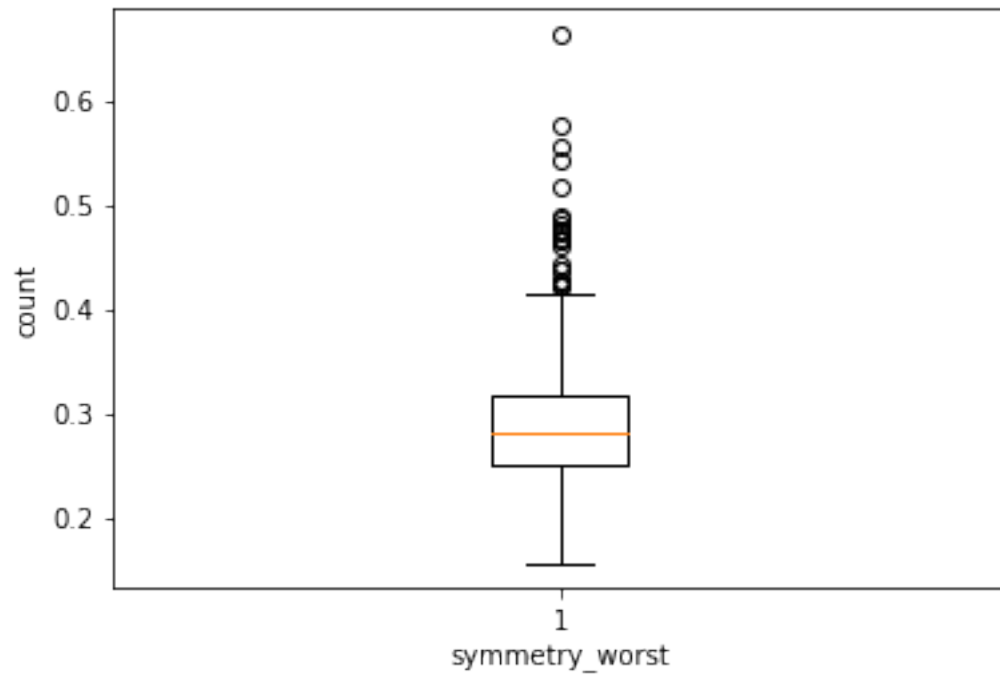










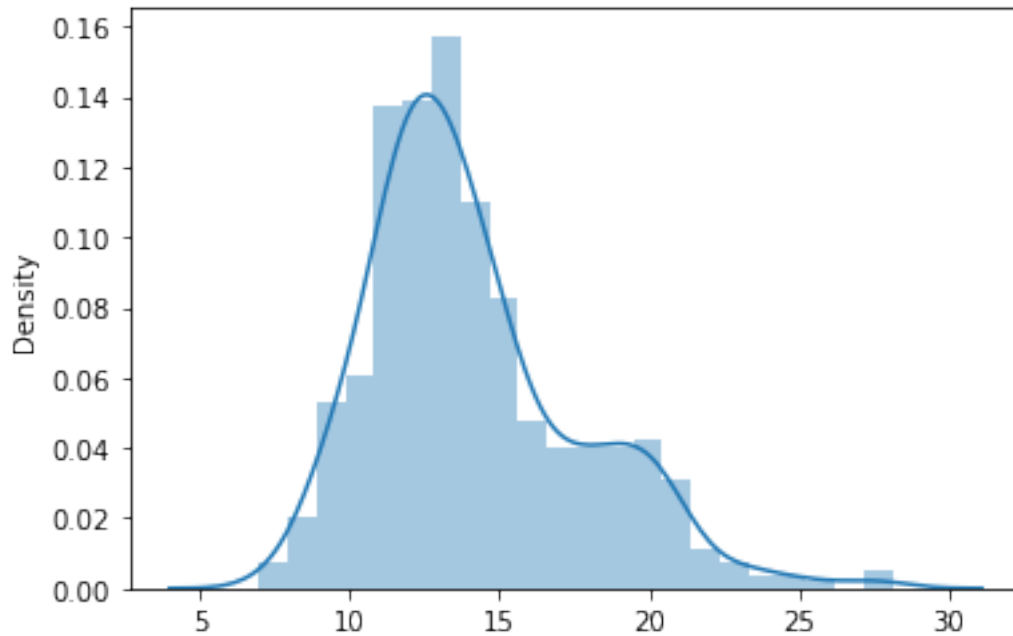


```
[23]: df.shape
```

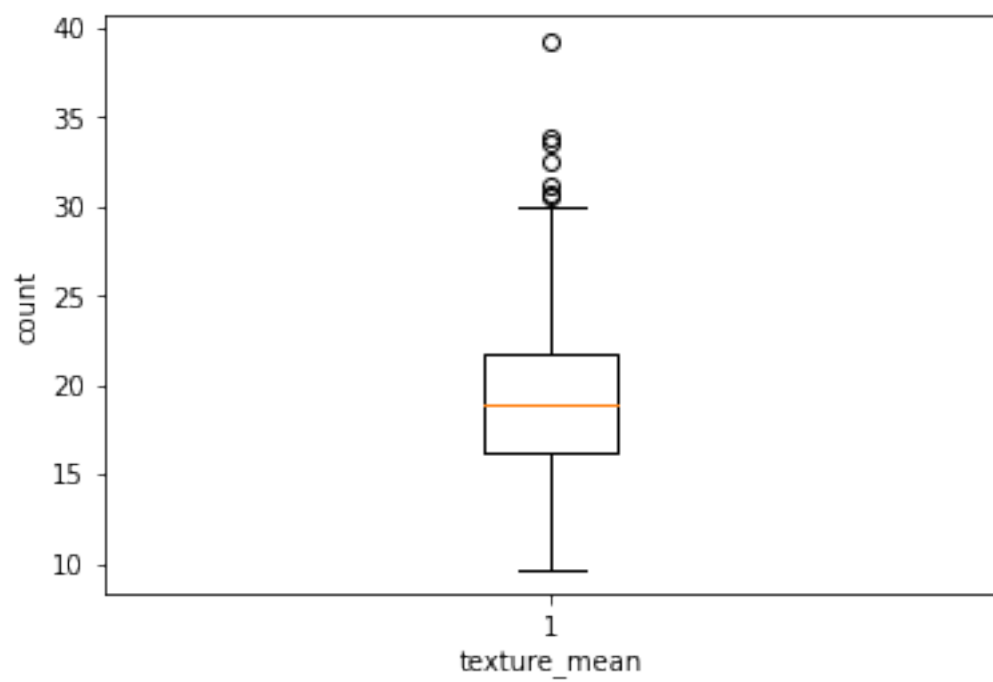
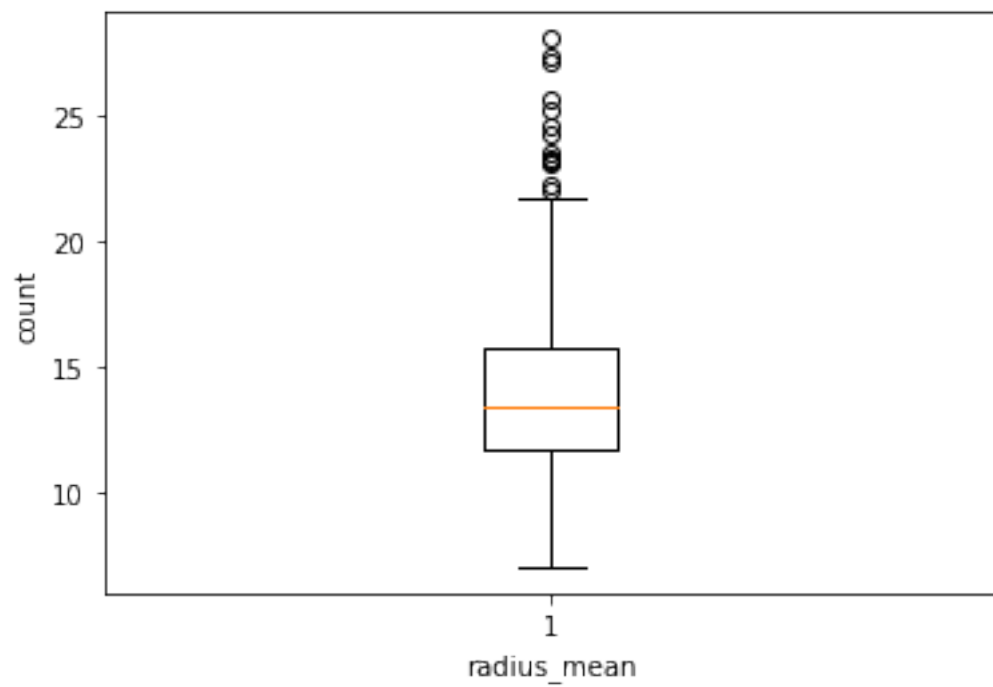
[23]: (569, 31)

