```
In [24]: from sklearn.linear_model import LogisticRegression
    from sklearn.model_selection import train_test_split
    from sklearn.metrics import accuracy_score
    import matplotlib.pyplot as plt
    import numpy as np
    import pandas as pd
    from sklearn.preprocessing import StandardScaler
```

In [5]: url = "https://archive.ics.uci.edu/ml/machine-learning-databases/wine-quality/wined
 data = pd.read_csv(url, sep=';')
 data

Out[5]:

•		fixed acidity	volatile acidity	citric acid	residual sugar	chlorides	free sulfur dioxide	total sulfur dioxide	density	рН	sulphates	alcoł
	0	7.0	0.27	0.36	20.7	0.045	45.0	170.0	1.00100	3.00	0.45	
	1	6.3	0.30	0.34	1.6	0.049	14.0	132.0	0.99400	3.30	0.49	
	2	8.1	0.28	0.40	6.9	0.050	30.0	97.0	0.99510	3.26	0.44	1
	3	7.2	0.23	0.32	8.5	0.058	47.0	186.0	0.99560	3.19	0.40	1
	4	7.2	0.23	0.32	8.5	0.058	47.0	186.0	0.99560	3.19	0.40	
	•••											
	4893	6.2	0.21	0.29	1.6	0.039	24.0	92.0	0.99114	3.27	0.50	1
	4894	6.6	0.32	0.36	8.0	0.047	57.0	168.0	0.99490	3.15	0.46	
	4895	6.5	0.24	0.19	1.2	0.041	30.0	111.0	0.99254	2.99	0.46	!
	4896	5.5	0.29	0.30	1.1	0.022	20.0	110.0	0.98869	3.34	0.38	1
	4897	6.0	0.21	0.38	0.8	0.020	22.0	98.0	0.98941	3.26	0.32	1

4898 rows × 12 columns

```
In [6]: # Load the winequality-white dataset
    # This example assumes that the dataset has already been preprocessed
    # and is available as a pandas dataframe called "data"

# Define the features and target variables
    X = data.drop('quality', axis=1)
    y = data['quality']

# Split the data into training and test sets
    X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state)
In [8]: # Create the logistic regression model
    model = LogisticRegression()
```

```
In [8]: # Create the logistic regression model
model = LogisticRegression()

# Train the model on the training data
model.fit(X_train, y_train)

# Predict the target variable for the test data
y_pred = model.predict(X_test)

# Estimate the class probabilities for the test data
probas = model.predict_proba(X_test)
```

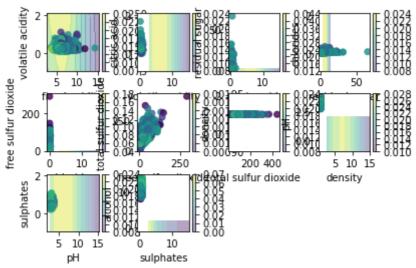
```
# Print the accuracy score for the model
         accuracy = accuracy_score(y_test, y_pred)
         print("Accuracy: ", accuracy)
         Accuracy: 0.45918367346938777
         C:\Users\maham\anaconda3\lib\site-packages\sklearn\linear_model\_logistic.py:814:
         ConvergenceWarning: lbfgs failed to converge (status=1):
         STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
         Increase the number of iterations (max_iter) or scale the data as shown in:
             https://scikit-learn.org/stable/modules/preprocessing.html
         Please also refer to the documentation for alternative solver options:
             https://scikit-learn.org/stable/modules/linear_model.html#logistic-regression
           n_iter_i = _check_optimize_result(
In [17]: # Create a meshgrid for all 11 features
         x_{min}, x_{max} = X.iloc[:, 0].min() - 1, <math>X.iloc[:, 0].max() + 1
         y_{min}, y_{max} = X.iloc[:, 1].min() - 1, <math>X.iloc[:, 1].max() + 1
         xx, yy = np.meshgrid(np.arange(x_min, x_max, 0.1),
                               np.arange(y_min, y_max, 0.1))
         for i in range(2, 12):
              plt.subplot(3, 4, i-1)
              plt.subplots_adjust(wspace=0.4, hspace=0.4)
             X_plot = np.zeros((xx.shape[0] * xx.shape[1], 11))
             X_plot[:, 0] = xx.ravel()
             X_plot[:, 1] = yy.ravel()
             for j in range(2, 12):
                  if i != j:
                      X_{plot}[:, j-1] = X[X.columns[j-1]].mean()
             Z = model.predict_proba(X_plot)[:, 1]
             Z = Z.reshape(xx.shape)
              # Plot the decision boundaries as a contour plot
              plt.contourf(xx, yy, Z, alpha=0.4)
             plt.colorbar()
             # Plot the data points on top of the decision boundaries
              plt.scatter(X.iloc[:, i-2], X.iloc[:, i-1], c=y, alpha=0.8)
             plt.xlabel(X.columns[i-2])
             plt.ylabel(X.columns[i-1])
```

plt.show()

C:\Users\maham\anaconda3\lib\site-packages\sklearn\base.py:450: UserWarning: X doe s not have valid feature names, but LogisticRegression was fitted with feature nam warnings.warn(C:\Users\maham\anaconda3\lib\site-packages\sklearn\base.py:450: UserWarning: X doe s not have valid feature names, but LogisticRegression was fitted with feature nam warnings.warn(C:\Users\maham\anaconda3\lib\site-packages\sklearn\base.py:450: UserWarning: X doe s not have valid feature names, but LogisticRegression was fitted with feature nam es warnings.warn(C:\Users\maham\anaconda3\lib\site-packages\sklearn\base.py:450: UserWarning: X doe s not have valid feature names, but LogisticRegression was fitted with feature nam es warnings.warn(C:\Users\maham\anaconda3\lib\site-packages\sklearn\base.py:450: UserWarning: X doe s not have valid feature names, but LogisticRegression was fitted with feature nam warnings.warn(C:\Users\maham\anaconda3\lib\site-packages\sklearn\base.py:450: UserWarning: X doe s not have valid feature names, but LogisticRegression was fitted with feature nam warnings.warn(C:\Users\maham\anaconda3\lib\site-packages\sklearn\base.py:450: UserWarning: X doe s not have valid feature names, but LogisticRegression was fitted with feature nam es warnings.warn(C:\Users\maham\anaconda3\lib\site-packages\sklearn\base.py:450: UserWarning: X doe s not have valid feature names, but LogisticRegression was fitted with feature nam es warnings.warn(C:\Users\maham\anaconda3\lib\site-packages\sklearn\base.py:450: UserWarning: X doe s not have valid feature names, but LogisticRegression was fitted with feature nam es warnings.warn(

C:\Users\maham\anaconda3\lib\site-packages\sklearn\base.py:450: UserWarning: X doe
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es

warnings.warn(



In []:

In Γ 1:

In []: