```
— 1
                             5 data=pd.read csv('C:/Users/Admin/Downloads/framingham train.csv')
                            6 data.head()
   Out[150]:
                            male
                                                   age education currentSmoker cigsPerDay BPMeds prevalentStroke prevalentHyp diabetes totChol
                                                                                                                                                                                                                                          svsBP
                                                                                                                                                                                                                                                           diaBP
                                                                                                                                                                                                                                                                                BMI heartRate
                           0 1 0.184211 0.333333 1 0.428571
                                                                                                                                        0 0 0 0 0.176591 0.191489 0.402116 0.190257 0.323232
                                                                                                                                          0
                                                                                                                                                                      0
                                                                                                                                                                                              0
                                     1 0.736842 0.000000
                                                                                                    0.000000
                                                                                                                                                                                                               0 0.262834 0.170213 0.328042 0.221037 0.161616
                          2 0 0.157895 0.666667 1 0.500000 0
                                                                                                                                                                 0 0 0.104723 0.125296 0.296296 0.201406 0.242424
                                     0 0.868421 0.333333
                                                                                                    0 0.000000
                                                                                                                                                                                                                0 0.293634 0.309693 0.507937 0.344401 0.464646
                           4 0 0.947368 0.666667 1 0.285714 0 0 1 0.0297741 0.352246 0.486772 0.390208 0.363636
   7 lr=LogisticRegression(solver='lbfgs')
                          10 lr=lr.fit(x_data,y_data)
11 accuracy = lr.score(x_data,y_data)
12 print(accuracy)
                          0.8522975929978118
—.2
: _ _ aatai=pa.reaa_csv( t:/users/Aamin/Downioaas/test_sampie_i.csv )
               y1_data = data1['TenYearCHD']
x1_data = data1.drop('TenYearCHD', axis = 1)
              accuracy = lr.score(x1_data,y1_data)
print(accuracy)
             a100=((1.28)**2)/(2*100)
b100=1.28*(((accuracy/100)-(((accuracy)**2)/100))+(((1.28)**2)/(4*((100)**2))))**0.5)
c100=(1+(((1.28)**2)/100))
print((accuracy+a100-b100)/c100)
print((accuracy+a100-b100)/c100)
              print((accuracy+a100+b100)/c100)
              v100=1.65*((((accuracy)**2)/100))+(((1.65)**2)/(4*((100)**2))))**0.5)

v100=1.65*(((1.65)**2)/100))

print((accuracy+u100-v100)/w100)
               print((accuracy+u100+v100)/w100)
              u100=((2.58)**2)/(2*100)
              \(\frac{1}{2}\) \(\frac{1}\) \(\frac{1}{2}\) \(\frac{1}{2}\) \(\frac{1}{2}\) \(\frac{1}{2}\) \(\frac{1}{2}\) \
              print((accuracy+u100+v100)/w100)
    31 data4=pd.read_csv('C:/Users/Admin/Downloads/test_sample_4.csv')
32 v4 data = data4['TenYear(HD']
    34 accuracy4 = lr.score(x4_data,y4_data)
35 print(accuracy4\)
  print(accuracy4)
       36
37 e1000=((1.28)**2)/(2*1000)
              F1000=1.28*((((accuracy4)1000)-(((accuracy4)**2)/1000))+(((1.28)**2)/(4*((1000)**2))))**0.5)
g1000=(1+(((1.28)**2)/1000))
print((accuracy4+e1000-f1000)/g1000)
       39
40
41
               print((accuracy4+e1000+f1000)/g1000)
```

```
x1000=((2.58)**2)/(2*1000)
y1000=2.58*((((accuracy4/1000)-(((accuracy4)**2)/1000))+(((2.58)**2)/(4*((1000)**2))))**0.5)
z1000=(1+(((2.58)**2)/1000))
print((accuracy4+x1000+y1000)/z1000)
print((accuracy4+x1000+y1000)/z1000)
     data5=pd.read_csv('C:/Users/Admin/Downloads/test_sample_5.csv')
y5_data = data5['TenYearCHD']
x5_data = data5.drop('TenYearCHD', axis = 1)
accuracy5 = lr.score(x5_data,y5_data)
print(accuracy5)
64 h1800=((1.28)**2)/(2*1800)
```