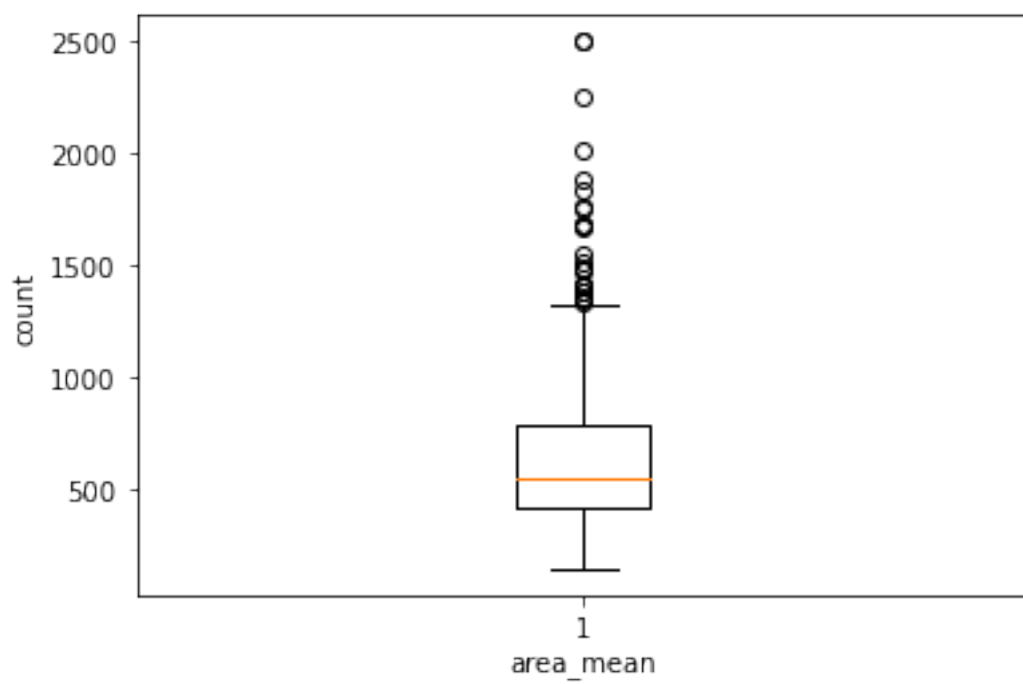
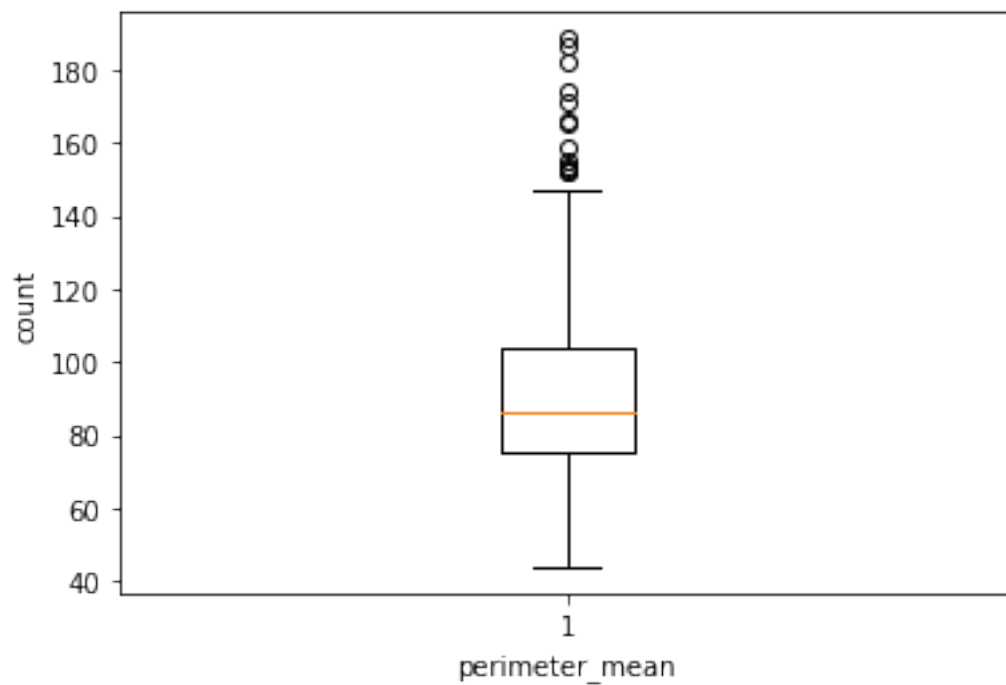
```
[24]: sns.distplot(x=df.radius_mean)
```

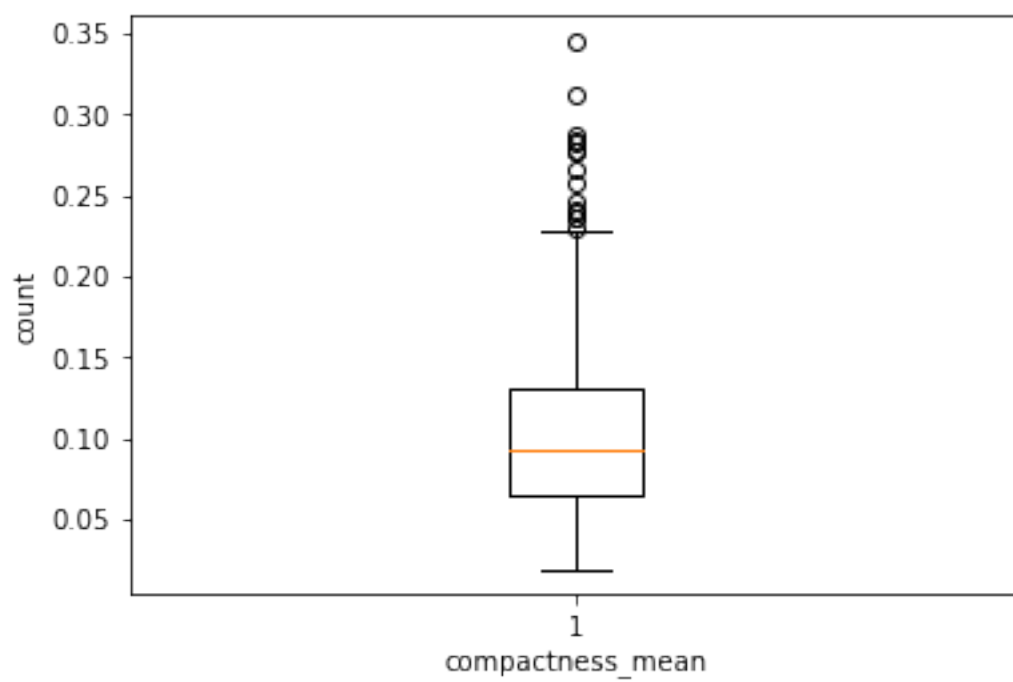
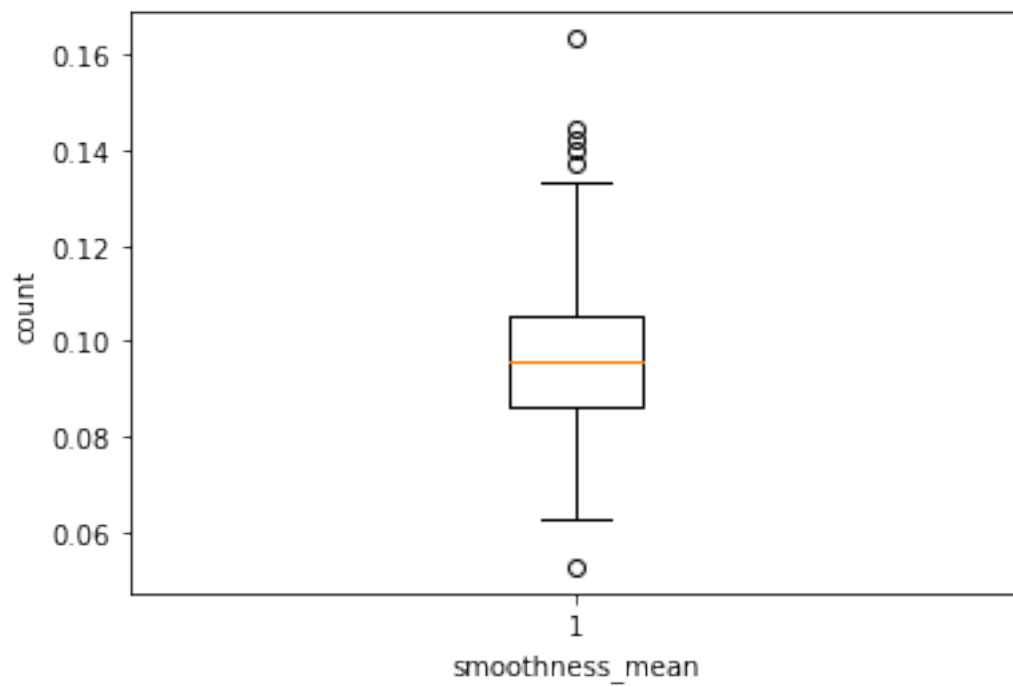
[24]: <AxesSubplot:ylabel='Density'>

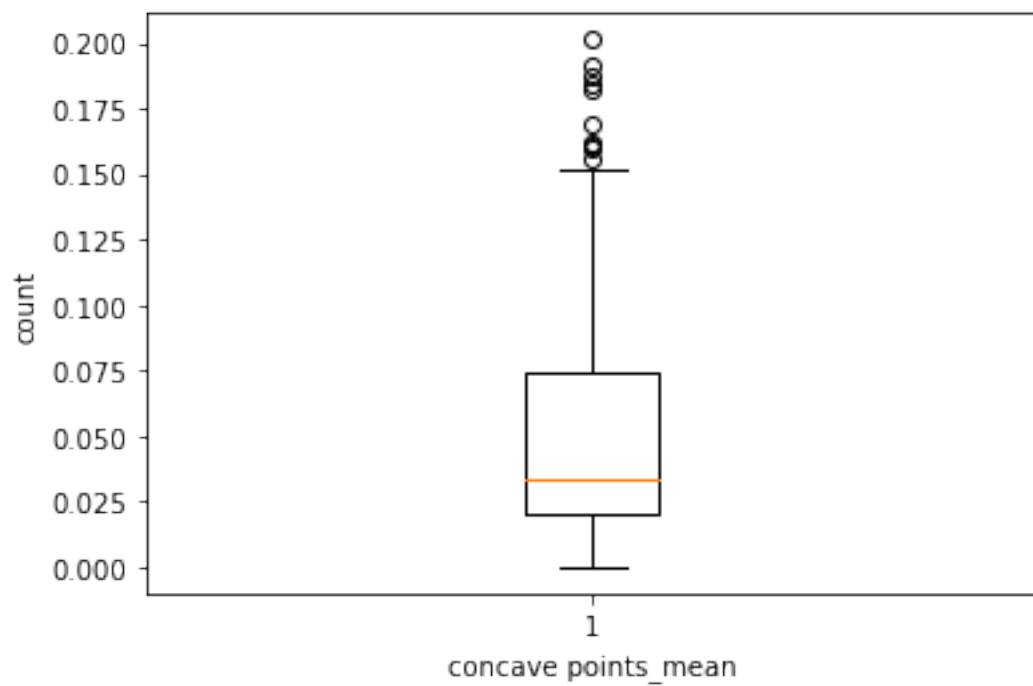
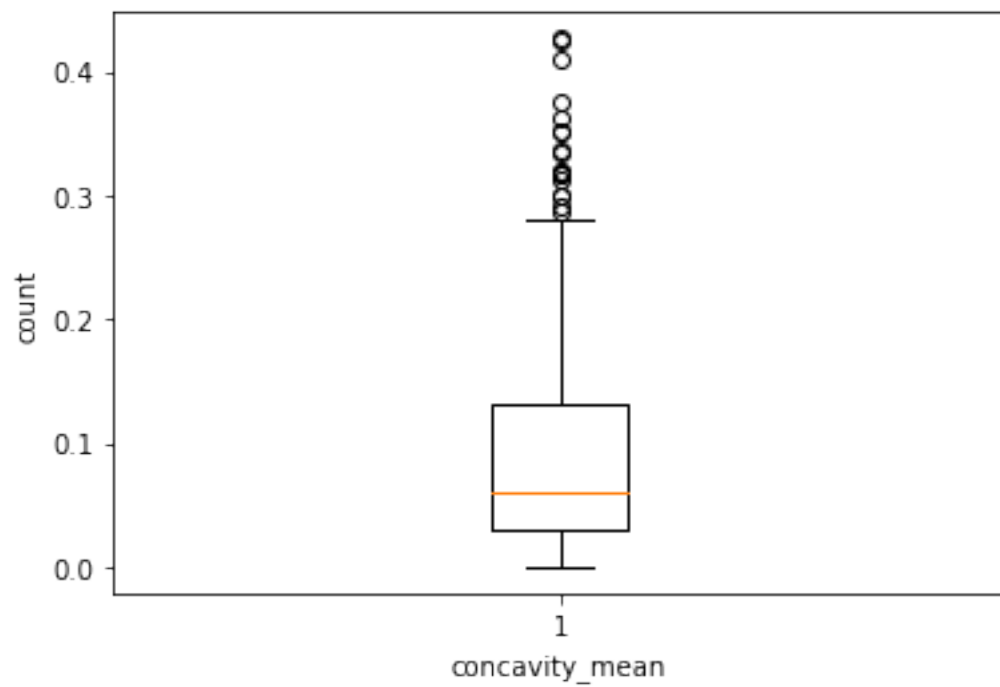


