資料探勘 06160485

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1. 資料

1-1 來源

UCI 糖尿病性的視網膜病變資料集

格式:.arff 打開可發現

@relation:代表有一個資料表叫做dr。

@attribute:共有20個屬性,有19個特徵及1個答案,型別為數字與{0,1}的兩個類別。

@data:資料的本身,很像CSV使用逗點隔開。

```
1 prelation dr
         @attribute 0 numeric
        @attribute 2 numeric
@attribute 3 numeric
      8 @attribute 5 numeric
      9 @attribute 6 numeric
      10 @attribute 7 numeric
8
        @attribute 8 numeric
      13 @attribute 10 numeric
     16 @attribute 13 numeric
     17 @attribute 14 numeric
     18 @attribute 15 numeric
     19 @attribute 16 numeric
        @attribute 17 numeric
        @attribute 18 numeric
        @attribute Class {0,1}
```

1-2 說明

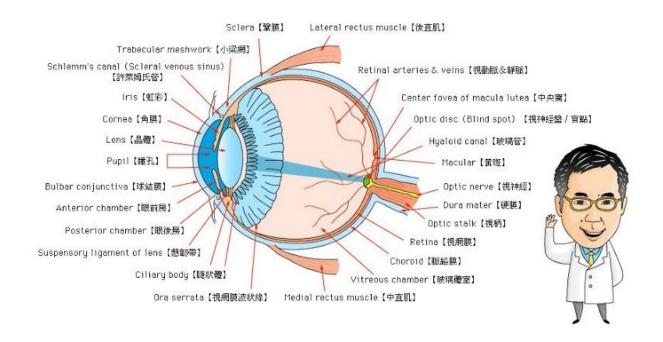
取自Messidor資料集,預測圖像是否包含糖尿病性視網膜病變的徵兆。使用在Balint Antal, Andras Hajdu:基於集合體的糖尿病性視網膜病變自動篩查系統,所有特徵都代表檢測到的病變資訊,且使用圖像分析和特徵提取以及分類。

適用問題	分類
資料筆數	1151
屬性數量	20
缺值	無

1-3 屬性介紹

特徵

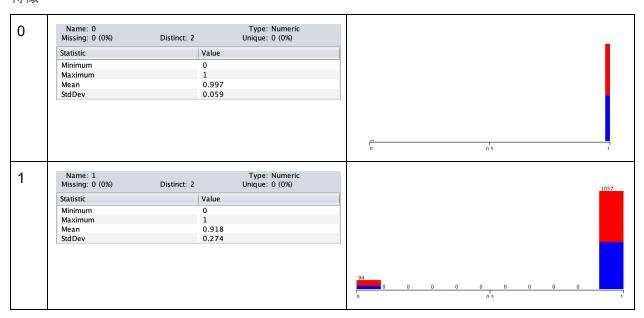
0. 檢測質量	0:差, 1:好
1. 預篩選的結果(嚴重視網膜異常)	1:有, 0:沒有
2-7. MA檢測結果,每一項特徵代表在信心水 準(alpha)下0.5 1.0	數字
8-15. 包含2-7的滲出液訊息,由病變數/ROI直 徑歸一化特徵,以縮減不同大小的差距	數字
16. 黃斑部到盲點的歐式距離,也使用ROI直 徑標準化	數字
17. 盲點直徑	數字
18. 基於AM/FM的分類結果	0:沒有糖尿病視網膜病變(DR), 1,2,3:代表不同的DR標籤

