

Consider again the data set in [week 4 assignment](#), and recall that we have transformed the data into classification and regression sets.

1. (Classification using GDA) Your task is to use Gaussian Discriminant Analysis (GDA) to build a classification model. To complete this assignment, make sure you:
 - a) Write your own code to implement the GDA algorithm. **(Do not use built-in classification functions.)**
 - b) Clearly explain how the GDA model works and why it can be used for classification, in particular this data set.
 - c) Train your model on the given dataset and report its accuracy. Be explicit about how you measure performance (e.g., accuracy on a test set, cross-validation, etc.).
 - d) Plot the decision boundary of your model and include the visualization in your report.

a) code 在另外的文字檔

b) 假設在每個類別 k 的資料 x 都是從 Gaussian distribution 中得到的 (此以分成 2 類說明)

ie, $\mu_k \in \mathbb{R}^n$

$$P(x|y=k) = N(\mu_k, \Sigma_k) = \frac{1}{(2\pi)^{\frac{n}{2}} |\Sigma_k|^{\frac{1}{2}}} \exp\left(-\frac{1}{2}(x-\mu_k)^T \Sigma_k^{-1} (x-\mu_k)\right), k=0,1$$

$$\text{且 } P(y) = \phi^y (1-\phi)^{1-y}$$

$$\left(\begin{array}{l} \text{若 } \Sigma_0 = \Sigma_1, \text{ 則是 LDA} \\ \text{若 } \mu_0 \neq \mu_1 \text{ or } \Sigma_0 \neq \Sigma_1, \text{ 則是 QDA} \end{array} \right)$$

將一部分資料 (training data) : $\{x^{(i)}, y^{(i)}\}$ 代入 likelihood function $L(\theta) = \prod_{i=1}^n P(x^{(i)}, y^{(i)})$

接著求出 $\theta^* = \arg\max_{\theta} L(\theta) = \arg\min_{\theta} -L(\theta)$, 即為 $\mu_k^*, \Sigma_k^*, \phi^*, k=0,1$

則給新的資料 \tilde{x} , 則可計算出

$$P(\tilde{x}|y=k) = N(\mu_k^*, \Sigma_k^*)$$

$$P(y) = (\phi^*)^y (1-\phi^*)^{1-y}$$

再透過貝氏定理, 計算

$$P(y=k|\tilde{x}) = \frac{P(\tilde{x}|y=k) P(y=k)}{P(\tilde{x}|y=0) P(y=0) + P(\tilde{x}|y=1) P(y=1)}, k=0,1$$

比較 $P(y=0|\tilde{x})$ 和 $P(y=1|\tilde{x})$ 的值.

若 $P(y=0|\tilde{x}) > P(y=1|\tilde{x})$, 則 \tilde{x} 判斷屬於類別 0

若 $P(y=0|\tilde{x}) < P(y=1|\tilde{x})$, 則 \tilde{x} 判斷屬於類別 1

c) 將 HW4 已分成 label 0, label 1 的 8040 筆資料分成

training data : 5628 筆 (70%), test data : 2412 筆 (30%)