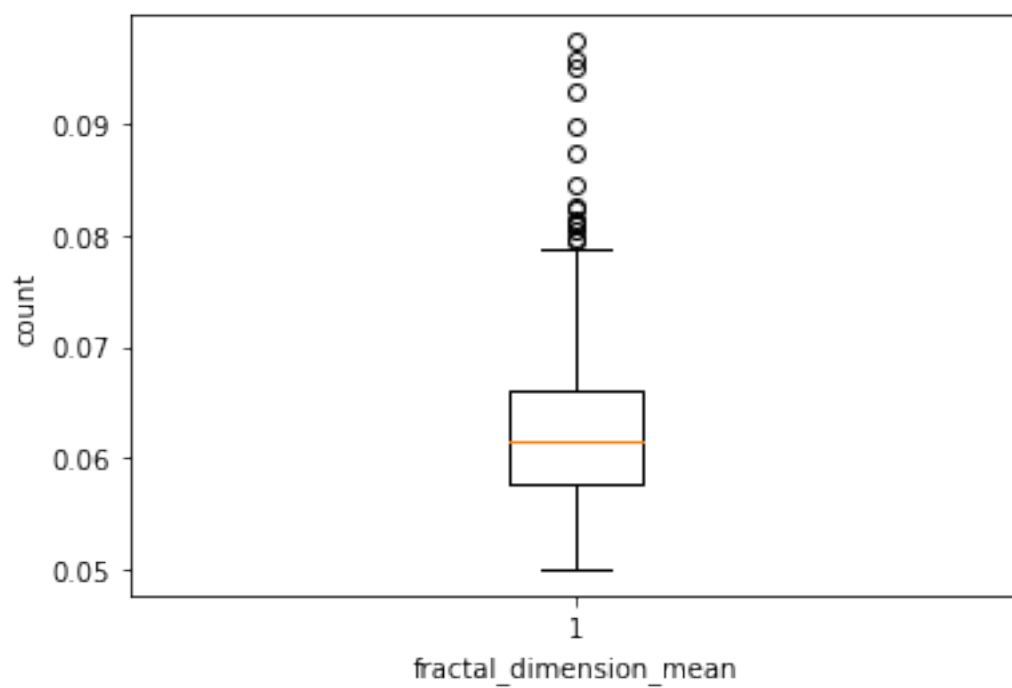
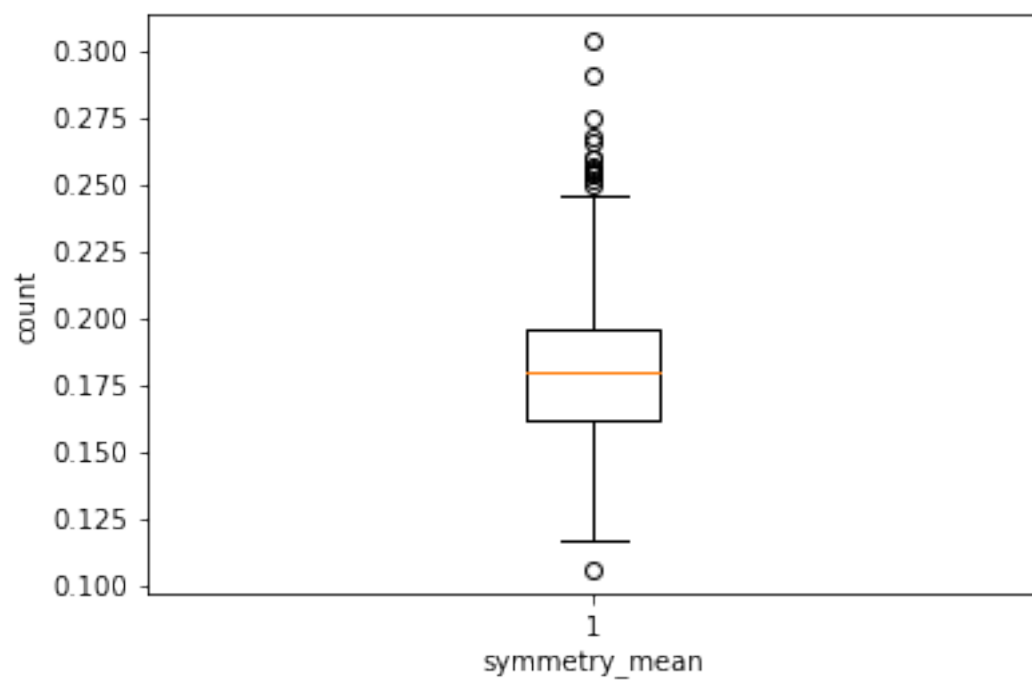
```
[25]: for col_name in df.columns:
        if(df[col_name].dtypes=='int64' or df[col_name].dtypes=='float64'):
            plt.boxplot(df[col_name] )
            plt.xlabel(col_name)
            plt.ylabel('count')
            plt.show()
```

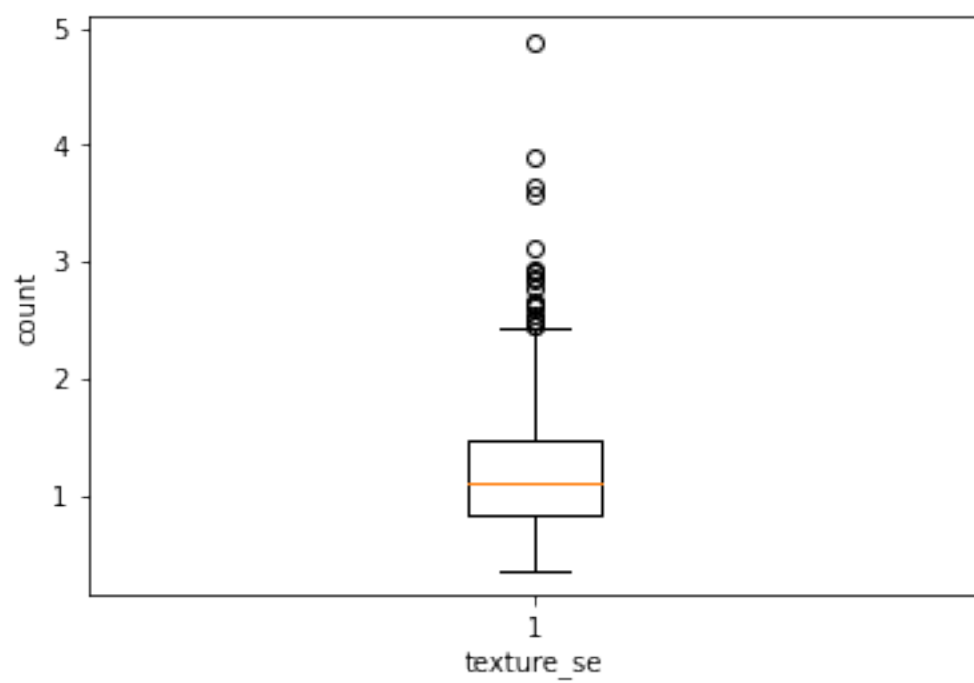
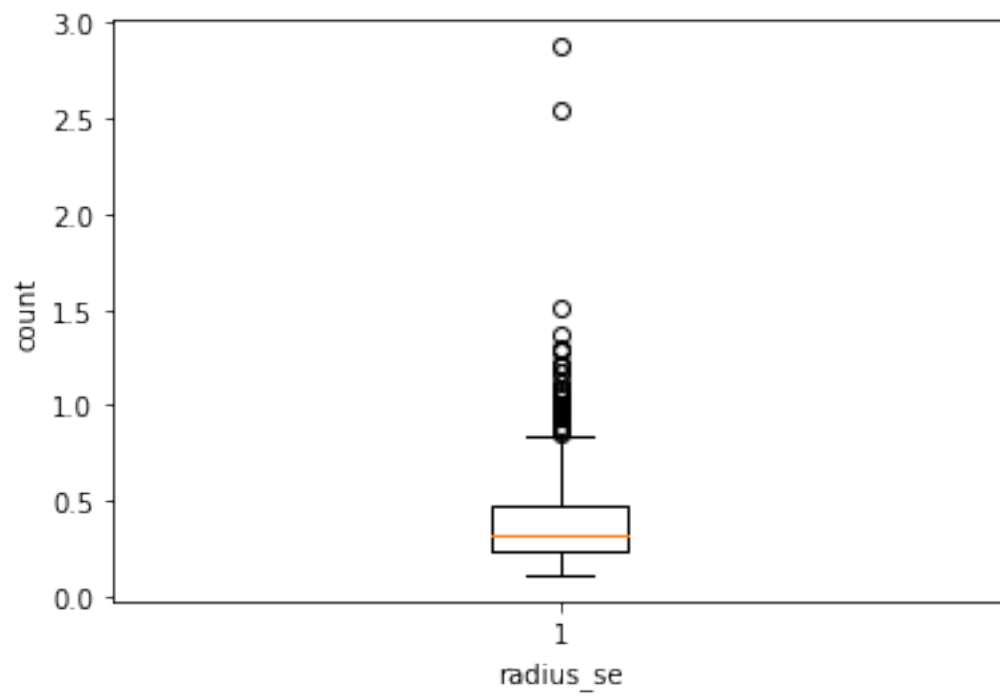


