```
In [29]: import pandas as pd
                             from sklearn.datasets import load_breast_cancer
                             from sklearn.model selection import train test split
                             from sklearn.naive_bayes import GaussianNB
                             from sklearn.metrics import confusion matrix
                             from matplotlib import pyplot as plt
                             import seaborn as sns
                             from sklearn import metrics
                             from sklearn.metrics import classification report
                             cancer=load_breast_cancer()
                             cancerdf=pd. DataFrame (cancer. data, columns=cancer. feature names)
                             print(cancerdf.head()) # head()默认显示前5行数据
                                    mean ra<sup>di</sup>us mean texture mean per<sup>i</sup>meter mean area mean smoot<sup>h</sup>ness
                                       17 99 10 38 122 80 1001 0 0 11840
20 57 17 77 132 90 1326 0 0 08474
19 69 21 25 130 00 1203 0 0 10960
11 42 20 38 77 58 386 1 0 14250
20 29 14 34 135 10 1297 0 0 10030
                                   mean compac<sup>t</sup>ness mean concav<sup>1</sup>ty mean concave po<sup>1</sup>n<sup>t</sup>s mean symme<sup>t</sup>ry
                                            O 27760 O 3001 O 14710 O 2419
O 07864 O 0869 O 07017 O 1812
O 15990 O 1974 O 12790 O 2069
O 28390 O 2414 O 10520 O 2597
O 13280 O 1980 O 10430 O 1809
                                    mean <sup>f</sup>rac<sup>t</sup>a<sup>1</sup> <sup>di</sup>mens<sup>i</sup>on wors<sup>t</sup> ra<sup>di</sup>us wors<sup>t</sup> <sup>t</sup>ex<sup>t</sup>ure wors<sup>t</sup> per<sup>i</sup>me<sup>t</sup>er

    Permitted of the control of the co
                            2
                                    worst area worst smoothness worst compactness worst concavity
                                      2019 0 0 1622 0 6656 0 7119
1956 0 0 1238 0 1866 0 2416
1700 0 0 1444 0 4245 0 4504
567 7 0 2098 0 8663 0 6869
1575 0 0 1374 0 2050 0 4000
                            2
                                    worst concave points worst symmetry worst fractal dimension
                                               0 11890
                                                                                                                                                                                             0 08902
                                                                                                                                                                                             0 17300
                                                                                                                                                                                             0 07678
                            Es rows x 30 columns
```

```
In [30]: print("肿瘤的分类: ", cancer['target_names']) print("肿瘤的分类: ", cancer['feature_names'])
```

```
肿瘤的分类。「'malignant' 'bonign']
          肿瘤的分类 ['mean ra<sup>dlus' mean text</sup>ure' mean per<sup>i</sup>me<sup>t</sup>er' mean area'
            mean smoothness mean compactness mean concav<sup>ity</sup>
           mean concave points mean symmetry mean fractal dimension
           ra<sup>di</sup>us error texture error per<sup>i</sup>meter error area error
           smoothness error compactness error concavity error
           concave points error symmetry error fractal dimension error
           worst radius worst texture worst perimeter worst area
            worst smoothness worst compactness worst concavity
           worst concave points worst symmetry worst fractal dimension
In [31]: x, y=cancer. data, cancer. target
          x_train, x_test, y_train, y_test=train_test_split(x, y, test_size=0.2, random_state = 100)
          print(x train. shape)# 查看训练集数据形态
          print(x_test. shape)# 查看测试集数据形态
          (455, 30)
          (114, 30)
In [32]:
          clf=GaussianNB()
          clf. fit(x_train, y_train) #对训练集进行拟合
          Y_pred= clf. predict(x_test)
          print("Accuracy:", metrics. accuracy_score(y_test, Y_pred))
          print("Precision:", metrics. precision score(y test, Y pred))
          print("Recall:", metrics. recall_score(y_test, Y_pred))
          A<sub>ccuracy:</sub> 0 0208245614035088
          P<sub>rec</sub>i<sub>s</sub>i<sub>on:</sub> 0 9253731343283582
          R<sub>eca</sub>11: 0 9538461538461539
In [33]: | pred=clf. predict(x_test)
          cm=confusion_matrix(pred, y_test)
          sns. heatmap(cm, cmap=sns. color_palette("Blues"), annot=True, fmt='d')
          plt. xlabel ('real')
          plt. ylabel('predict')
          plt. show()
                                                                                      60
                                                                                     50
                               44
                                                              3
              0
                                                                                     40
                                                                                    - 30
                                                                                    - 20
                                5
                                                             62
              ٦ -
                                                                                    - 10
                                0
                                                              1
                                              real
```

```
In [108... import numpy as np
           import pandas as pd
           from sklearn. decomposition import PCA
           from sklearn.model_selection import train_test_split
           from sklearn.linear model import LogisticRegression
           from sklearn import metrics
           from sklearn.metrics import classification_report
           from sklearn.datasets import load_breast_cancer
           from matplotlib import pyplot as plt
           import seaborn as sns
           from sklearn.metrics import confusion_matrix
           breast = load_breast_cancer()
           breast_data = breast.data
           breast input = pd. DataFrame(breast data)
           breast_labels = breast.target
           labels = np. reshape(breast_labels, (569, 1))
           final_breast_data = np. concatenate([breast_data, labels], axis=1)
           breast_dataset = pd. DataFrame(final_breast_data)
           features = breast.feature_names
           features
           array (l' mean radius , 'mean texture', 'mean perimeter', 'mean area',
Out[108]:
                  mean smoothness, mean compactness, mean concav<sup>1t</sup>y, mean concave points, mean symmetry, mean fractal dimension,
                  radius error, 'texture error', 'perimeter error', 'area error',
                  smoothness error . compactness error . concavity error .
                  concave points error, symmetry error,
                  'fractal dimension error', worst radius, worst texture',
                  worst perimeter, worst area, worst smoothness,
                  worst compactness, worst concavity, worst concave points,
                  worst symmetry, worst fractal dimension dtype='<u23')
In [109... | features_labels = np. append(features, 'label')
           breast_dataset.columns = features_labels
           breast_dataset['label']. replace('Benign', 0, inplace=True)
           breast_dataset['label']. replace('Malignant', 1, inplace=True)
           breast_dataset. tail()
Out[109]:
```

