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

In [2]: ds=pd.read_csv("C:/Users/USER/Desktop/Datasets/wbcd.csv")
 ds

Out[2]:		id	diagnosis	radius_mean	texture_mean	perimeter_mean	area_mean	smoothness_r
	0	87139402	В	12.32	12.39	78.85	464.1	0.1
	1	8910251	В	10.60	18.95	69.28	346.4	0.0
	2	905520	В	11.04	16.83	70.92	373.2	0.1
	3	868871	В	11.28	13.39	73.00	384.8	0.1
	4	9012568	В	15.19	13.21	97.65	711.8	0.0
				•••	•••			
	564	911320502	В	13.17	18.22	84.28	537.3	0.0
	565	898677	В	10.26	14.71	66.20	321.6	0.0
	566	873885	М	15.28	22.41	98.92	710.6	0.0
	567	911201	В	14.53	13.98	93.86	644.2	0.1
	568	9012795	М	21.37	15.10	141.30	1386.0	0.1
			_					

569 rows × 32 columns

In [3]: del ds["id"]

[n [4]: ds

Out	[4]	
Out	4	

	diagnosis	radius_mean	texture_mean	perimeter_mean	area_mean	smoothness_mean	comp
0	В	12.32	12.39	78.85	464.1	0.10280	
1	В	10.60	18.95	69.28	346.4	0.09688	
2	В	11.04	16.83	70.92	373.2	0.10770	
3	В	11.28	13.39	73.00	384.8	0.11640	
4	В	15.19	13.21	97.65	711.8	0.07963	
564	В	13.17	18.22	84.28	537.3	0.07466	
565	В	10.26	14.71	66.20	321.6	0.09882	
566	М	15.28	22.41	98.92	710.6	0.09057	
567	В	14.53	13.98	93.86	644.2	0.10990	
568	М	21.37	15.10	141.30	1386.0	0.10010	

569 rows × 31 columns

4

In [5]: ds.describe()

	radius_mean	texture_mean	perimeter_mean	area_mean	smoothness_mean	compactness_
count	569.000000	569.000000	569.000000	569.000000	569.000000	569.0
mean	14.127292	19.289649	91.969033	654.889104	0.096360	0.1
std	3.524049	4.301036	24.298981	351.914129	0.014064	0.0
min	6.981000	9.710000	43.790000	143.500000	0.052630	0.0
25%	11.700000	16.170000	75.170000	420.300000	0.086370	0.0
50%	13.370000	18.840000	86.240000	551.100000	0.095870	0.0
75%	15.780000	21.800000	104.100000	782.700000	0.105300	0.1
max	28.110000	39.280000	188.500000	2501.000000	0.163400	0.3

8 rows × 30 columns

4

In [6]: ds.median(numeric_only=True)

Out[6]:	radius_mean	13.370000
	texture_mean	18.840000
	perimeter_mean	86.240000
	area_mean	551.100000
	smoothness_mean	0.095870
	compactness_mean	0.092630
	concavity_mean	0.061540
	points_mean	0.033500
	symmetry_mean	0.179200
	dimension_mean	0.061540
	radius_se	0.324200
	texture_se	1.108000
	perimeter_se	2.287000
	area_se	24.530000
	smoothness_se	0.006380
	compactness_se	0.020450
	concavity_se	0.025890
	points_se	0.010930
	symmetry_se	0.018730
	<pre>dimension_se</pre>	0.003187
	radius_worst	14.970000
	texture_worst	25.410000
	perimeter_worst	97.660000
	area_worst	686.500000
	smoothness_worst	0.131300
	compactness_worst	0.211900
	concavity_worst	0.226700
	points_worst	0.099930
	symmetry_worst	0.282200
	<pre>dimension_worst dtype: float64</pre>	0.080040

In [7]: ds.dtypes

Out[7]:	diagnosis	object
	radius_mean	float64
	texture_mean	float64
	perimeter_mean	float64
	area_mean	float64
	smoothness_mean	float64
	compactness_mean	float64
	concavity_mean	float64
	points_mean	float64

symmetry_mean dimension_mean float64 radius_se float64 float64 texture_se

float64

float64 perimeter_se area_se float64 float64 smoothness se

float64 compactness_se concavity_se float64 float64 points_se symmetry_se float64

dimension_se float64 radius worst float64 float64 texture_worst

perimeter_worst float64 area_worst float64 float64 smoothness_worst compactness_worst float64

concavity worst float64 points_worst float64 symmetry_worst float64 dimension_worst float64

dtype: object

In [8]: ds.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 569 entries, 0 to 568
Data columns (total 31 columns):