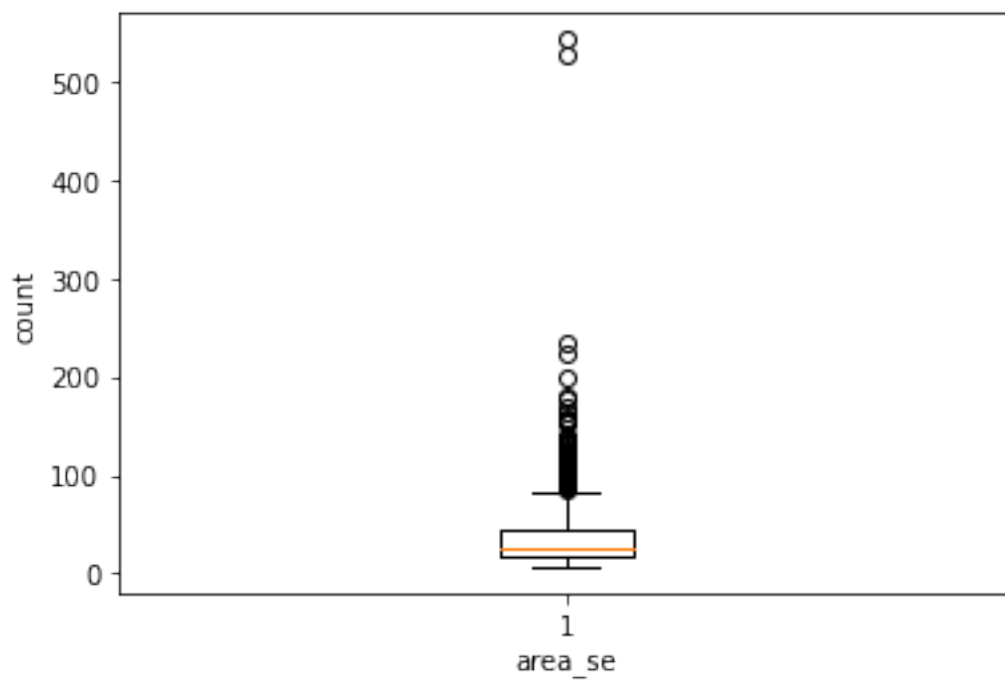
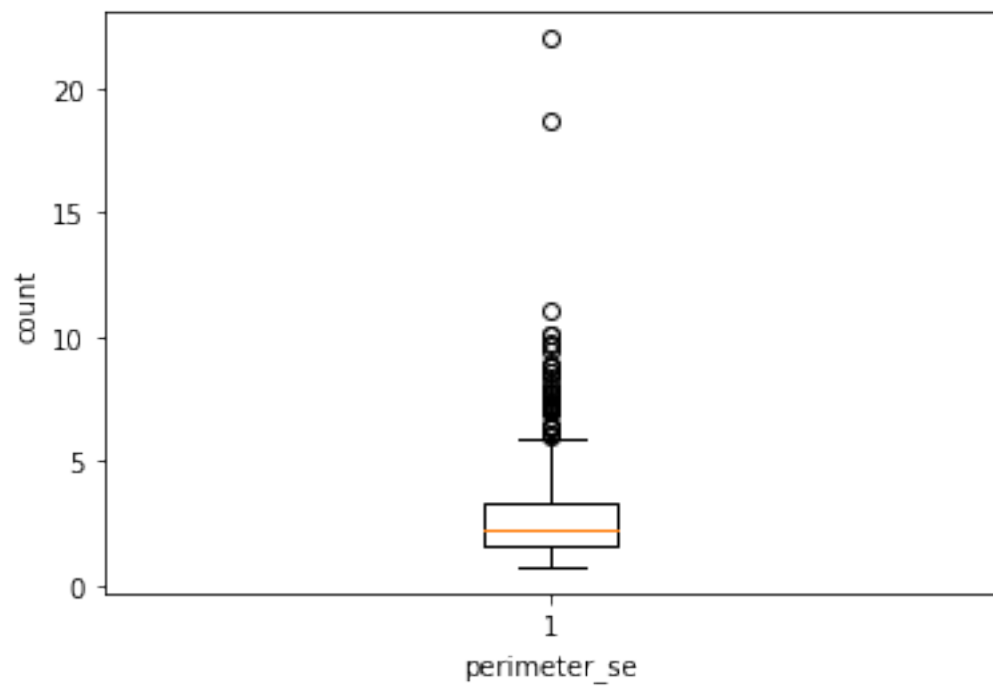


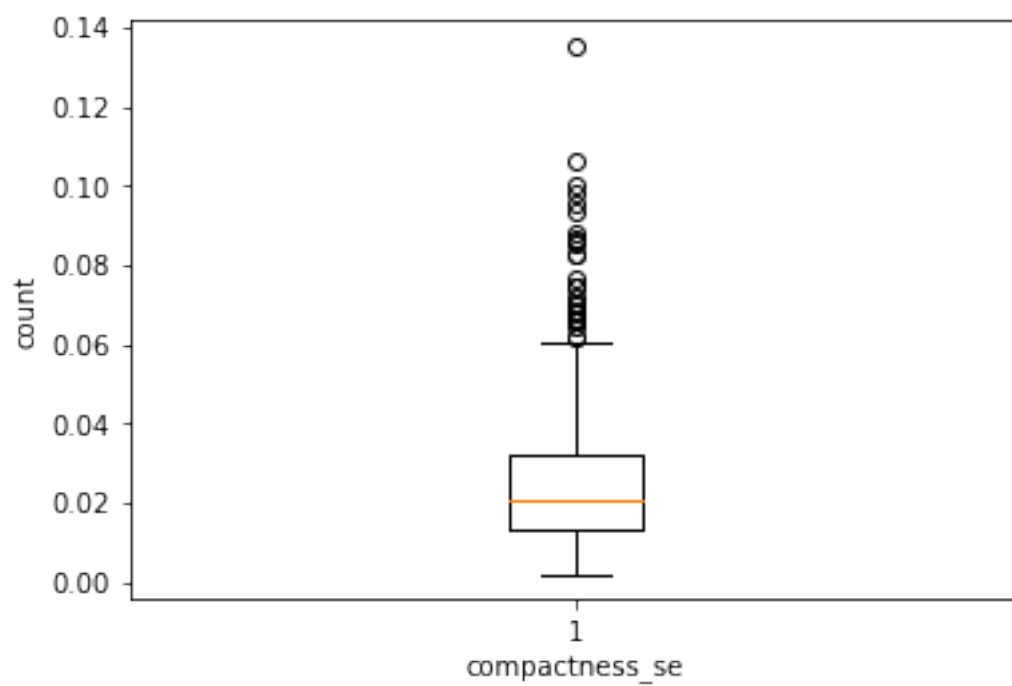
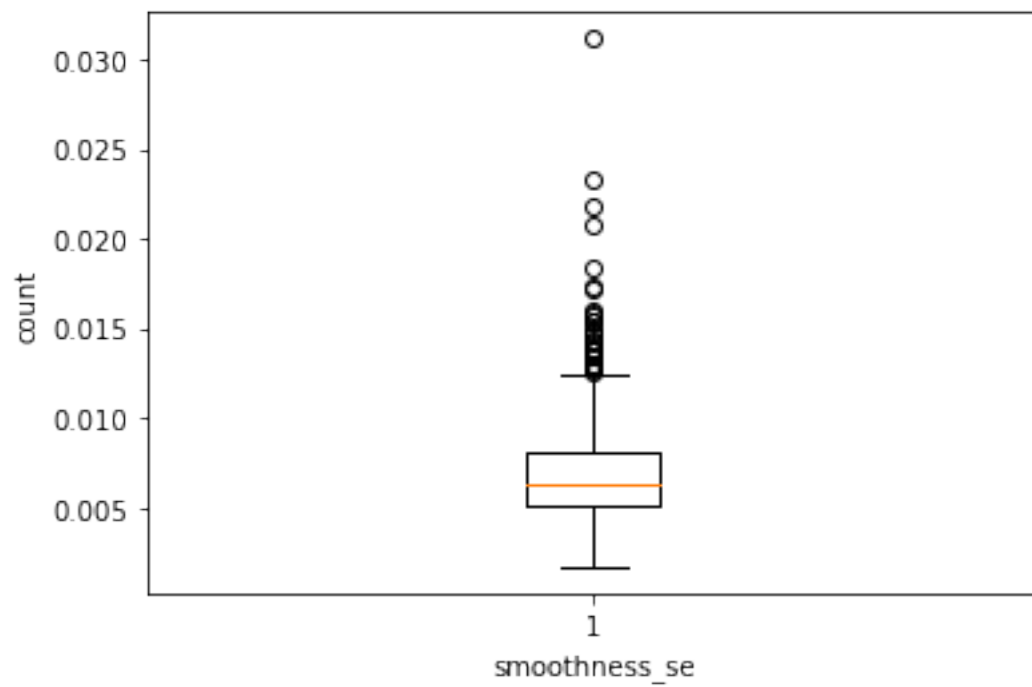


