```
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