#	Column	Non-Null Count	Dtype
		56011	
0	diagnosis	569 non-null	object
1	radius_mean	569 non-null	float64
2	texture_mean	569 non-null	float64
3	perimeter_mean	569 non-null	float64
4	area_mean	569 non-null	float64
5	smoothness_mean	569 non-null	float64
6	compactness_mean	569 non-null	float64
7	concavity_mean	569 non-null	float64
8	points_mean	569 non-null	float64
9	symmetry_mean	569 non-null	float64
10	dimension_mean	569 non-null	float64
11	radius_se	569 non-null	float64
12	texture_se	569 non-null	float64
13	perimeter_se	569 non-null	float64
14	area_se	569 non-null	float64
15	smoothness_se	569 non-null	float64
16	compactness_se	569 non-null	float64
17	concavity_se	569 non-null	float64
18	points_se	569 non-null	float64
19	symmetry_se	569 non-null	float64
20	dimension_se	569 non-null	float64
21	radius_worst	569 non-null	float64
22	texture_worst	569 non-null	float64
23	perimeter_worst	569 non-null	float64
24	area_worst	569 non-null	float64
25	smoothness_worst	569 non-null	float64
26	compactness_worst	569 non-null	float64
27	concavity_worst	569 non-null	float64
28	points_worst	569 non-null	float64
29	symmetry_worst	569 non-null	float64
30	dimension_worst	569 non-null	float64
	es: float64(30), oh		

dtypes: float64(30), object(1)

memory usage: 137.9+ KB

```
In [9]: ds.isnull().sum()
 Out[9]: diagnosis
                                0
          radius_mean
                                0
                                0
          texture_mean
                                0
          perimeter mean
          area_mean
                                0
          smoothness_mean
                                0
                                0
          compactness_mean
          concavity_mean
                                0
                                0
          points mean
          symmetry_mean
                                0
                                0
          dimension_mean
          radius_se
                                0
                                0
          texture_se
          perimeter_se
          area_se
                                0
                                0
          smoothness se
                                0
          compactness_se
          concavity_se
                                0
                                0
          points_se
          symmetry_se
          dimension_se
                                0
          radius worst
                                0
                                0
          texture_worst
          perimeter_worst
                                0
          area_worst
          smoothness_worst
          compactness_worst
                                0
                                0
          concavity worst
                                0
          points_worst
          symmetry_worst
                                0
                                0
          dimension_worst
          dtype: int64
In [10]: ds["diagnosis"].value_counts()
Out[10]: B
               357
               212
```

```
localhost:8889/notebooks/K Nearest Neighbor .ipynb
```

Name: diagnosis, dtype: int64

```
In [15]: |ds.hist()
Out[15]: array([[<AxesSubplot:title={'center':'radius_mean'}>,
                  <AxesSubplot:title={'center':'texture mean'}>,
                  <AxesSubplot:title={'center':'perimeter_mean'}>,
                  <AxesSubplot:title={'center':'area mean'}>,
                  <AxesSubplot:title={'center':'smoothness mean'}>],
                 [<AxesSubplot:title={'center':'compactness mean'}>,
                  <AxesSubplot:title={'center':'concavity mean'}>,
                  <AxesSubplot:title={'center':'points_mean'}>,
                  <AxesSubplot:title={'center':'symmetry mean'}>,
                  <AxesSubplot:title={'center':'dimension mean'}>],
                 [<AxesSubplot:title={'center':'radius_se'}>,
                  <AxesSubplot:title={'center':'texture_se'}>,
                  <AxesSubplot:title={'center':'perimeter_se'}>,
                  <AxesSubplot:title={'center':'area se'}>,
                  <AxesSubplot:title={'center':'smoothness_se'}>],
                 [<AxesSubplot:title={'center':'compactness se'}>,
                  <AxesSubplot:title={'center':'concavity_se'}>,
                  <AxesSubplot:title={'center':'points_se'}>,
                  <AxesSubplot:title={'center':'symmetry se'}>,
                  <AxesSubplot:title={'center':'dimension_se'}>],
                 [<AxesSubplot:title={'center':'radius worst'}>,
                  <AxesSubplot:title={'center':'texture worst'}>,
                  <AxesSubplot:title={'center':'perimeter worst'}>,
                  <AxesSubplot:title={'center':'area_worst'}>,
                  <AxesSubplot:title={'center':'smoothness_worst'}>],
                 [<AxesSubplot:title={'center':'compactness worst'}>,
                  <AxesSubplot:title={'center':'concavity_worst'}>,
                  <AxesSubplot:title={'center':'points worst'}>,
                  <AxesSubplot:title={'center':'symmetry worst'}>,
                  <AxesSubplot:title={'center':'dimension_worst'}>]], dtype=object)
             radius meatexture meanimeter meanea meanoothness mean
           100
           congpla
           100
             0
                     el o t
            250
            com
                                          δV
                                $æ
                                                   —30€P
            2000
             0ra/e
                                        ₩oorsal
           100
           congp
            200
                     1|do
             0
                       0
                                            0.25 0.50
                              1
                                 0.00
                                       0.25
                                                       0.1
                                                           0.2
```