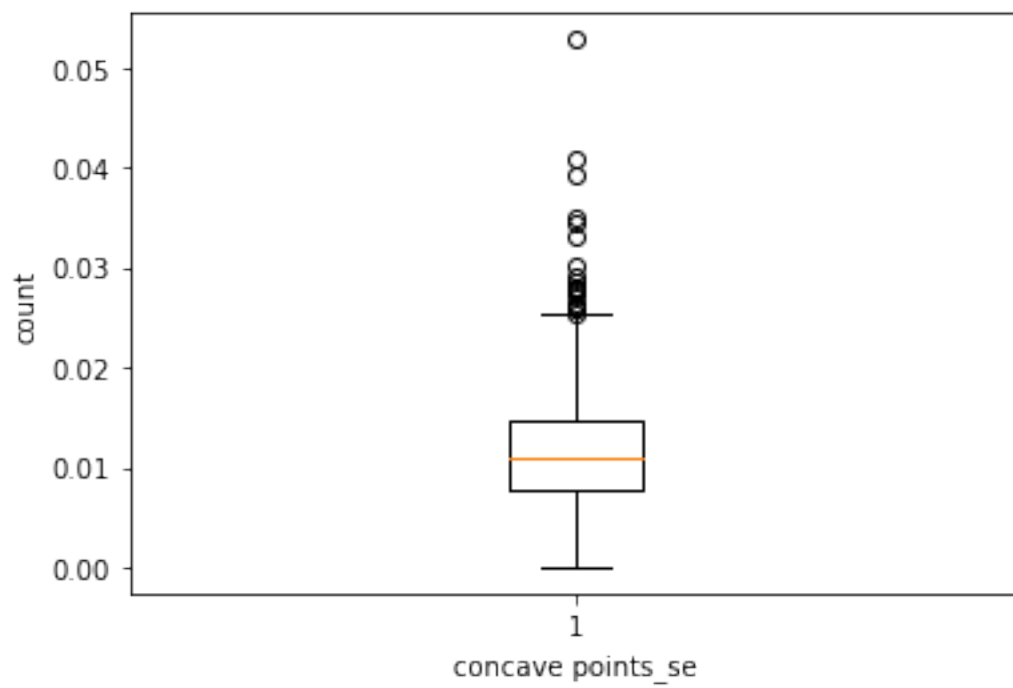
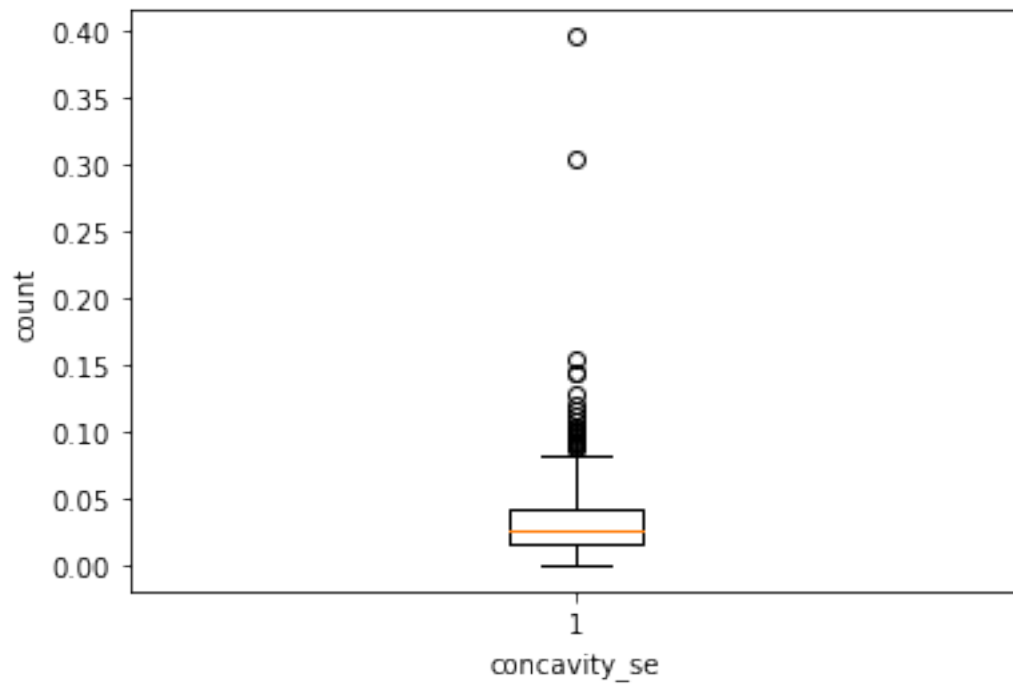


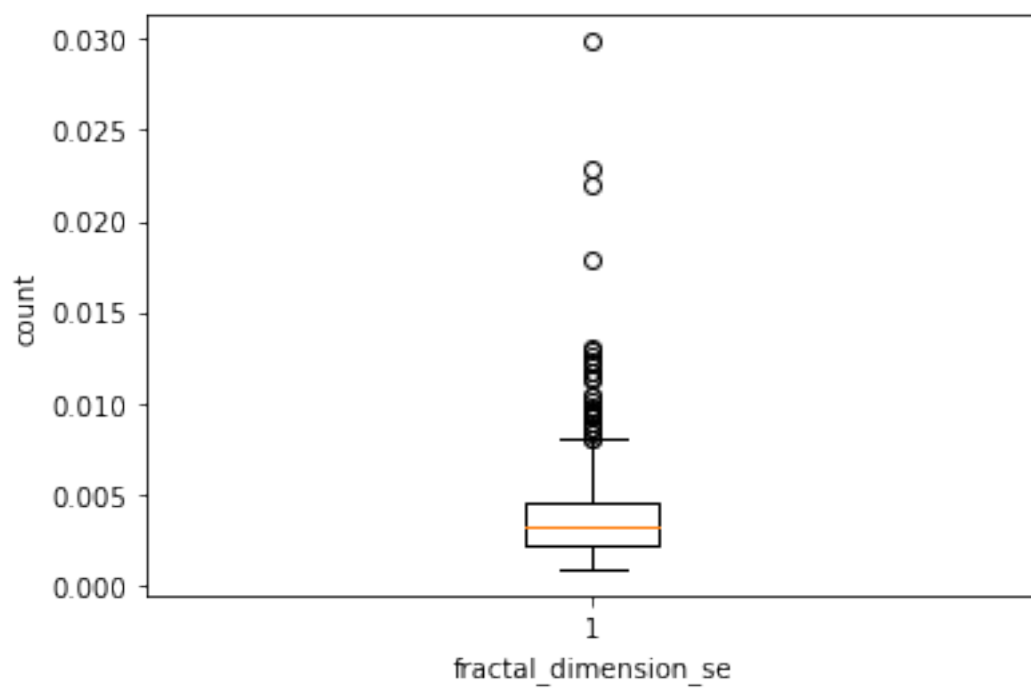
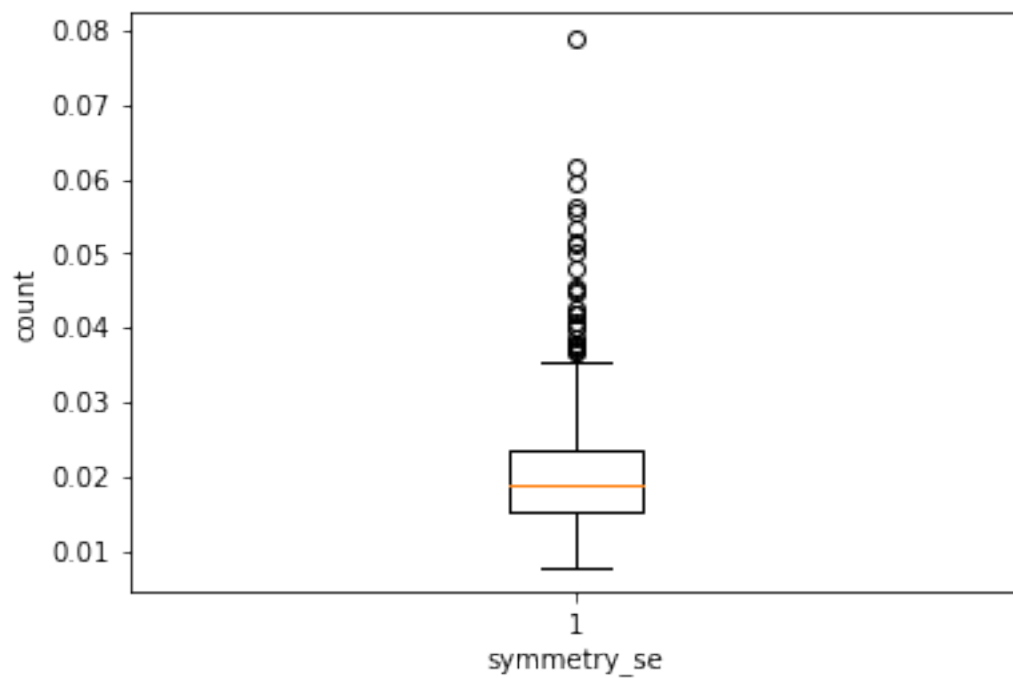


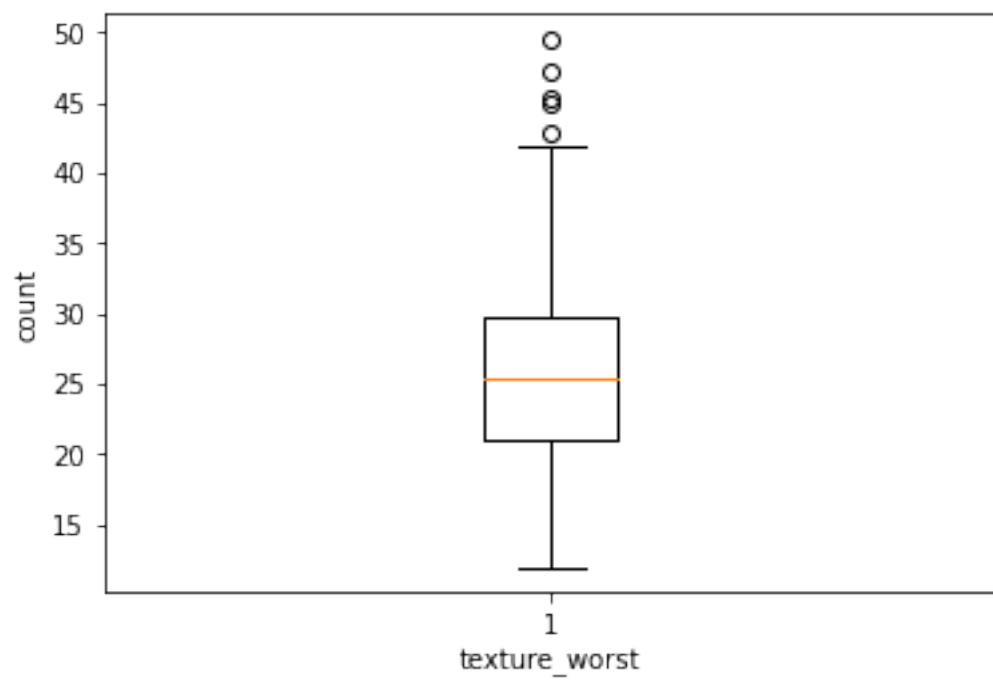
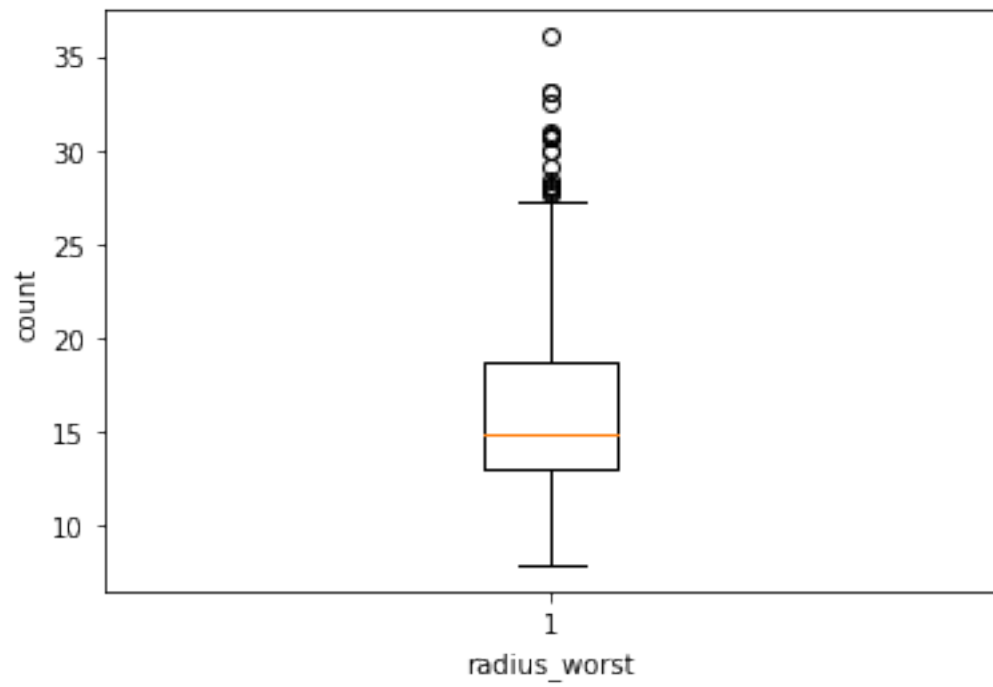