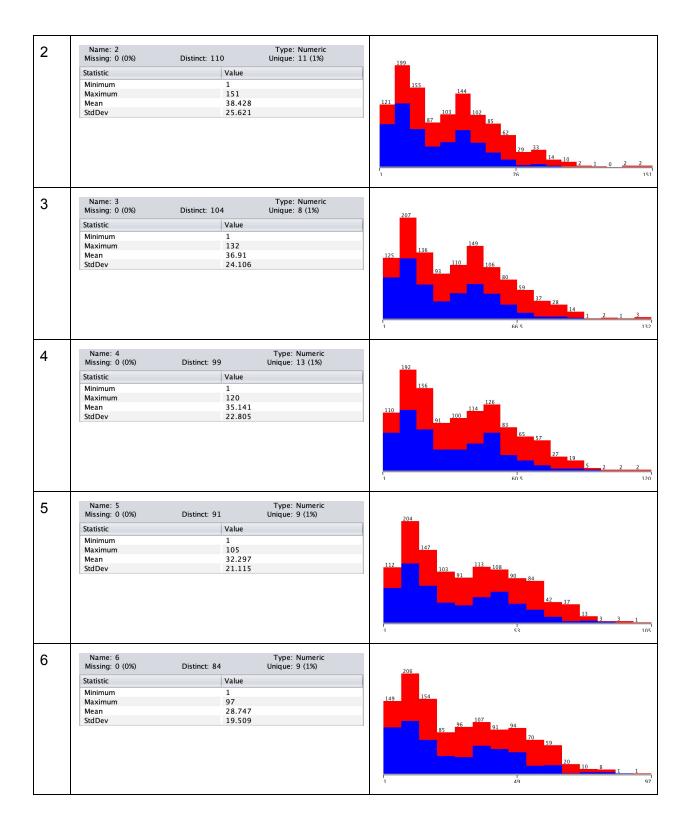


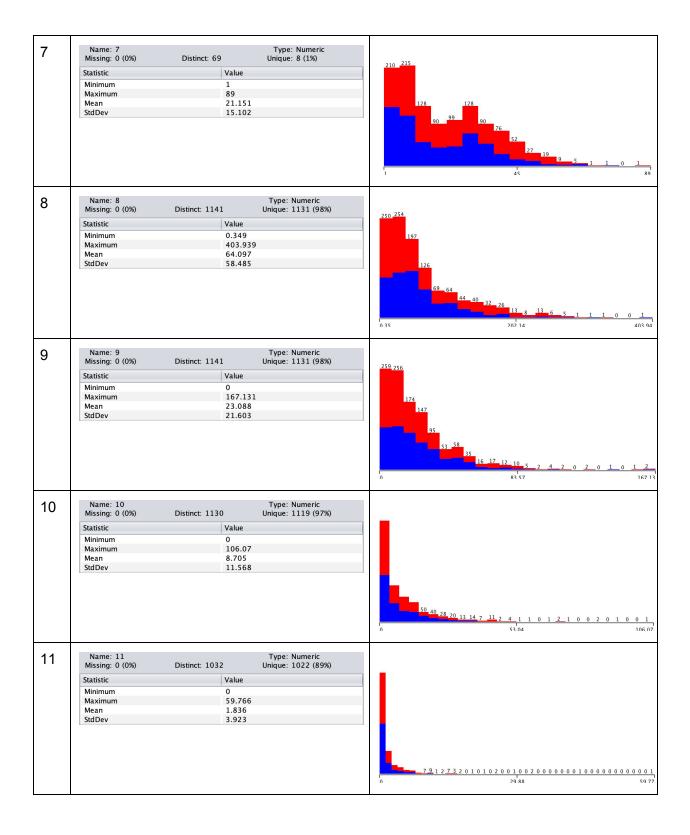
1-4 屬性資訊

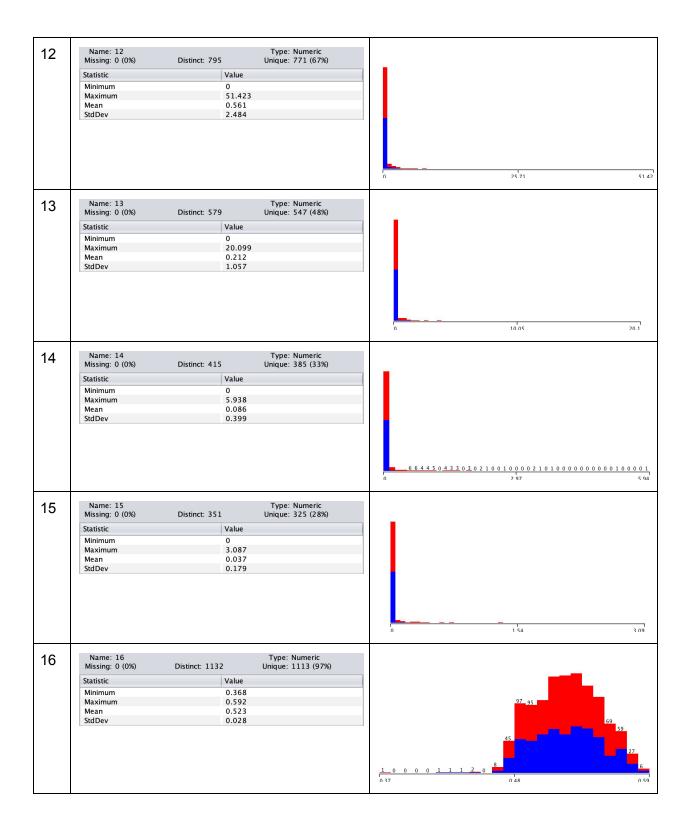
我們將原始資料讀入weka並視覺化結果

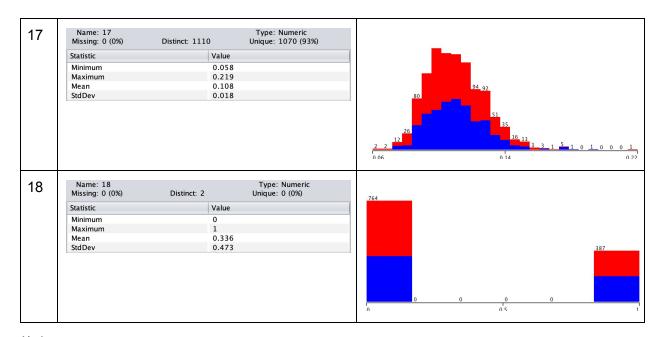
特徵



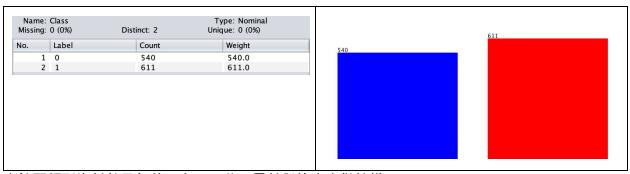








答案



