PS5

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```
In [1]: import pandas as pd
        import numpy as np
        import matplotlib.pyplot as plt
   Q1
1
In [2]: import sklearn
        from sklearn import preprocessing
        from sklearn.linear_model import LogisticRegression
        from sklearn.cross_validation import train_test_split
        from sklearn import metrics
        from sklearn.metrics import classification_report, confusion_matrix
        from sklearn.model_selection import LeaveOneOut
        import seaborn as sb
        from sklearn.model_selection import KFold
        import statsmodels.api as sm
        from scipy.interpolate import LSQUnivariateSpline
/anaconda/lib/python3.6/site-packages/sklearn/cross_validation.py:41: DeprecationWarning: This m
  "This module will be removed in 0.20.", DeprecationWarning)
/anaconda/lib/python3.6/site-packages/statsmodels/compat/pandas.py:56: FutureWarning: The pandas
  from pandas.core import datetools
In [3]: sd = pd.read_csv("data/strongdrink.txt")
        xvar = ["alco", "malic", "tot_phen", "color_int"]
        yvar = ["cultivar"]
        X_train, X_test, y_train, y_test = \
            train_test_split(sd[xvar], sd[yvar], test_size = 0.25,
               random_state=20)
        MultLogReg = LogisticRegression(multi_class='multinomial',
                                        solver='newton-cg')
        MultLogReg.fit(X_train, y_train)
        coef = pd.DataFrame(MultLogReg.coef_[0:2], index = ["cultivar = 1", "cultivar = 2"], col
        coef.loc[:,"Intercept"] = MultLogReg.intercept_[:2]
        coef
```

```
/anaconda/lib/python3.6/site-packages/sklearn/utils/validation.py:578: DataConversionWarning: A
 y = column_or_1d(y, warn=True)
Out [3]:
                                   malic tot_phen color_int Intercept
                          alco
        cultivar = 1 1.700368 -0.265602 1.223892
                                                     0.022762 -24.010548
        cultivar = 2 -1.468065 -0.333055 0.664017 -0.922714 22.802713
In [4]: pd.crosstab(index = y_test["cultivar"], columns = 'count')
Out[4]: col_0
                  count
        cultivar
        1
                     13
        2
                     21
                     10
In [5]: print(classification_report(y_test, MultLogReg.predict(X_test)))
                          recall f1-score
             precision
                                             support
                  0.87
                            1.00
                                      0.93
          1
                                                  13
          2
                  1.00
                            0.90
                                      0.95
                                                  21
          3
                  1.00
                            1.00
                                      1.00
                                                  10
avg / total
                  0.96
                            0.95
                                      0.96
                                                  44
```

Category 3 is best at predicting, which, on the contrary to the question, has fewest observations.

1.1 b) LOOCV

/anaconda/lib/python3.6/site-packages/sklearn/utils/validation.py:578: DataConversionWarning: A y = column_or_1d(y, warn=True)

```
In [7]: print(classification_report(classify[:,0], classify[:,1]))
```

	precision	recall	f1-score	support	
1.0	0.90	0.93	0.92	59	
2.0	0.91	0.90	0.91	71	
3.0	0.96	0.93	0.95	46	
avg / total	0.92	0.92	0.92	176	

The precision rates for each category is in the first column. The precision is worse than that from part (a).

1.2 c) k-fold cross validation

1.0

0.87

0.93