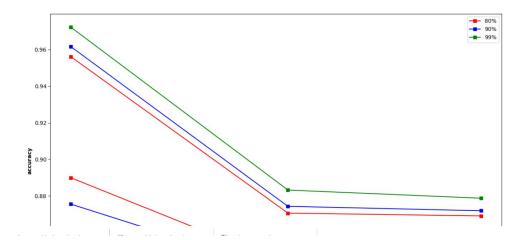
## 一.3

```
2 from matplotlib.font_manager import FontProperties
3 plt.figure(figsize=(15,10),dpi=100,linewidth = 2)
4 samples=[100,1000,1800]
5
6 plt.plot(samples,bb['808'],'s-',color = 'r',label='808')
7 plt.plot(samples,bb['998'],'s-',color = 'g',label='998')
8 plt.plot(samples,db['998'],'s-',color = 'g',label='998')
9 plt.plot(samples,dd['998'],'s-',color = 'r')
10 plt.plot(samples,dd['998'],'s-',color = 'b')
11 plt.plot(samples,dd['998'],'s-',color = 'g')
12 plt.xlabel('*samples,'dd['998'],'s-',color = 'g')
13 plt.ylabel("accuracy", fontweight = "bold")
14 plt.legend(loc = 'best')
15 plt.title("confidence interval", fontsize = 15, fontweight = "bold", y = 1.1)

Out[155]: Text(0.5, 1.1, 'confidence interval')
```



## confidence interval



## —.4

```
In [156]: 1 #(a) 樣本數越多:信賴區間越小,代表越準確,越小的範圍中資料反而比較多 2 #(b) 信心水準越大,信賴區間越大,表示較多的資料在這範圍中
```

## 二.1

```
In [158]: 1 from sklaern.model.selection import train_texts.

2 from sklaern.model.selection import train_texts.

3 from sklaern.model.selection import train_texts.

4 import nummy as np

5 from sklaern.model.selection import train_texts.

6 from sklaern.model.selection import train_texts.

7 clf = GaussianMB()

8 X-adax.derog('diagnosis'; axis = 1)

9 yealst['diagnosis']

10 (If = Clf.fit(X,y)

11 [159]: 1 | accercoss_val_predict(clf,X,y,cv=10),scorings'accuracy')

2 | scores = cross_val_predict(clf,X,y,cv=10),scorings'accuracy')

2 | scores = cross_val_predict(clf,X,y,cv=10),scorings'accuracy')

10 | file | f
```

```
In [163]: 1 confusion_matrix(y,scores)
               Out[163]: array([[357, 0], [212, 0]], dtype=int64)
             In [165]: 1 | score= cross_val_score(clf,X,y,cv=10,scoring='accuracy')
2 | scores = cross_val_predict(clf,X,y,cv=10)
3 | print(score)
4 | print(score.mean())
5 | print(scores)
                                                                                                 5 print(scores)
[0.9122887 0.84210526 0.9122807 0.89473684 0.98245614 0.89473684 0.89473684 0.983473684 0.89473684 0.89473684 0.89473684 0.89473684 0.89473684 0.89473684 0.89473684 0.89473684 0.89473684 0.89473684 0.89473684 0.98245614 0.89473684 0.89473684 0.98245614 0.89473684 0.89473684 0.98245614 0.89473684 0.89473684 0.98245614 0.89473684 0.89473684 0.98245614 0.89473684 0.98245614 0.89473684 0.98245614 0.89473684 0.98245614 0.89473684 0.98245614 0.89473684 0.98245614 0.89473684 0.98245614 0.89473684 0.98245614 0.89473684 0.98245614 0.89473684 0.98245614 0.89473684 0.98245614 0.89473684 0.98245614 0.89473684 0.98245614 0.89473684 0.98245614 0.89473684 0.98245614 0.89473684 0.98245614 0.89473684 0.98245614 0.89473684 0.98245614 0.89473684 0.98245614 0.89473684 0.98245614 0.89473684 0.98245614 0.89473684 0.98245614 0.89473684 0.98245614 0.89473684 0.98245614 0.89473684 0.98245614 0.89473684 0.98245614 0.89473684 0.98245614 0.89473684 0.98245614 0.89473684 0.98245614 0.89473684 0.98245614 0.89473684 0.98245614 0.89473684 0.98245614 0.982473684 0.98245614 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.98247364 0.982473684 0.982473684 0.982473684 0.982473684 0.982473684 0.
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                                      In [166]: 1 confusion_matrix(y,scores) 2 #由上述比較可知,decision預測準確度>SVM>naive
                                      Out[166]: array([[336, 21], [30, 182]], dtype=int64)
```