localhost:8889/notebooks/K Nearest Neighbor .ipynb

In [16]: x=pd.DataFrame(ds.iloc[:,1:])

y=pd.DataFrame(ds.iloc[:,0])

In [17]: x

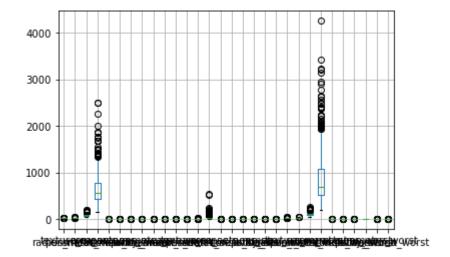
Out[17]:

	radius_mean	texture_mean	perimeter_mean	area_mean	smoothness_mean	compactness_me
0	12.32	12.39	78.85	464.1	0.10280	0.069
1	10.60	18.95	69.28	346.4	0.09688	0.114
2	11.04	16.83	70.92	373.2	0.10770	0.078
3	11.28	13.39	73.00	384.8	0.11640	0.113
4	15.19	13.21	97.65	711.8	0.07963	0.069
564	13.17	18.22	84.28	537.3	0.07466	0.059
565	10.26	14.71	66.20	321.6	0.09882	0.091
566	15.28	22.41	98.92	710.6	0.09057	0.105
567	14.53	13.98	93.86	644.2	0.10990	0.092
568	21.37	15.10	141.30	1386.0	0.10010	0.151

569 rows × 30 columns

In [18]: boxplot=x.boxplot()
boxplot

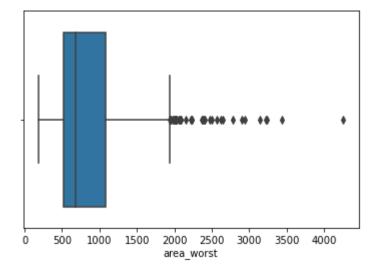
Out[18]: <AxesSubplot:>



In [19]: |sns.boxplot(x["area_worst"])

C:\Users\USER\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarn
ing: Pass the following variable as a keyword arg: x. From version 0.12, the on
ly valid positional argument will be `data`, and passing other arguments withou
t an explicit keyword will result in an error or misinterpretation.
 warnings.warn(

Out[19]: <AxesSubplot:xlabel='area_worst'>

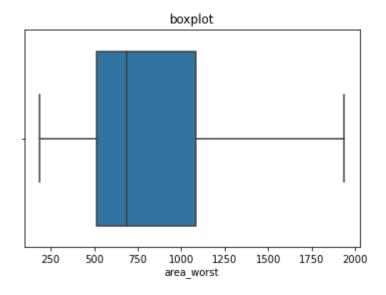


In [20]: from feature_engine.outliers import Winsorizer
 win=Winsorizer(capping_method="iqr", tail="both", fold=1.5, variables=["area_worst"] new_area_worst=win.fit_transform(x[["area_worst"]])

```
In [21]: sns.boxplot(new_area_worst.area_worst)
    plt.title("boxplot")
    plt.show
```

C:\Users\USER\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarn
ing: Pass the following variable as a keyword arg: x. From version 0.12, the on
ly valid positional argument will be `data`, and passing other arguments withou
t an explicit keyword will result in an error or misinterpretation.
warnings.warn(

Out[21]: <function matplotlib.pyplot.show(close=None, block=None)>



```
In [22]: x.insert(loc=23,column="new_area_worst", value=new_area_worst)
```

```
In [23]: del x["area_worst"]
```