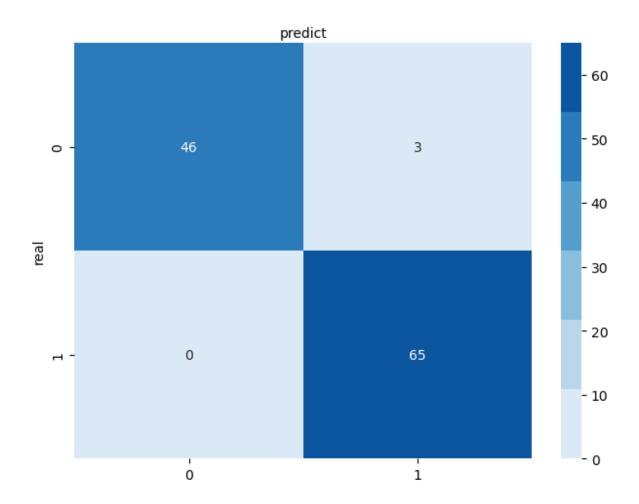
	mean radius	mean texture	mean perimeter	mean area	mean smoothness	mean compactness	mean concavity	mean concave points	mea symmetr
564	21.56	22.39	142.00	1479.0	0.11100	0.11590	0.24390	0.13890	0.172
565	20.13	28.25	131.20	1261.0	0.09780	0.10340	0.14400	0.09791	0.175
566	16.60	28.08	108.30	858.1	0.08455	0.10230	0.09251	0.05302	0.159
567	20.60	29.33	140.10	1265.0	0.11780	0.27700	0.35140	0.15200	0.239
568	7.76	24.54	47.92	181.0	0.05263	0.04362	0.00000	0.00000	0.158

5 rows × 31 columns

```
In [110... from sklearn.preprocessing import StandardScaler
          x = breast dataset. loc[:, features]. values
          x = StandardScaler().fit_transform(x)
```

```
In [111... k=9 #设置降维的占比
           pca= PCA(n_components=k)#调用PCA函数, 先实例化
           pcaCom = pca. fit_transform(x)
           pcaCom = pd. DataFrame(pcaCom)
           print("主成分的数量: ", pca. n_components_)
           X = pcaCom.iloc[:, [0, 1, 2, 3, 4, 5, 6, 7, 8]].values
           \#X = pcaCom.iloc[:, [0, 1, 2, 3, 4, 5, 6, 7, 8]].values
           Y = breast_dataset.iloc[:, 30].values
           X_train, X_test, Y_train, Y_test = train_test_split(X, Y, train_size=0.8, test_size=
           主成分的数量: 9
In [112... | model=LogisticRegression()
           model. fit(X_train, Y_train)
           Y_pred= model.predict(X_test)
           Y pred[0:9]
           array CEO I O I I I O O I I
Out[112]:
           print("Accuracy:", metrics. accuracy_score(Y_test, Y_pred))
In [113...
           print("Precision:", metrics. precision_score(Y_test, Y_pred))
           print("Recall:", metrics. recall_score(Y_test, Y_pred))
           A<sub>CCUPACY</sub>: 0 9730842105263158
           P<sub>ree</sub>i<sub>s</sub>i<sub>on:</sub> 0 9558823529411765
           Reca<sup>11</sup>: 1 0
In [114... | class_names = [0, 1]
           fig, ax = plt. subplots()
           tick_marks = np. arange(len(class_names))
           plt. xticks(tick_marks, class_names)
           plt. yticks(tick_marks, class_names)
           # create heatmap
           cm = confusion_matrix(Y_test, Y_pred)
           sns. heatmap(pd. DataFrame(cm), cmap=sns. color_palette("Blues"), annot=True, fmt='d')
           ax. xaxis. set_label_position("top")
           plt. tight_layout()
           plt. title ('Confusion matrix', y=1.1)
           plt. ylabel ('real')
           plt. xlabel('predict')
           plt. show()
```