由於兩類別資料數目都差不多,因此不需針對答案去做抽樣

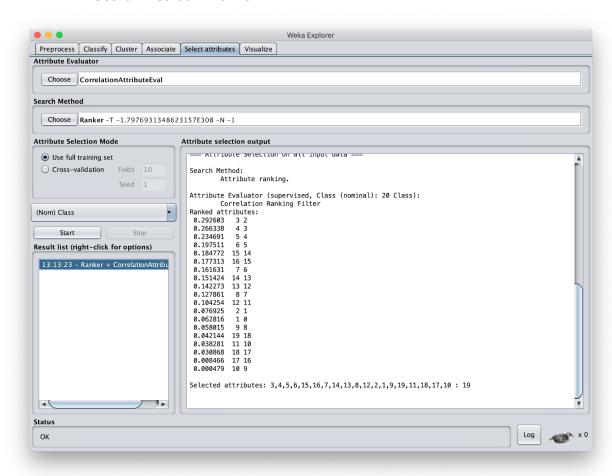
1-5 特徵工程

1-5-1 基於關聯性

● Weka操作步驟:

Select attributes

Attribute Evaluater > CorrelationAttributeEval Search Method > Ranker

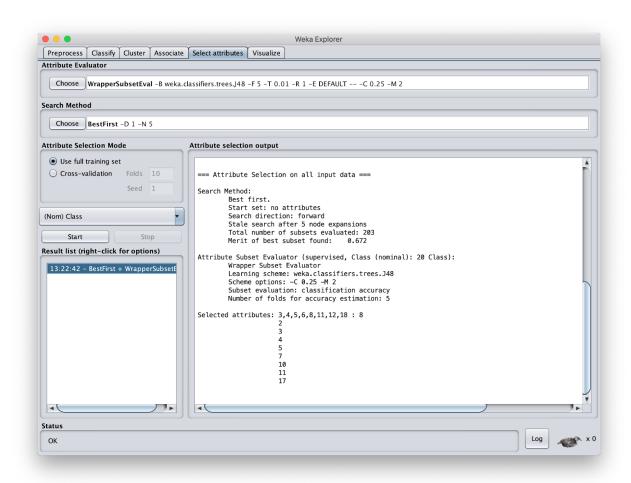


1-5-2 基於學習

● Weka操作步驟:

Select attributes

Attribute Evaluater > WrapperSubsetEval > classifer > tree > J48 Search Method > BestFirst



1-6 前處理原因

1. 觀察資料值域

由1-4可知,特徵19個標籤最大值從0.219-403不等,因此資料前處理要做正規化的動作, 避免訓練模型時,因特徵值域範圍不同造成某項特徵重要性權重的錯誤。

2. 抽樣

我們將原始資料進行抽樣訓練,使得訓練時使用的時間及空間會較少。

3. 降低維度

一般資料都有大量的特徵, 我們使用主成分分析法(PCA)而不是單純的特徵選擇來選擇對模型比較有幫助的特徵, 把資料集的複雜度變小有以下的優點:

- 避免維度詛咒
- 較好視覺化
- 消除資料本身雜訊
- 更容易解釋

2. 資料前處理

2-1 資料正規化

● Weka操作步驟:

Preprocess>Filter

filters>unsupervised>attribute>Normalize

efore			after		
Name: 8 Missing: 0 (0%)	Distinct: 1141	Type: Numeric Unique: 1131 (98%)	Name: 8 Missing: 0 (0%)	Distinct: 1141	Type: Numeric Unique: 1131 (98
Statistic	Value		1	100 100	
Minimum	0.349		Statistic	Value	
Maximum	403.9	39	Minimum	0	
Mean	64.09		Maximum	1	
StdDev	58.48		Mean	0.158	
			StdDev	0.145	
Statistic Minimum Maximum Mean StdDev	Value 0 167.1 23.08 21.60	3	Statistic Minimum Maximum Mean StdDev	Value 0 1 0.138 0.129	
Name: 17 Missing: 0 (0%)	Distinct: 1110	Type: Numeric Unique: 1070 (93%)	Name: 17 Missing: 0 (0%)	Distinct: 1110	Type: Numeric Unique: 1070 (93
Statistic	Value		Statistic	Value	
Minimum	0.058				
Maximum	0.219		Minimum	0	
Mean	0.108		Maximum	1	
StdDev	0.018		Mean	0.313	
			StdDev	0.111	

● 說明:

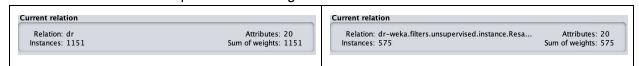
由上表可知,原始資料中特徵17的值域是[0.058,0.219]、特徵8的的值域是[0.349,403.939],標準差為0.018及58.485,平均為0.108、64.097,兩者相差640倍,有可能會因值域造成重要性的錯誤,因此我們將其正規化,使特徵的值域範圍在[0,1]之間,每一個特徵的重要性都相同。

2-2 資料抽樣

● Weka操作步驟:

Preprocess>Filter

filters>unsupervised>instance>resample sampleSizePresentage:50

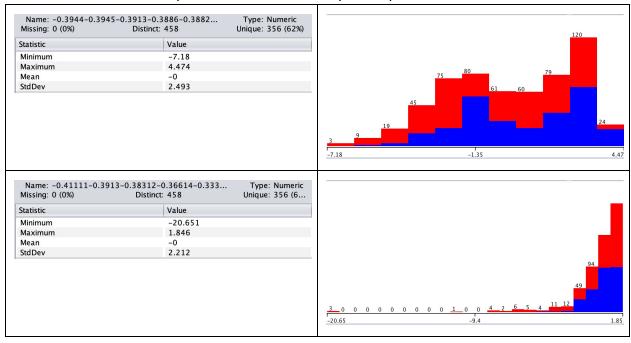


2-3 主成分分析法(PCA)