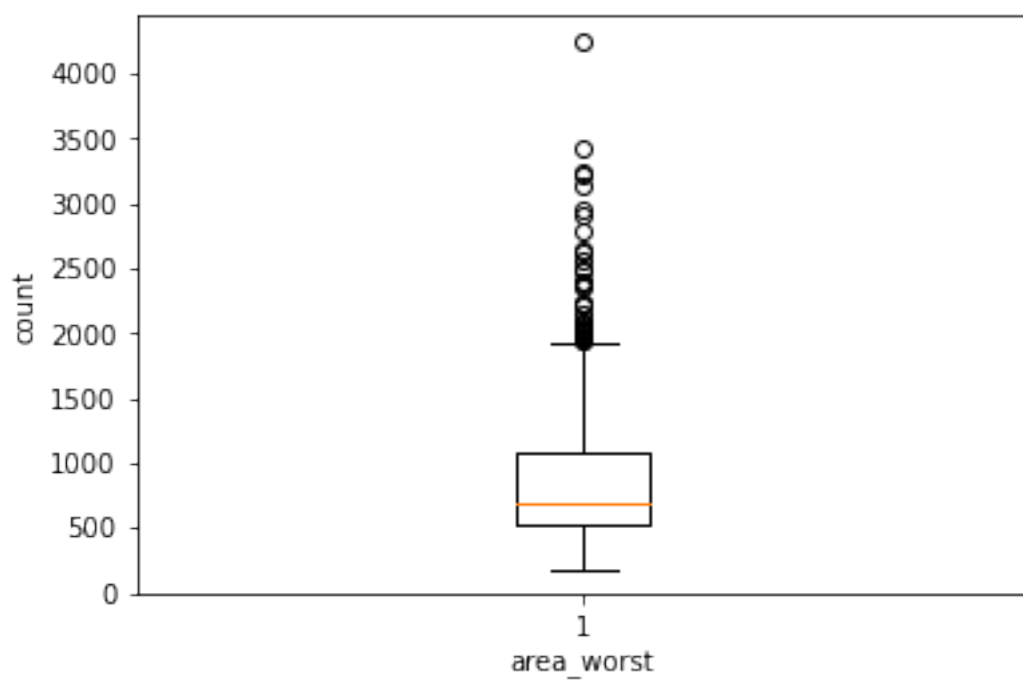
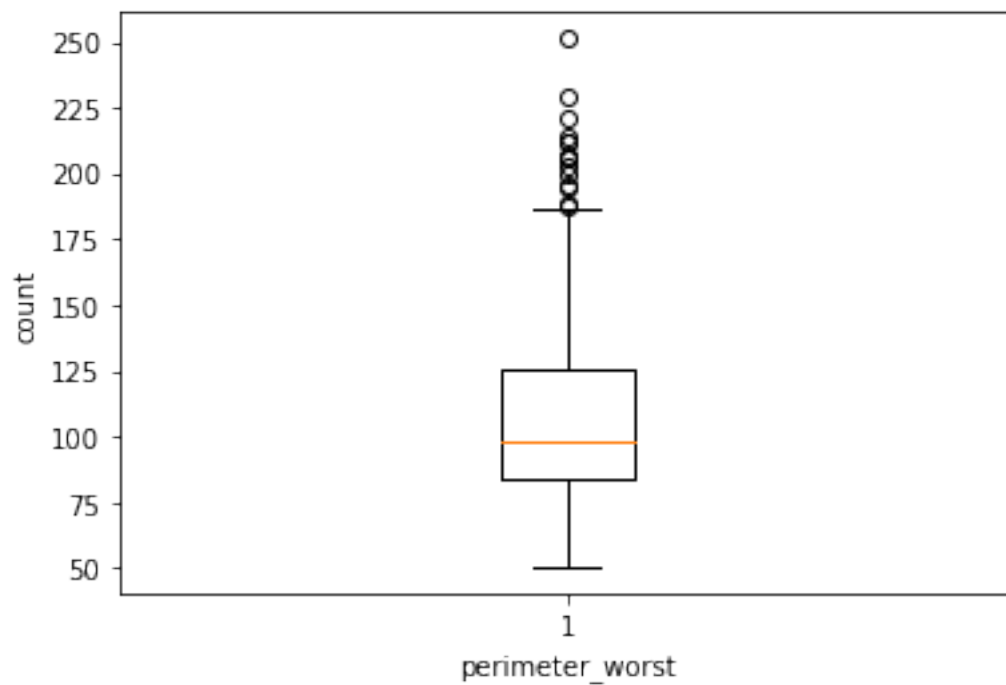


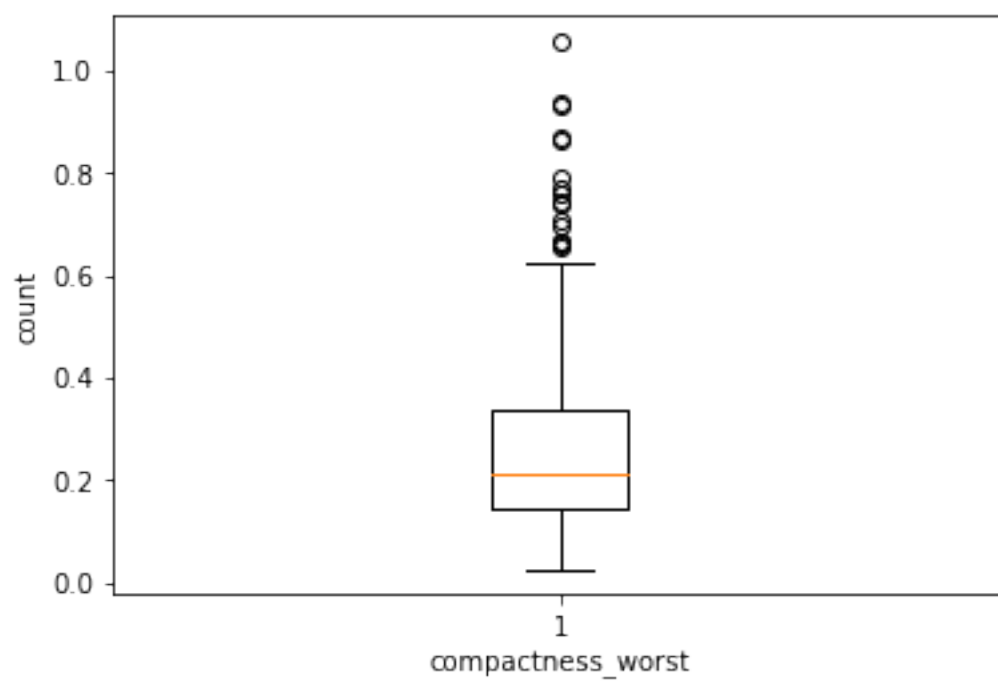
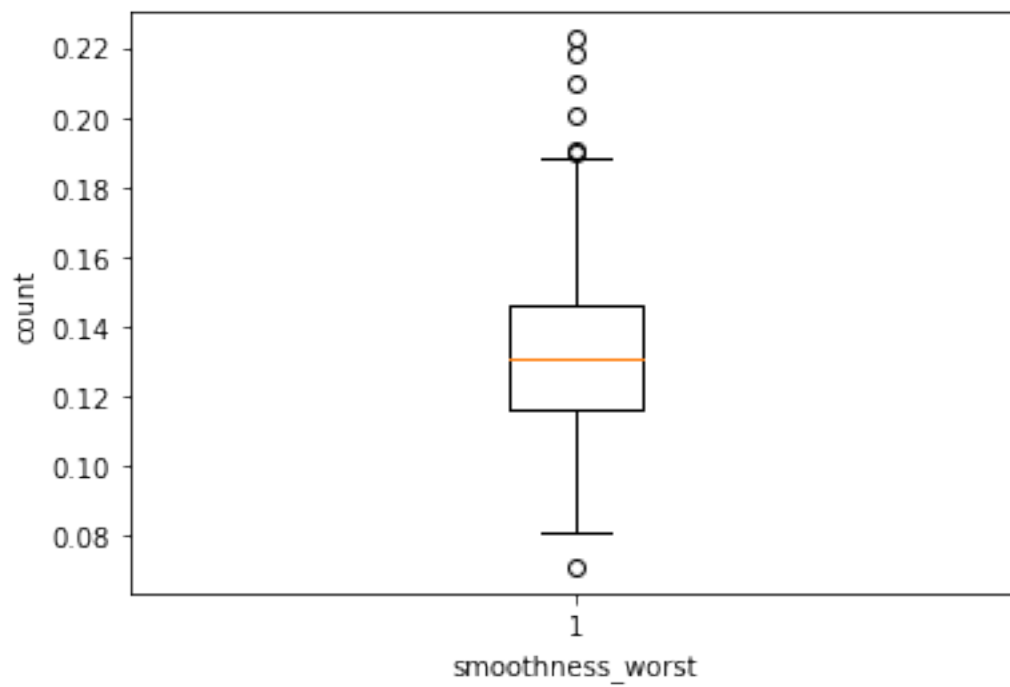


