

Discriminant Analysis

Introduction

Perform Discriminant analysis using linear discriminant analysis from sklearn.discriminant_analysis.

The Data

The data set contain details of number of visits (1 or 2) to resort by the customer with some features.

```
import pandas as pd
df= pd.read_csv(r"C:\Users\my pc\Desktop\MBA - BA II\Multivariate analysis lab\4.DA\DAdata.csv")
df.columns = df.columns.str.replace(" ", "_")
df=df.rename(columns = {'Annual_family_income_(000s)':'Annual_family_income'})
#Dropping unnecessary columns
df.drop(['Respondent_Number'],axis = 1, inplace=True)
df.info()

Data columns (total 7 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   Resort_visit                          30 non-null    int64
1   Annual_family_income                  30 non-null    float64
2   Attitude_towards_travel              30 non-null    int64
3   Importance_attached_to_family_skiing_holiday  30 non-null    int64
4   Household_size                        30 non-null    int64
5   Age_of_head_of_household             30 non-null    int64
6   Amount_spent_on_family_skiing        30 non-null    int64

#split the feature and target variable
x = df.drop(['Resort_visit'],axis = 1)
x.info()
y = df['Resort_visit']
```

Summary statistics and visualization of dataset

Group Frequency :

```
#group frequency
count = df.groupby(['Resort_visit']).size()
print(count)
```

output

```
Resort_visit
1           15
2           15
```

Here we observed that, we have equal number of data on both class.

Group mean

```
#group mean
class_feature_means = pd.DataFrame(columns=y)
for c, rows in df.groupby('Resort_visit'):
    class_feature_means[c] = rows.mean()
class_feature_means = class_feature_means.drop('Resort_visit')
class_feature_means
```

output:

Resort_visit	1	2
Annual_family_income	60.520000	41.913333
Attitude_towards_travel	5.400000	4.333333
Importance_attached_to_family_skiing_holiday	5.800000	4.066667
Household_size	4.333333	2.800000
Age_of_head_of_household	53.733333	50.133333
Amount_spent_on_family_skiing	2.600000	1.400000

These are mean value of all feature class 1 and class 2

Perform one-way MANOVA

We conduct this monova analysis to our find data is statistically significant to perform to lda.

```
from statsmodels.multivariate.manova import MANOVA
fit = MANOVA.from_formula('Annual_family_income + Attitude_towards_travel + \
    Importance_attached_to_family_skiing_holiday + \
    Household_size + \
    Age_of_head_of_household + \
    Amount_spent_on_family_skiing ~ Resort_visit', data=df)
print(fit.mv_test())
```

output:

Multivariate linear model

=====					
Intercept	Value	Num DF	Den DF	F Value	Pr > F
Wilks' lambda	0.0525	6.0000	23.0000	69.1554	0.0000
Pillai's trace	0.9475	6.0000	23.0000	69.1554	0.0000
Hotelling-Lawley trace	18.0405	6.0000	23.0000	69.1554	0.0000
Roy's greatest root	18.0405	6.0000	23.0000	69.1554	0.0000
=====					
Resort_visit	Value	Num DF	Den DF	F Value	Pr > F
Wilks' lambda	0.3021	6.0000	23.0000	8.8556	0.0000
Pillai's trace	0.6979	6.0000	23.0000	8.8556	0.0000
Hotelling-Lawley trace	2.3102	6.0000	23.0000	8.8556	0.0000
Roy's greatest root	2.3102	6.0000	23.0000	8.8556	0.0000
=====					

The Wilks' lambda test statistics is statistically significant [Wilks' lambda = 0.3021, $F(6, 23) = 8.8556$, $p = 0.000$] and indicates that resort visit has a statistically significant association with all the features.

Linear Discriminant Analysis

Here we will perform the linear discriminant analysis (LDA) using sklearn to see the differences between each group. LDA will discriminate the groups using information from both the dependent variables.

```
#LinearDiscriminantAnalysis
from sklearn.discriminant_analysis import LinearDiscriminantAnalysis
import seaborn as sns
import numpy as np
lda = LinearDiscriminantAnalysis(n_components = 1)
da = lda.fit(x,y)
y_pred = lda.predict(x)
print(y_pred)
```

output:

```
[1 1 1 2 1 1 2 1 1 1 1 1 1 1 1 1 2 2 2 2 2 2 2 2 2 2 2 2 2 2]
```

```
# get Prior probabilities of groups:
```