```
In [9]: k = 4
        kf = KFold(n_splits=k, random_state=10, shuffle=True)
        kf.get_n_splits(Xvars)
        classify_kf = np.empty((N, 2))
        k_ind = 0
        for train_index, test_index in kf.split(Xvars):
            X_train, X_test = Xvars[train_index], Xvars[test_index]
            y_train, y_test = yvals[train_index], yvals[test_index]
            LogReg = LogisticRegression(multi_class='multinomial',
                                        solver='newton-cg')
            LogReg.fit(X_train, y_train)
            y_pred = LogReg.predict(X_test)
            classify_kf[test_index,0:1] = y_test
            classify_kf[test_index,1] = y_pred
/anaconda/lib/python3.6/site-packages/sklearn/utils/validation.py:578: DataConversionWarning: A
 y = column_or_1d(y, warn=True)
In [10]: print(classification_report(classify_kf[:,0], classify_kf[:,1]))
             precision recall f1-score
                                             support
```

0.90

59

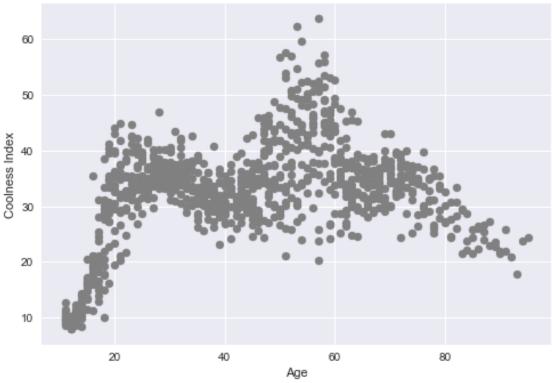
```
2.0 0.91 0.87 0.89 71
3.0 0.96 0.93 0.95 46
avg / total 0.91 0.91 0.91 176
```

The precision rates for each category is in the first column. The precision is worse than that from part (a) and part (b).

2 Q2

2.1 Scatterplot

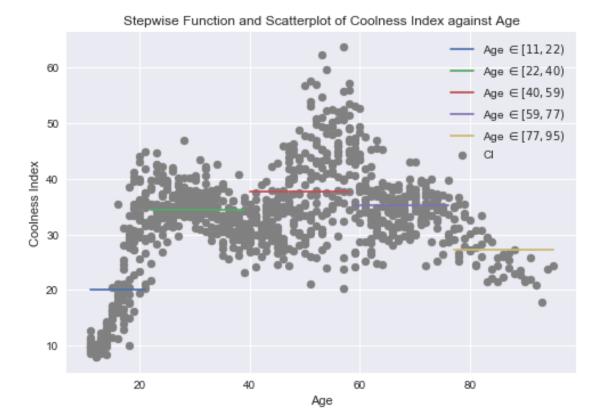




2.2 Stepwise Function

```
In [13]: cutoff = np.array([21, 39, 58, 76])
         ci.loc[:,"binnum"] = np.searchsorted(cutoff, ci.age)
         ci = pd.concat((ci,
             pd.get_dummies(ci.binnum, prefix = "bin")),
             axis = 1)
In [14]: reg = sm.OLS(ci.CI, ci.iloc[:,-5:]).fit()
In [15]: bins = [
             "$\in [11, 22)$",
             "$\in [22, 40)$",
             "$\in [40, 59)$",
             "$\in [59, 77)$",
             "$\in [77, 95)$"
         ]
         for i in range(5):
             ax.plot(ci.loc[ci.binnum == i, "age"], np.ones((ci.binnum == i).sum()) * reg.params
                     label = 'Age ' + bins[i])
         ax.set_title("Stepwise Function and Scatterplot of Coolness Index against Age")
         ax.legend()
         fig
```

Out[15]:



In [16]: print("beta 0 to beta 5 are as follows:")
 reg.params

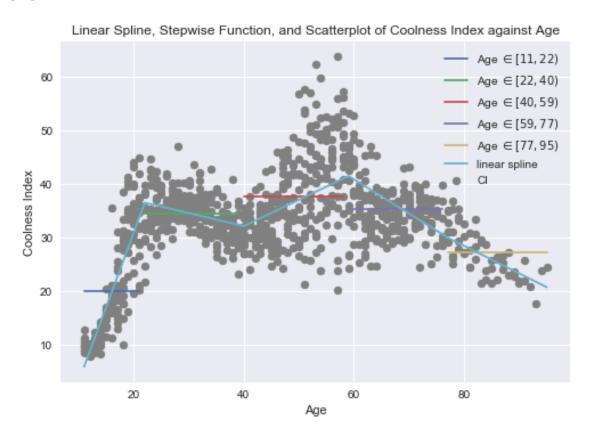
beta 0 to beta 5 are as follows:

Out[16]: bin_0 20.102457 bin_1 34.475788 bin_2 37.635105 bin_3 35.225400 bin_4 27.296378 dtype: float64

Predicted coolness of a 73-year old is

Out[17]: 35.225400040242754

Out[18]:

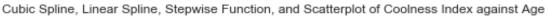


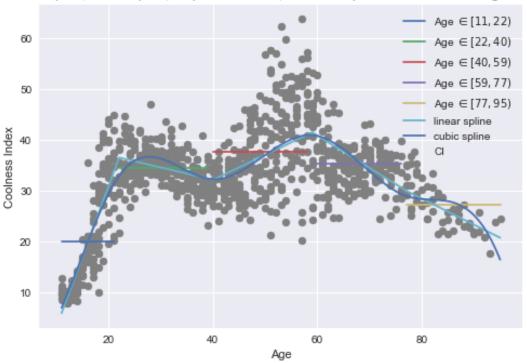
In [19]: print("Predicted coolness of a 73-year old from linear spline is")
 spl_ci(73)

Predicted coolness of a 73-year old from linear spline is

ax.legend()
ax.set_title("Cubic Spline, Linear Spline, Stepwise Function, and Scatterplot of Coolne
fig

Out[20]:





Predicted coolness of a 73-year old from cubic spline is

Out[21]: array(31.262429389257864)