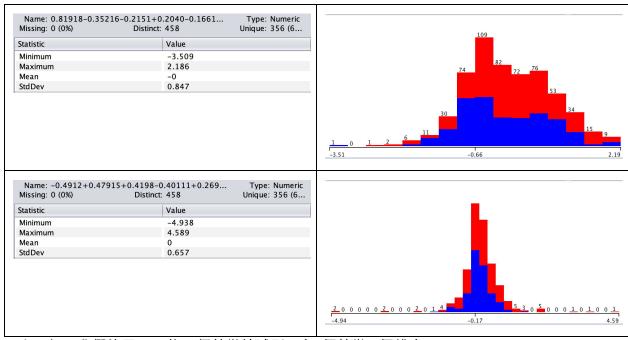
● Weka操作步驟:

Preprocess>Filter

filters>unsupervised>attribute>PrincipleComponent

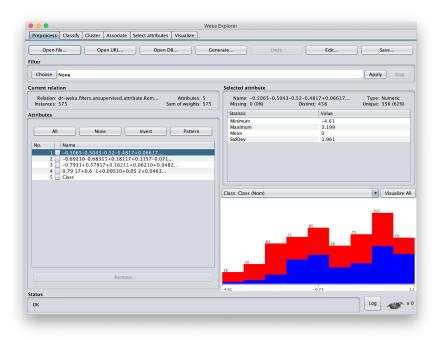


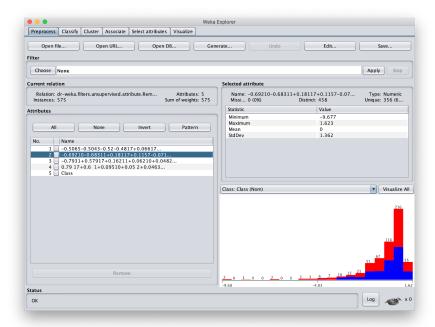
Missing: 0 (0%)	0+0.4098-0.34115-0.3241 Distinct: 458	Type: Numeric Unique: 356 (6	
Statistic	Value		
Minimum	-12.477		
Maximum	7.065		
Mean	0		
StdDev	1.463		
			1 0 0 0 0 0 0 1 0 0 2 0 1 3 1 3 7 1 5 1 3 0 - 12.48 -2.71 7.
Name: 0.71816-0.5483 Missing: 0 (0%)	17+0.2830-0.2521+0.1161 Distinct: 458	Type: Numeric Unique: 356 (6	92 <u>93</u> 92
Statistic	Value		
Minimum	-7.765		70_
Maximum	4.228		62
Mean	0		
StdDev	1.07		24
			1 0 0 0 0 0 0 0 0 0 1 0 1 0 1 0 1 0 0 0 0 0 0 0 1 0 1 0 1 0
Name: 0.7931-0.4917 Missing: 0 (0%)	-0.2530+0.12618-0.10211 Distinct: 458	Type: Numeric Unique: 356 (62%)	
tatistic	Value		
Minimum	-4.266		
Maximum	6.312		
Mean	0		
StdDev	1.021		
			3 9 10 11 4 0 5 11 5 0 1 0 0 0 0 0 0 0 0 0 0 0 0 0
Name: 0.8930+0.3011 Missing: 0 (0%)	-0.23918-0.18316+0.0671 Distinct: 458	Type: Numeric Unique: 356 (6	
Missing: 0 (0%)			
Missing: 0 (0%) statistic	Distinct: 458		
Missing: 0 (0%) Statistic Minimum	Distinct: 458 Value		
Missing: 0 (0%) Statistic Minimum Maximum	Distinct: 458 Value -21.268		
Missing: 0 (0%) statistic Minimum Maximum Mean	Distinct: 458 Value		
Missing: 0 (0%) Statistic Minimum Maximum Mean	Distinct: 458 Value		
Missing: 0 (0%) statistic Minimum Maximum Mean	Distinct: 458 Value		-21.27 -9.98 1
Missing: 0 (0%) statistic Minimum Maximum Mean Std Dev Name: -0.6417-0.5413 Missing: 0 (0%)	Distinct: 458 Value		-21.27 -9.98 1 88 91 88 91
Missing: 0 (0%) Statistic Minimum Maximum Mean StdDev Name: -0.6417-0.5411 Missing: 0 (0%)	Distinct: 458 Value -21.268 1.308 -0 0.997 0.6-0.3818-0.3531-0.0790 Distinct: 458 Value	Unique: 356 (6	
Missing: 0 (0%) Statistic Minimum Maximum Mean Std Dev Name: -0.6417-0.5411 Missing: 0 (0%) Statistic Minimum	Distinct: 458 Value	Unique: 356 (6	
Missing: 0 (0%) Statistic Minimum Maximum Mean Stid Dev Name: -0.6417-0.5411 Missing: 0 (0%) Statistic Minimum Maximum	Distinct: 458 Value	Unique: 356 (6	
Missing: 0 (0%) Statistic Minimum Maximum Mean Std Dev Name: -0.6417-0.5411 Missing: 0 (0%) Statistic Minimum Maximum Maximum Mean	Distinct: 458 Value	Unique: 356 (6	88 91 88 91
Missing: 0 (0%) Statistic Minimum Maximum Mean Std Dev	Distinct: 458 Value	Unique: 356 (6	