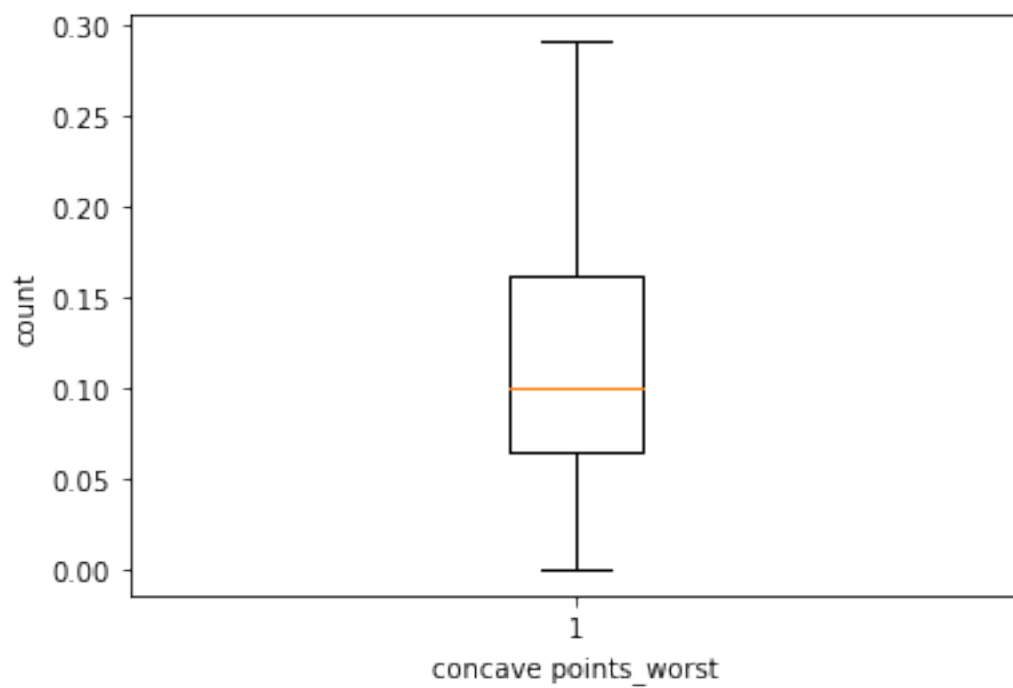
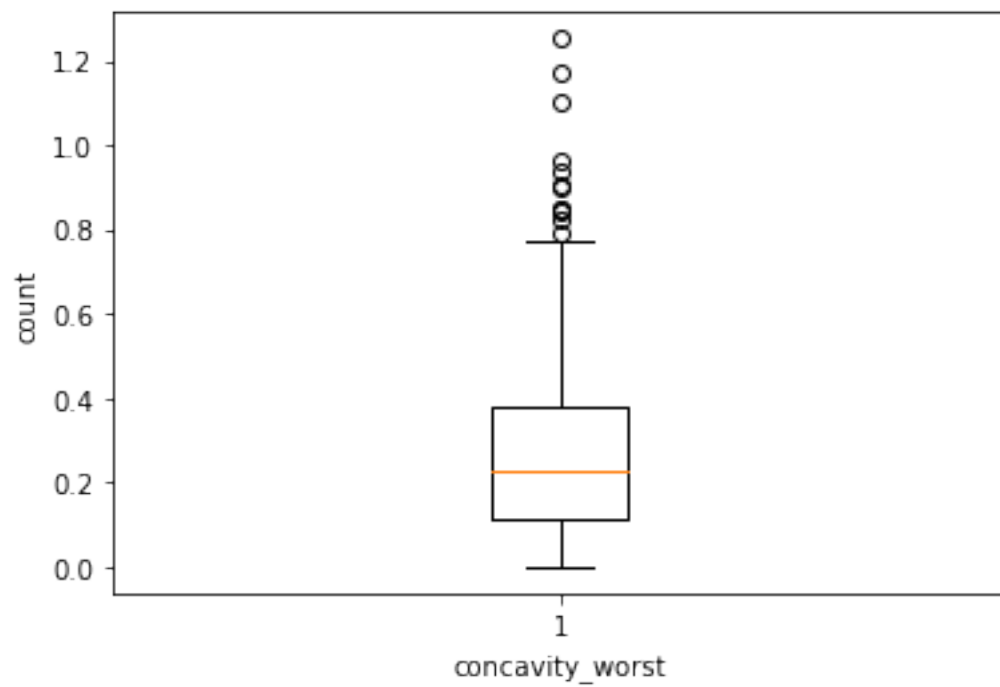


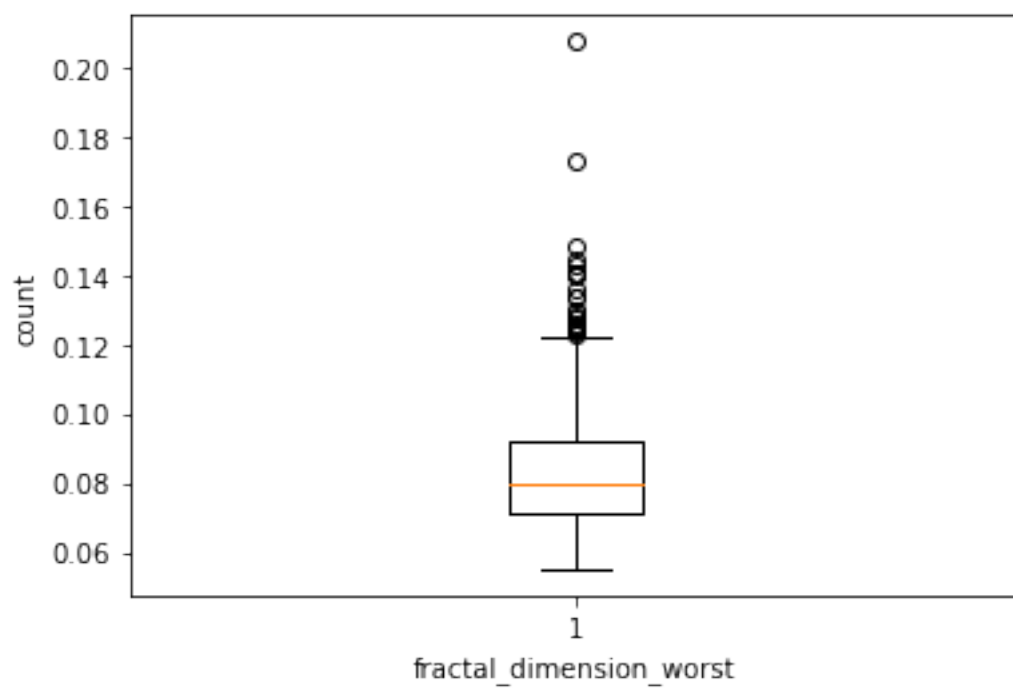
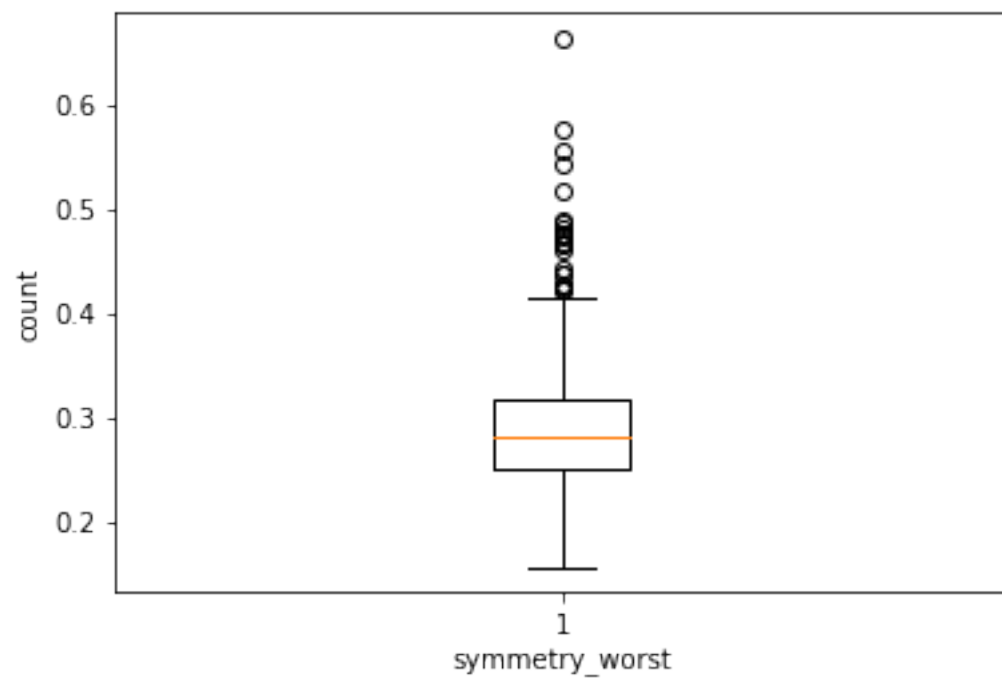