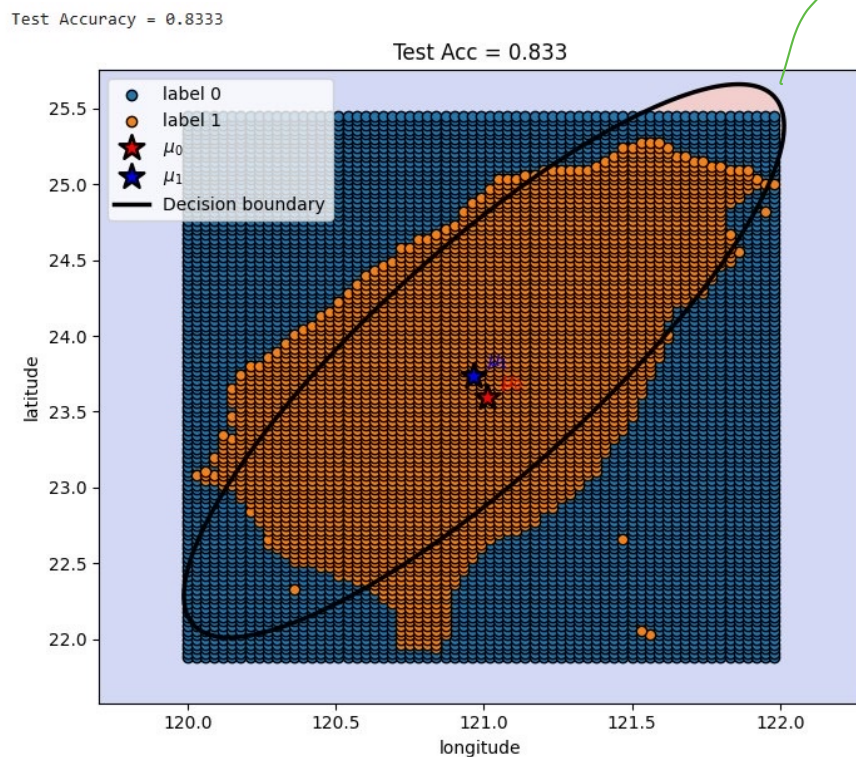
此沒有 stop criteria, 所以不須

validation data

此 model 的 accuracy 用 $\text{test accuracy} \equiv \frac{\text{正確預測的 test data 數量}}{\text{test data 數量}}$ 表示

結果為 $0.8333 = 83.33\%$

d)



- $C(\vec{x})$ be your classification model, and
- $R(\vec{x})$ be your regression model.

$$h(\vec{x}) = \begin{cases} R(\vec{x}), & \text{if } C(\vec{x}) = 1 \\ -999, & \text{if } C(\vec{x}) = 0. \end{cases}$$

d) Include plots or tables that demonstrate the behavior of your model.

Coefficient of Determination $R^2 = 1 - \frac{\sum_{i=1}^n (y_i - \hat{y}_i)^2}{\sum_{i=1}^n (y_i - \bar{y})^2} = 0.158$, where \bar{y} is the average of the actual values

↑
越接近 1 越好

↑
代表 R^2 可以解釋大約 15.8% 的 y 的變化

[資料偵測]

分類特徵欄位: ['longitude', 'latitude'] 分類標籤欄: label

回歸特徵欄位: ['longitude', 'latitude'] 回歸目標欄: value

[切分]

分類: train=6432, test=1608

回歸: train=2796, test=699

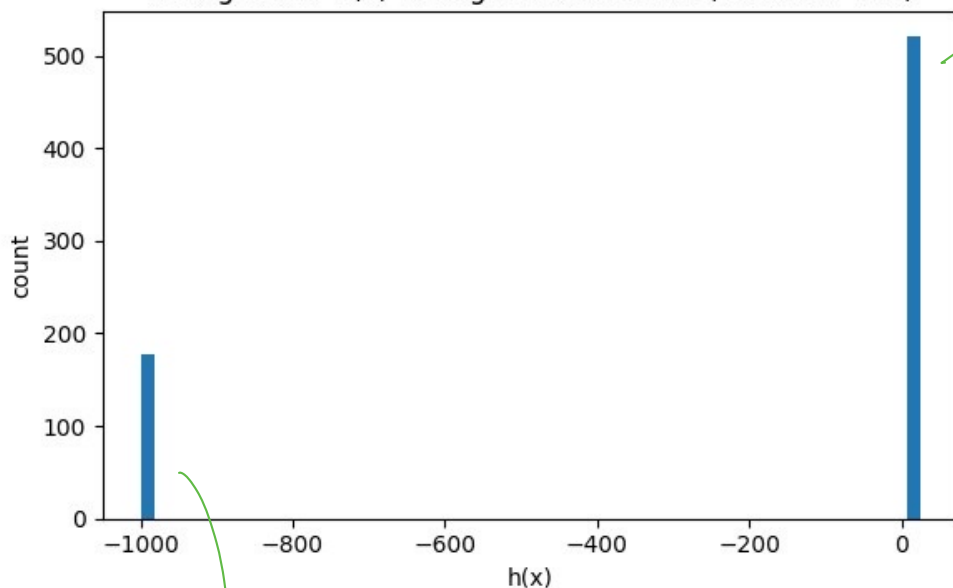
[評估結果]

Classification Test Accuracy (on classification_dataset): 0.8302

在 regression 測試集中, $C(x)=1$ 的樣本數: 521 / 699

Regression metrics on those $C(x)=1$ samples: {'MAE': 4.383282718402041, 'RMSE': 5.726880065482881, 'R2': 0.15836613517403975}

