Psychoinformatics - Week 9 (Exercises)

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```
In []: import numpy as np
    from sklearn import *
    from sklearn import model_selection
    from matplotlib.pyplot import *
    %matplotlib inline
```

1 檢查 machine learning pipeline (8 points)

1.1 請打亂原本的Y觀察正確率是否和chance level (0.33)有差異?若有, why? (4 points)

```
In [ ]: # 本題在研究打亂X和打亂Y有差別嗎?
   iris = datasets.load_iris()
   X=iris.data
   Y=iris.target
   Y2=np.random.permutation(Y)
   print(Y)
   clf=neighbors.KNeighborsClassifier(1)
   clf.fit(X,Y2)
   accuracy=np.mean(clf.predict(X)==Y2)
   print(accuracy)
   2 21
   0.9933333333333333
```

1.1.1 修正 pipeline

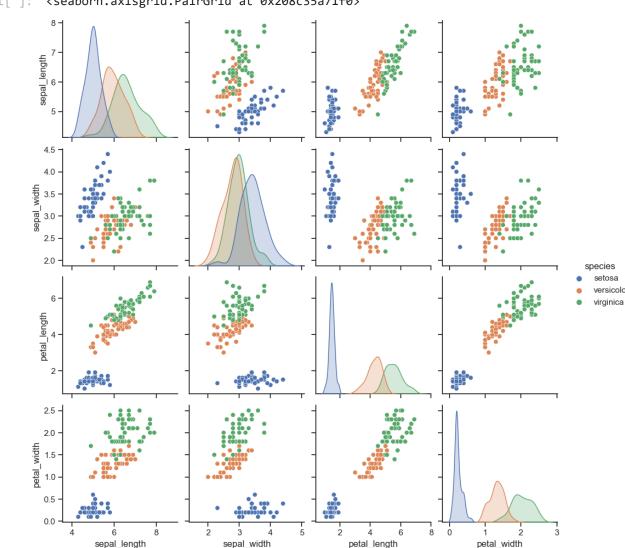
```
In [ ]: clf = neighbors.KNeighborsClassifier(150)
    clf.fit(X, Y2)

accuracy = np.mean(clf.predict(X) == Y2)
    print(accuracy)
```

0.3333333333333333

1.1.2 視覺化特徵

```
import seaborn as sns
iris = sns.load_dataset('iris')
sns.pairplot(iris, hue='species')
```



Out[]: <seaborn.axisgrid.PairGrid at 0x208c35a71f0>

1.1.3 驗證假說

由於初始 n_neighbors 參數被設定為 1 · 即每次只考慮一個最近鄰 · 其中數據點之間可能有一些相似性(如資料集特徵分布) · 模型仍然可以預測一些數據點的標籤 · 這導致了 accuracy 相當接近於 1.0 ·

因此我改將 n_neighbors 參數設定為 150 · 即考慮所有資料點 · 最後得到的 accuracy 為 0.33 · 與 chance level 相同 。

1.2 請用母數或無母數統計檢定以下accuracies中的結果是否和 chance level (0.5)有差異? 若有, why? (4 points)

In []: Y=np.remainder(range(200),2) print(Y) #Y的0和1個數一樣多

```
 \begin{bmatrix} 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0
```

```
In []: # 跑一百次測試:
clf=svm.SVC()
accuracies=[]
for i in range(100):
    X=np.random.rand(200,2) # X取亂數
    kf=model_selection.KFold(len(Y),shuffle=True) # Leave-one-out cross-validation
    sc=model_selection.cross_val_score(clf,X,Y,cv=kf)
    accuracies.append(sc.mean())
```

1.2.1 母數統計檢定

```
In [ ]: # Please do your statistical tests here:
    from scipy.stats import ttest_1samp
    ttest_1samp(accuracies,0.5)
```

Out[]: TtestResult(statistic=-1.6667597470884317, pvalue=0.09872298492582864, df=99)

1.2.2 無母數統計檢定

Out[]: WilcoxonResult(statistic=1896.0, pvalue=0.11431889247617877)

1.2.3 隨機森林分類器

```
In [ ]: from sklearn.ensemble import RandomForestClassifier
    from sklearn.model_selection import train_test_split
    from sklearn.metrics import accuracy_score

X_train, X_test, Y_train, Y_test = train_test_split(X, Y, test_size=0.2, random_state)

clf = RandomForestClassifier()

clf.fit(X_train, Y_train)
    Y_pred = clf.predict(X_test)

accuracy = accuracy_score(Y_test, Y_pred)
    print(f"accuracy: {accuracy}")
```

accuracy: 0.5

1.2.4 驗證假說

模型的預測能力與 chance level 沒有顯著差異(從 ttest 與 wilcoxon p-value 皆大於顯著性水平 0.05 可知)可能是由於數據集特徵不足以有效區分兩個類別·且數據集的 0 和 1 的數量相等(即平衡數據集)·使得模型很難找到明顯的規律。 雖嘗試過其他的分類器·如 Random Forest·但結果皆相同。

Please write your discussion here, if any.

Please submit your notebook in PDF to NTU Cool by next Friday (11/10).