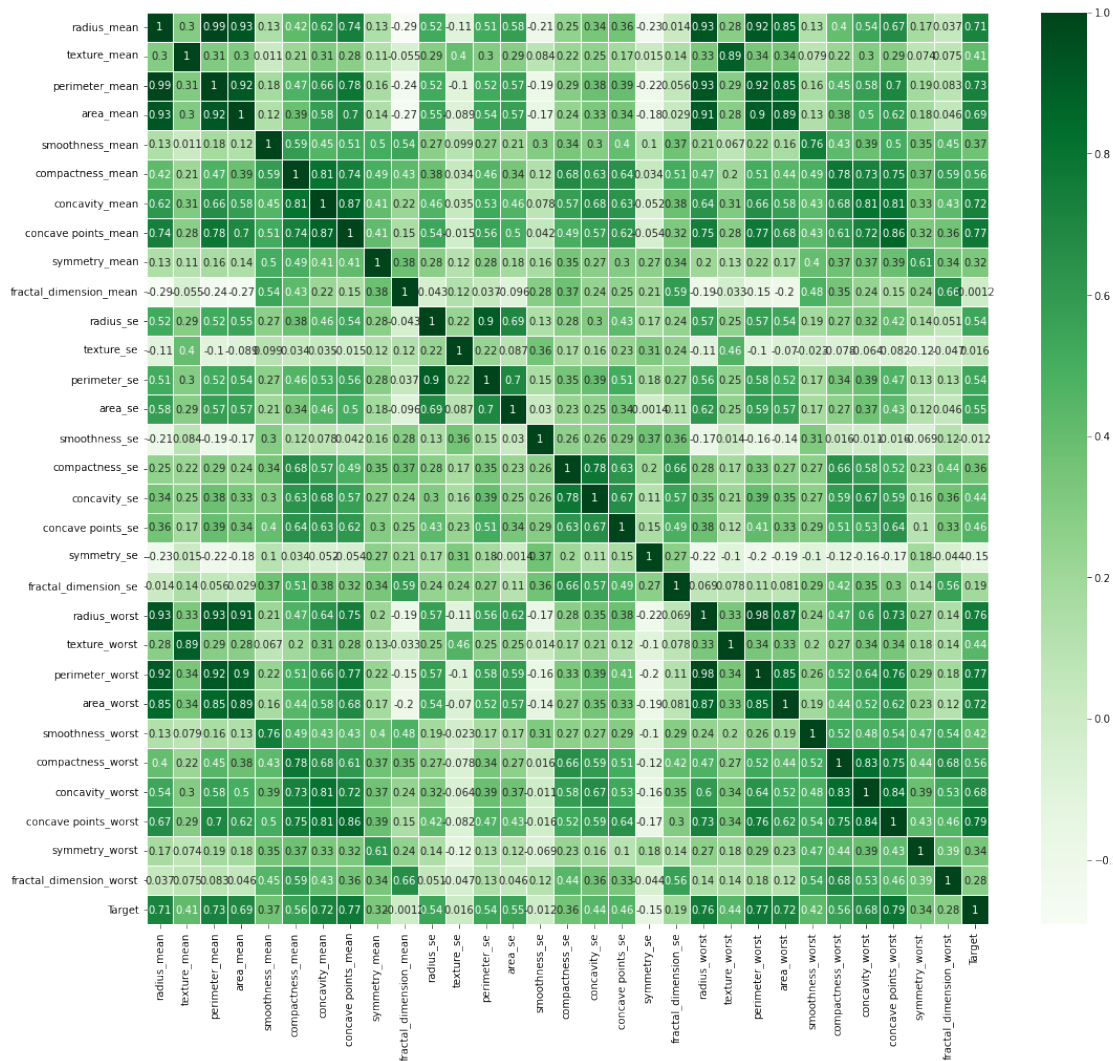
1 Get the values of Outliers

```
[26]: Q1 = df.quantile(0.25)
      Q3 = df.quantile(0.75)
      IQR = Q3 - Q1
      ((df < (Q1 - 1.5 * IQR)) | (df > (Q3 + 1.5 * IQR))).sum()
```

```
[26]: radius_mean      14
      texture_mean      7
      perimeter_mean    13
      area_mean         25
      smoothness_mean    6
      compactness_mean   16
      concavity_mean     18
      concave points_mean 10
      symmetry_mean      15
      fractal_dimension_mean 15
      radius_se         38
      texture_se        20
      perimeter_se      38
      area_se          65
      smoothness_se     30
      compactness_se    28
      concavity_se      22
      concave points_se  19
      symmetry_se       27
      fractal_dimension_se 28
      radius_worst      17
      texture_worst      5
      perimeter_worst    15
      area_worst        35
      smoothness_worst   7
      compactness_worst  16
      concavity_worst    12
      concave points_worst 0
      symmetry_worst     23
      fractal_dimension_worst 24
      Target            0
      dtype: int64
```

```
[45]: plt.figure(figsize=(18,16))
      sns.heatmap(df.corr(),annot=True,linewidth=.5,cmap='Greens')
```

```
[45]: <AxesSubplot:>
```



```
[43]: Q1 = df.quantile(0.25)
      Q3 = df.quantile(0.75)
      IQR = Q3 - Q1
      ((df < (Q1 - 1.5 * IQR)) | (df > (Q3 + 1.5 * IQR))).sum()
```

```
[43]: radius_mean      6
      texture_mean     2
      perimeter_mean   4
      area_mean       40
      smoothness_mean  1
      compactness_mean 6
      concavity_mean   8
      concave points_mean 9
      symmetry_mean    1
```

```

fractal_dimension_mean      7
radius_se                   28
texture_se                   5
perimeter_se                21
area_se                     52
smoothness_se               9
compactness_se              9
concavity_se                13
concave points_se           6
symmetry_se                 14
fractal_dimension_se        13
radius_worst                21
texture_worst                0
perimeter_worst             13
area_worst                  48
smoothness_worst            1
compactness_worst           12
concavity_worst              2
concave points_worst        0
symmetry_worst              11
fractal_dimension_worst     11
Target                      0
dtype: int64

```

2 Outlier Treatment

```

[41]: def outlier_detect(df):
        for i in df.describe().columns:
            Q1=df.describe().at['25%',i]
            Q3=df.describe().at['75%',i]
            IQR=Q3 - Q1
            LTV=Q1 - 1.5 * IQR
            UTV=Q3 + 1.5 * IQR
            x=np.array(df[i])
            p=[]
            for j in x:
                if j < LTV or j>UTV:
                    p.append(df[i].median())
                else:
                    p.append(j)
            df[i]=p
        return df

```

```

[30]: outlier_detect(df)

```

```

[30]:      radius_mean  texture_mean  perimeter_mean  area_mean  smoothness_mean  \
0          17.99      10.38      122.80      1001.0      0.11840
1          20.57      17.77      132.90      1326.0      0.08474
2          19.69      21.25      130.00      1203.0      0.10960
3          11.42      20.38       77.58      386.1      0.09587
4          20.29      14.34      135.10      1297.0      0.10030
..          ...          ...          ...          ...          ...
564         21.56      22.39      142.00      551.1      0.11100
565         20.13      28.25      131.20      1261.0      0.09780
566         16.60      28.08      108.30      858.1      0.08455
567         20.60      29.33      140.10      1265.0      0.11780
568          7.76      24.54       47.92      181.0      0.09587

      compactness_mean  concavity_mean  concave points_mean  symmetry_mean  \
0          0.09263      0.06154      0.14710      0.2419
1          0.07864      0.08690      0.07017      0.1812
2          0.15990      0.19740      0.12790      0.2069
3          0.09263      0.24140      0.10520      0.1792
4          0.13280      0.19800      0.10430      0.1809
..          ...          ...          ...          ...
564         0.11590      0.24390      0.13890      0.1726
565         0.10340      0.14400      0.09791      0.1752
566         0.10230      0.09251      0.05302      0.1590
567         0.09263      0.06154      0.15200      0.2397
568         0.04362      0.00000      0.00000      0.1587

      fractal_dimension_mean  ... texture_worst  perimeter_worst  area_worst  \
0          0.07871  ...      17.33      184.60      686.5
1          0.05667  ...      23.41      158.80      686.5
2          0.05999  ...      25.53      152.50      1709.0
3          0.06154  ...      26.50       98.87      567.7
4          0.05883  ...      16.67      152.20      1575.0
..          ...  ...          ...          ...          ...
564         0.05623  ...      26.40      166.10      686.5
565         0.05533  ...      38.25      155.00      1731.0
566         0.05648  ...      34.12      126.70      1124.0
567         0.07016  ...      39.42      184.60      1821.0
568         0.05884  ...      30.37       59.16      268.6

      smoothness_worst  compactness_worst  concavity_worst  \
0          0.16220      0.21190      0.7119
1          0.12380      0.18660      0.2416
2          0.14440      0.42450      0.4504
3          0.13130      0.21190      0.6869
4          0.13740      0.20500      0.4000
..          ...          ...          ...
564         0.14100      0.21130      0.4107

```

565	0.11660	0.19220	0.3215
566	0.11390	0.30940	0.3403
567	0.16500	0.21190	0.2267
568	0.08996	0.06444	0.0000

	concave	points_worst	symmetry_worst	fractal_dimension_worst	Target
0		0.2654	0.2822	0.11890	1
1		0.1860	0.2750	0.08902	1
2		0.2430	0.3613	0.08758	1
3		0.2575	0.2822	0.08004	1
4		0.1625	0.2364	0.07678	1
..	
564		0.2216	0.2060	0.07115	1
565		0.1628	0.2572	0.06637	1
566		0.1418	0.2218	0.07820	1
567		0.2650	0.4087	0.08004	1
568		0.0000	0.2871	0.07039	0

[569 rows x 31 columns]

3 Splitting the data

```
[32]: X= df.drop(columns='Target', axis=1)
      Y=df['Target']
      from sklearn.model_selection import train_test_split
      X_train, X_test, Y_train, Y_test=train_test_split(X,Y, test_size=20,
      ↪random_state=2)
```

4 Model Training

```
[33]: from sklearn.linear_model import LogisticRegression
      model=LogisticRegression()
```

```
[34]: model.fit(X_train,Y_train)
```

```
[34]: LogisticRegression()
```

5 Model Evaluation

```
[35]: #accuracy of training data
      from sklearn.metrics import accuracy_score
      x_train_prediction= model.predict(X_train)
      train_accuracy=accuracy_score(Y_train,x_train_prediction )
```

```
[36]: print('training accuracy =',train_accuracy*100 )
```

```
training accuracy = 91.07468123861567
```

```
[37]: #accuracy of test data
      from sklearn.metrics import accuracy_score
      x_test_prediction= model.predict(X_test)
      test_accuracy=accuracy_score(Y_test,x_test_prediction )
```

```
[38]: print('testing accuracy =',test_accuracy*100 )
```

```
testing accuracy = 85.0
```

6 building prection system

```
[40]: input_value= (17.99,10.38,122.8,1001,0.1184,0.2776,0.3001,0.1471,0.2419,0.
      ↪07871,1.095,0.9053,8.589,153.4,0.006399,0.04904,0.05373,0.01587,0.03003,0.
      ↪006193,25.38,17.33,184.6,2019,0.1622,0.6656,0.7119,0.2654,0.4601,0.1189)
```

```
new_input=np.asarray(input_value)
#reshaping
reshape_input= new_input.reshape(1,-1)
prediction=model.predict(reshape_input)
print(prediction)
if (prediction[0]==1):
    print("the brist cancer is malignant")
else:
    print("the brist cancer is benign")
```

```
[1]
```

```
the brist cancer is malignant
```

```
[ ]:
```