## Confusion matrix



```
In [45]: import numpy as np
          import pandas as pd
          from sklearn. decomposition import PCA
          from sklearn.model_selection import train_test_split
          from sklearn.linear model import LogisticRegression
          from sklearn import metrics
          from sklearn.metrics import classification_report
          from sklearn.datasets import load_breast_cancer
          from matplotlib import pyplot as plt
          import seaborn as sns
          from sklearn.metrics import confusion matrix
          from sklearn.naive bayes import GaussianNB
          breast = load breast cancer()
          breast data = breast.data
          breast input = pd. DataFrame(breast data)
          breast_labels = breast.target
          labels = np. reshape(breast_labels, (569, 1))
          final_breast_data = np. concatenate([breast_data, labels], axis=1)
          breast_dataset = pd. DataFrame(final_breast_data)
          features = breast. feature names
          features
          array CC' mean radius, 'mean texture', 'mean perimeter', 'mean area',
Out[45]:
                 mean smoothness, mean compactness, mean concavity,
                 mean concave points, mean symmetry, mean fractal dimension,
                 radius error texture error perimeter error area error
                 smoothness error eompactness error concavity error
                 concave po<sup>i</sup>n<sup>t</sup>s error symme<sup>t</sup>ry error '
'r<sub>ractal</sub> dimension error worst radius' worst texture'
                 worst perimeter, worst area, worst smoothness,
                 worst compactness, worst concavity, worst concave points,
                 worst symmetry worst fractal dimension dtype='<u23')
         features_labels = np. append(features, 'label')
In [46]:
          breast_dataset.columns = features_labels
          breast_dataset['label']. replace('Benign', 0, inplace=True)
          breast_dataset['label']. replace('Malignant', 1, inplace=True)
          breast_dataset.tail()
                                                                                   ....
Out[46]:
```

		mean radius	mean texture	mean perimeter	mean area	mean smoothness	mean compactness	mean concavity	concave points	mear symmetr
	564	21.56	22.39	142.00	1479.0	0.11100	0.11590	0.24390	0.13890	0.172
	565	20.13	28.25	131.20	1261.0	0.09780	0.10340	0.14400	0.09791	0.175
	566	16.60	28.08	108.30	858.1	0.08455	0.10230	0.09251	0.05302	0.159
	567	20.60	29.33	140.10	1265.0	0.11780	0.27700	0.35140	0.15200	0.239
	568	7.76	24.54	47.92	181.0	0.05263	0.04362	0.00000	0.00000	0.158

5 rows × 31 columns

```
In [47]: from sklearn.preprocessing import StandardScaler

x = breast_dataset.loc[:, features].values
x = StandardScaler().fit_transform(x)
```

```
#设置降维的占比
In [48]:
          k=5
          pca= PCA(n_components=k)#调用PCA函数, 先实例化
          pcaCom = pca. fit_transform(x)
          pcaCom = pd. DataFrame(pcaCom)
          print("主成分的数量: ", pca. n_components_)
          X = pcaCom.iloc[:, [0, 1, 2, 3, 4]].values
          #X = pcaCom.iloc[:, [0, 1, 2, 3, 4, 5, 6, 7, 8]].values
          Y = breast_dataset.iloc[:, 30].values
          X_train, X_test, Y_train, Y_test = train_test_split(X, Y, train_size=0.8, test_size=
          主成分的数量: 5
In [49]: | model = GaussianNB()
          model.fit(X_train, Y_train)
          Y_pred= model.predict(X_test)
          Y_pred[0:9]
          array ([0 1 0 1 1 1 0 0 1 ])
Out[49]:
In [50]: print("Accuracy:", metrics. accuracy_score(Y_test, Y_pred))
          print("Precision:", metrics. precision_score(Y_test, Y_pred))
          print("Recall:", metrics. recall_score(Y_test, Y_pred))
          A<sub>ccuracy</sub>: 0 9210526315789473
          Precision: 0 9
          R<sub>eca</sub>11: 0 9692307692307692
In [51]: class_names=[0,1]
          fig, ax = plt. subplots()
          tick_marks = np. arange(len(class_names))
          plt. xticks(tick_marks, class_names)
          plt. yticks(tick_marks, class_names)
          # create heatmap
          cm = confusion_matrix(Y_test, Y_pred)
          sns. heatmap(pd. DataFrame(cm), cmap=sns. color_palette("Blues"), annot=True, fmt='d')
          ax. xaxis. set_label_position("top")
          plt. tight_layout()
          plt. title('Confusion matrix', y=1.1)
          plt. ylabel('real')
          plt. xlabel('predict')
          plt. show()
```

## Confusion matrix