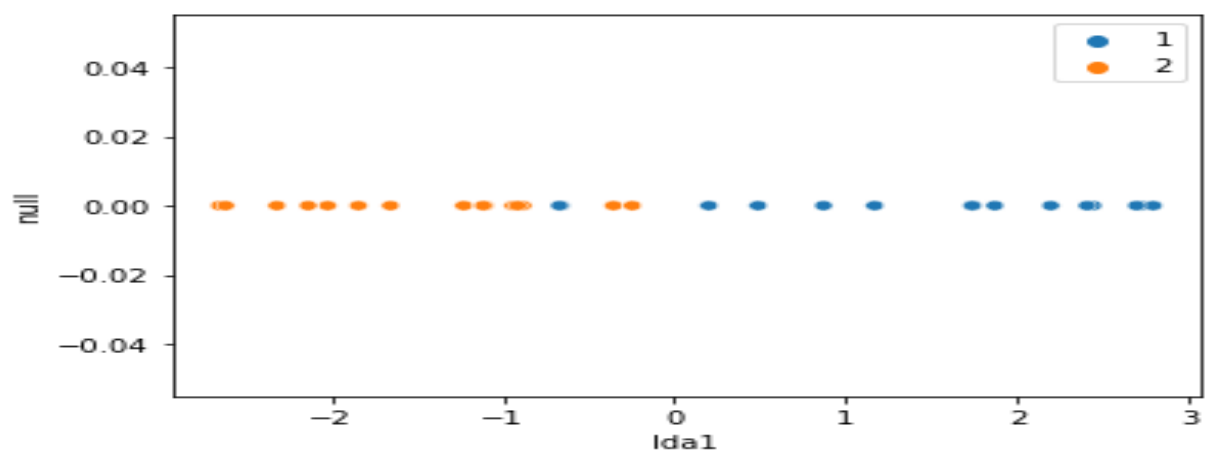
```
da.priors_
```

output:

```
array([0.5, 0.5])
```

Plot

```
#plot
X_new = pd.DataFrame(da.transform(x), columns=["lda1"])
val = 0. # this is the value where you want the data to appear on the y-axis.
X_new['null'] = np.zeros_like(X_new) + val;
sns.scatterplot(data=X_new, x="lda1", y="null", hue=df.Resort_visit.tolist(), palette=["C0", "C1"])
```

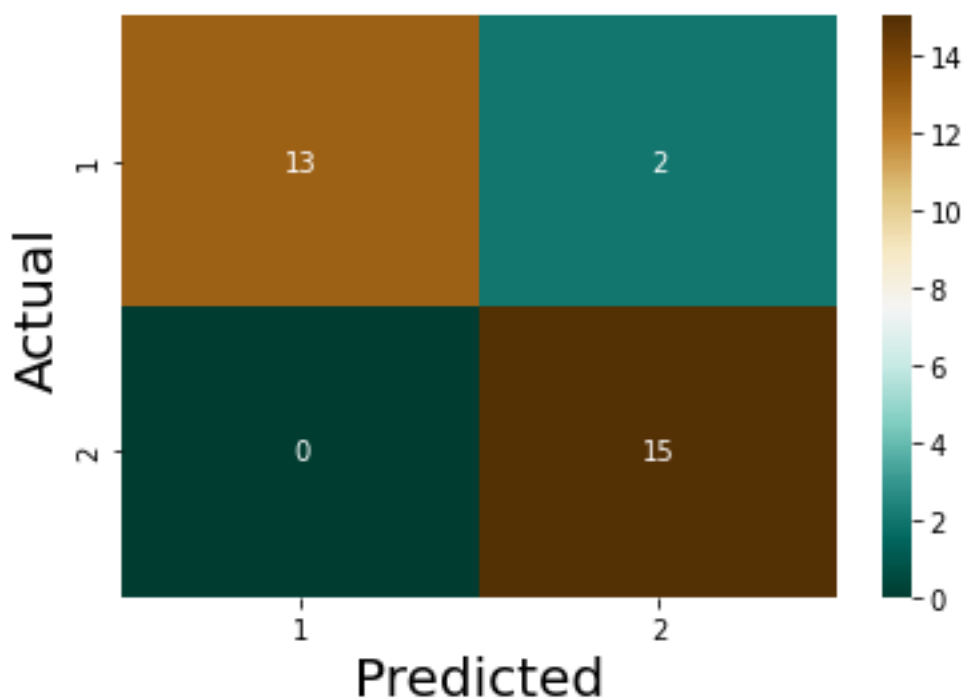


Confusion Matrix

```
from sklearn import metrics
cm=metrics.confusion_matrix(y,y_pred)
cm
x_axis = [1,2]
y_axis = [1,2]
p=sns.heatmap(cm, annot=True, cmap='BrBG_r',xticklabels=x_axis,yticklabels=y_axis)
p.set_xlabel("Predicted", fontsize = 20)
p.set_ylabel("Actual", fontsize = 20)
```