In [24]: x

α	1 ') /1	
out	1 44 1	

	radius_mean	texture_mean	perimeter_mean	area_mean	smoothness_mean	compactness_me
0	12.32	12.39	78.85	464.1	0.10280	0.069
1	10.60	18.95	69.28	346.4	0.09688	0.114
2	11.04	16.83	70.92	373.2	0.10770	0.078
3	11.28	13.39	73.00	384.8	0.11640	0.113
4	15.19	13.21	97.65	711.8	0.07963	0.069
564	13.17	18.22	84.28	537.3	0.07466	0.059
565	10.26	14.71	66.20	321.6	0.09882	0.091
566	15.28	22.41	98.92	710.6	0.09057	0.105
567	14.53	13.98	93.86	644.2	0.10990	0.092
568	21.37	15.10	141.30	1386.0	0.10010	0.151

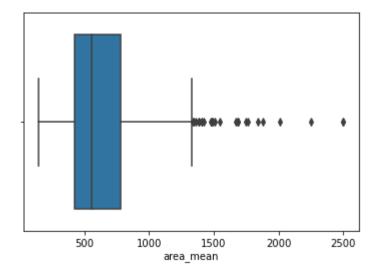
569 rows × 30 columns

4

In [25]: sns.boxplot(x["area_mean"])

C:\Users\USER\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarn
ing: Pass the following variable as a keyword arg: x. From version 0.12, the on
ly valid positional argument will be `data`, and passing other arguments withou
t an explicit keyword will result in an error or misinterpretation.
warnings.warn(

Out[25]: <AxesSubplot:xlabel='area_mean'>

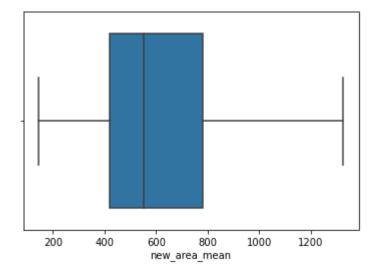


In [28]: from feature_engine.outliers import Winsorizer
 win=Winsorizer(capping_method="iqr", tail="both", fold=1.5, variables=["area_mear
 new_area_mean=win.fit_transform(x[["area_mean"]])
 x.insert(loc=3,column="new_area_mean", value=new_area_mean)

In [29]: sns.boxplot(x["new_area_mean"])

C:\Users\USER\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarn
ing: Pass the following variable as a keyword arg: x. From version 0.12, the on
ly valid positional argument will be `data`, and passing other arguments withou
t an explicit keyword will result in an error or misinterpretation.
 warnings.warn(

Out[29]: <AxesSubplot:xlabel='new_area_mean'>

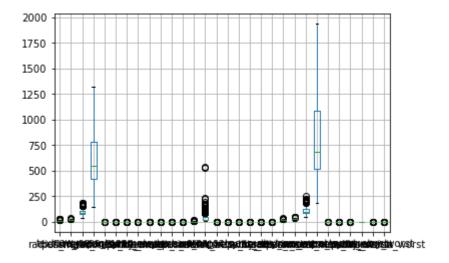


In [30]: del x["area_mean"]
x

	radius_mean	texture_mean	perimeter_mean	new_area_mean	smoothness_mean	compact
0	12.32	12.39	78.85	464.1	0.10280	
1	10.60	18.95	69.28	346.4	0.09688	
2	11.04	16.83	70.92	373.2	0.10770	
3	11.28	13.39	73.00	384.8	0.11640	
4	15.19	13.21	97.65	711.8	0.07963	
		•••				
564	13.17	18.22	84.28	537.3	0.07466	
565	10.26	14.71	66.20	321.6	0.09882	
566	15.28	22.41	98.92	710.6	0.09057	
567	14.53	13.98	93.86	644.2	0.10990	
568	21.37	15.10	141.30	1326.3	0.10010	,
						•

```
In [31]: boxplot=x.boxplot()
boxplot
```

Out[31]: <AxesSubplot:>

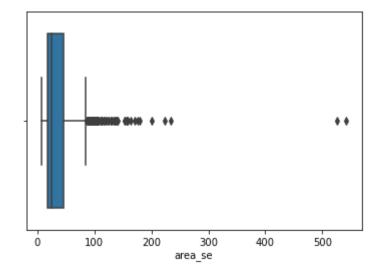


```
In [35]: names=x.columns
names
```