以上可知, 我們使用PCA將19個特徵縮減到只有9個特徵(9個維度)

2-4 特徵選擇且資料處理過





3. 心得

在此次的期中作業中,我們使用了weka學習如何對資料做前處理。從此次練習中可以觀察到,weka是一個很簡單可以快速對資料做處理的資料探勘工具,對於現在正在學習資料探勘技術是非常有幫助的,除了上課學習的理論基礎,更有實作的配合。此外,以前在線性代數老師有提到的主成分分析法(PCA)對於當時的我覺得很抽象,但在做期中專案的時候就了解原來這個是多麼的重要,可以避免一些訓練上的問題,如維度詛咒等等。

綜合上述,我覺得此次期中作業練習除了了解整個做資料探勘的步驟,更對於課堂上的理論及更早線性代數的資料處理演算法有更一步的了解,透過實作能更了解每一個步驟的意義,因此此次作業對於資料探勘是非常重要且有意義的。

3-1 處理完的資料樣式

```
lap{\ } result.arff 	imes
           @relation 'dr-weka.filters.unsupervised.instance.Resample-S1-Z50.0-weka.filters.unsupervised.attribute.Norm
          @attribute -0.3944-0.3945-0.3913-0.3886-0.3882... numeric
          @attribute -0.41111-0.3913-0.38312-0.36614-0.33315... numeric
       5 @attribute 0.4559+0.42110+0.4098-0.34115-0.32414... numeric
          @attribute 0.71816-0.54817+0.2830-0.2521+0.11612... numeric
        7 @attribute 0.7931-0.4917-0.2530+0.12618-0.10211... numeric
       8 @attribute 0.8930+0.3011-0.23918-0.18316+0.06715... numeric
9 @attribute -0.6417-0.54116-0.3818-0.3531-0.0790... numeric
      10 @attribute 0.81918-0.35216-0.2151+0.2040-0.16611... numeric
      11 @attribute -0.4912+0.47915+0.4198-0.40111+0.26914... numeric
      12 @attribute Class {0,1}
          2.581926.0.678383.-0.899611.1.07922.1.33701.-0.077475.0.531243.1.253136.-0.475156.1
          1.411857,0.908347,-1.205703,1.744797,0.894571,-0.429475,-1.099126,0.705921,-0.362572,1
      17 \qquad \textbf{2.730955,0.60702,-1.194172,1.507683,-3.577453,-1.018252,0.382158,-0.796566,0.903264,1}
          -1.445156,1.229457,-0.330051,0.065543,0.833124,0.307533,1.140851,0.17468,-0.208049,0
          -2.076613,1.387654,-0.510786,-0.737506,-0.003501,0.249906,-0.136778,0.010597,-0.27066,1
          2.704716,0.442988,-0.304449,1.31163,0.661428,0.244181,0.441774,-1.223026,0.382889,1
      21 2.727248, 1.133922, -1.623178, 0.33959, 0.767179, 0.539493, 1.067678, -0.586893, -0.25247, 1
      22 2.677917,1.115074,-1.410628,1.14311,0.772882,0.331523,0.526734,-0.974087,-0.014943,0
          -2.976502,1.401719,-0.064854,-0.010017,0.757155,0.197097,0.874817,0.376337,-0.183091,1
      24 2.48028,0.999815,-1.192551,-0.123226,-0.241847,0.382696,-0.641259,-1.1856,0.073129,1
```