output:

```
array([[13,  2],
       [ 0, 15]])
```



Using test data set predict the number of visit

#test dataset

```
df_test= pd.read_csv(r"C:\Users\my pc\Desktop\MBA - BA II\Multivariate analysis
lab\4.DA\DAdata_test.csv")
df_test.columns = df_test.columns.str.replace(" ", "_")
df_test=df_test.rename(columns = {'Annual_family_income_(000s)':'Annual_family_income'})
df_test.drop(['Respondent_Number'],axis = 1, inplace=True)
```

```
test_pred = lda.predict(df_test)
test_pred
```

output :

```
array([1, 2, 2, 2, 1, 1, 2, 1, 1, 2])
```

Conclusion

We done the Discriminant analysis on given data set and predict the number of visit of resort using Linear discriminant analysis model.

Python code

```
# -*- coding: utf-8 -*-
"""
Created on Sun Jan 30 09:13:07 2022

@author: harigaran
"""

import pandas as pd
df= pd.read_csv(r"C:\Users\my pc\Desktop\MBA - BA II\Multivariate analysis lab\4.DA\DAdata.csv")
df.columns = df.columns.str.replace(" ", "_")
df=df.rename(columns = {'Annual_family_income_(000s)':'Annual_family_income'})
#Dropping unnecessary columns
df.drop(['Respondent_Number'],axis = 1, inplace=True)
df.info()
#split the feature and target variable
x = df.drop(['Resort_visit'],axis = 1)
x.info()
y = df['Resort_visit']

#group frequency
count = df.groupby(['Resort_visit']).size()
print(count)
#group mean
class_feature_means = pd.DataFrame(columns=y)
for c, rows in df.groupby('Resort_visit'):
    class_feature_means[c] = rows.mean()
class_feature_means = class_feature_means.drop('Resort_visit')
class_feature_means

from statsmodels.multivariate.manova import MANOVA
fit = MANOVA.from_formula('Annual_family_income + Attitude_towards_travel + \
Importance_attached_to_family_skiing_holiday+ \
Household_size+ \
Age_of_head_of_household + \
Amount_spent_on_family_skiing ~ Resort_visit', data=df)
```

```

print(fit.mv_test())

#LinearDiscriminantAnalysis
from sklearn.discriminant_analysis import LinearDiscriminantAnalysis
import seaborn as sns
import numpy as np
lda = LinearDiscriminantAnalysis(n_components = 1)
da = lda.fit(x,y)
y_pred = lda.predict(x)
print(y_pred)

# get Prior probabilities of groups:
da.priors_

#plot
X_new = pd.DataFrame(da.transform(x), columns=["lda1"])
val = 0. # this is the value where you want the data to appear on the y-axis.
X_new["null"] = np.zeros_like(X_new) + val;
sns.scatterplot(data=X_new, x="lda1", y="null", hue=df.Resort_visit.tolist(),palette=["C0", "C1"])


from sklearn import metrics
cm=metrics.confusion_matrix(y,y_pred)
cm
x_axis = [1,2]
y_axis = [1,2]
p=sns.heatmap(cm, annot=True, cmap='BrBG_r',xticklabels=x_axis,yticklabels=y_axis)
p.set_xlabel("Predicted", fontsize = 20)
p.set_ylabel("Actual", fontsize = 20)


#test dataset
df_test= pd.read_csv(r"C:\Users\my pc\Desktop\MBA - BA IIMultivariate analysis
lab\4.DA\DAdata_test.csv")
df_test.columns = df_test.columns.str.replace(" ", "_")
df_test=df_test.rename(columns = {'Annual_family_income_(000s)': 'Annual_family_income'})
df_test.drop(['Respondent_Number'],axis = 1, inplace=True)
test_pred = lda.predict(df_test)
test_pred

```