In [38]: sns.boxplot(x["area_se"])

C:\Users\USER\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarn
ing: Pass the following variable as a keyword arg: x. From version 0.12, the on
ly valid positional argument will be `data`, and passing other arguments withou
t an explicit keyword will result in an error or misinterpretation.
 warnings.warn(

Out[38]: <AxesSubplot:xlabel='area_se'>



In [40]: from feature_engine.outliers import Winsorizer
 win=Winsorizer(capping_method="iqr", tail="both", fold=1.5, variables=["area_se"]
 new_area_sc=win.fit_transform(x[["area_se"]])
 x.insert(loc=13,column="new_area_se", value=new_area_sc)

```
In [42]: del x["area se"]
```

in [42].	X	x[urea_se]	ı				
Out[42]:		radius mean	texture mean	perimeter mean	new area mean	smoothness mean	compactness

	radius_mean	texture_mean	perimeter_mean	new_area_mean	smoothness_mean	compactness
0	12.32	12.39	78.85	464.1	0.10280	
1	10.60	18.95	69.28	346.4	0.09688	
2	11.04	16.83	70.92	373.2	0.10770	
3	11.28	13.39	73.00	384.8	0.11640	
4	15.19	13.21	97.65	711.8	0.07963	
564	13.17	18.22	84.28	537.3	0.07466	
565	10.26	14.71	66.20	321.6	0.09882	
566	15.28	22.41	98.92	710.6	0.09057	
567	14.53	13.98	93.86	644.2	0.10990	
568	21.37	15.10	141.30	1326.3	0.10010	

569 rows × 30 columns

```
In [43]: names=x.columns
         names
```

- Out[43]: Index(['radius_mean', 'texture_mean', 'perimeter_mean', 'new_area_mean', 'smoothness_mean', 'compactness_mean', 'concavity_mean', 'points_mean', 'symmetry_mean', 'dimension_mean', 'radius_se', 'texture_se',
 'perimeter_se', 'new_area_se', 'smoothness_se', 'compactness_se',
 'concavity_se', 'points_se', 'symmetry_se', 'dimension_se',
 'radius_worst', 'texture_worst', 'perimeter_worst', 'new_area_worst', 'smoothness_worst', 'compactness_worst', 'concavity_worst', 'points_worst', 'symmetry_worst', 'dimension_worst'], dtype='object')
- In [44]: from sklearn.model selection import train test split x_train, x_test, y_train, y_test=train_test_split(x,y, train_size=0.80)
- In [46]: | from sklearn.preprocessing import StandardScaler scale = StandardScaler() x train = scale.fit transform(x train) x test = scale.transform(x test)
- In [47]: | from sklearn.neighbors import KNeighborsClassifier clf=KNeighborsClassifier(n neighbors=3)

```
In [49]: clf.fit(x train, y train)
        C:\Users\USER\anaconda3\lib\site-packages\sklearn\neighbors\_classification.py:
        179: DataConversionWarning: A column-vector y was passed when a 1d array was ex
        pected. Please change the shape of y to (n samples,), for example using ravel
        ().
          return self._fit(X, y)
Out[49]: KNeighborsClassifier(n_neighbors=3)
In [50]: y_pred=clf.predict(x_test)
        y pred
                                     'B', 'B', 'B', 'B',
Out[50]: array(['M', 'B', 'B', 'M', 'B',
                                                       'B', 'B',
                                                       'Μ',
               'B', 'B', 'M', 'B', 'B', 'B', 'B', 'M', 'B',
                                                            'B', 'M', 'B',
                                                            'B',
                                                       'B',
                                                                'M',
               'B', 'M', 'M', 'B', 'M',
                                     'B', 'M',
                                              'B', 'B',
               'M', 'M', 'M', 'B',
                                     'B', 'M',
                                              'B', 'B',
                                                                'B',
                                                       'B', 'B',
               'B', 'M', 'M', 'B', 'B', 'M', 'M',
                                              'B', 'M',
                                                       'M', 'M', 'B', 'B',
                                     'B', 'B', 'B', 'B', 'B', 'M', 'M',
               'M', 'M', 'B', 'B', 'B',
               In [51]: from sklearn.metrics import confusion_matrix, accuracy_score
        print(confusion matrix(y test, y pred))
        [[75 0]
         [ 4 35]]
In [53]: | accuracy_score(y_test, y_pred)
Out[53]: 0.9649122807017544
In [ ]:
```