Histogram of $h(x)$ on regression test set (includes -999)

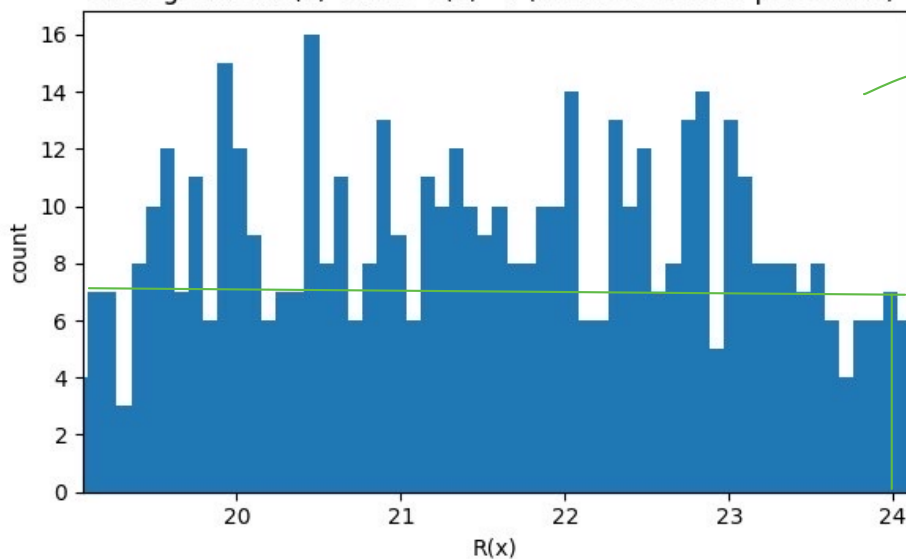


699筆data中被model C
判斷為0的data有178筆

→ 699筆data中被model C
判斷為1的data有521筆

[連續分佈範圍參考] 1% ~ 99% 百分位: [19.063, 24.089]

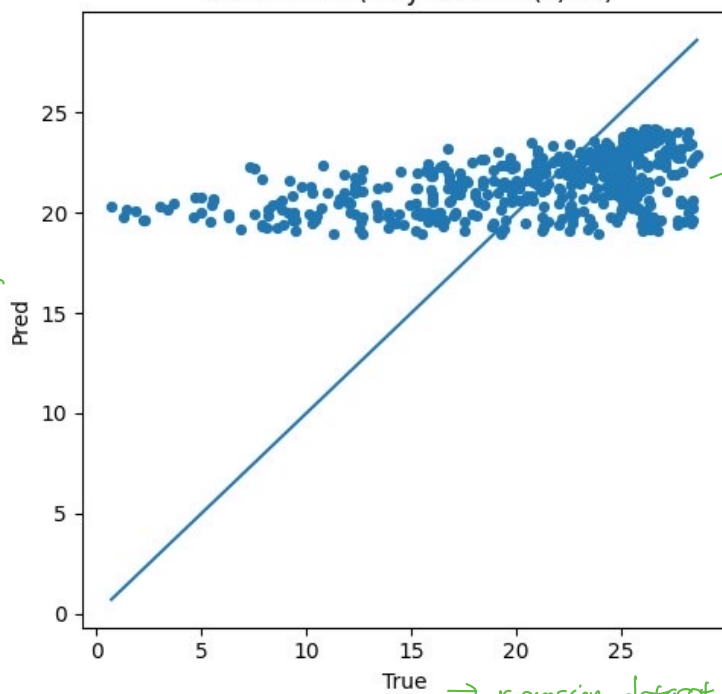
Histogram of $R(x)$ where $C(x)=1$ (zoomed to 1-99 percentile)



→ 將上圖右邊長條形放大來看,

在521筆data中,有7筆data
的 $h(x)$ 值為24

True vs Pred (only where $C(x)=1$)



model $R(x)$ 的值
(預測) ←

藍點為被判斷 $C(x)=1$ 的 521 筆 data

可看出 model h 預測所有的 data 的值都差不多 (19 ~ 24 之間)

表示 model 只學到一個接近平均的結果

→ regression_dataset 裡真實的值

[Piecewise 行為檢查]

$h(x)$ 中等於 -999 的個數：178

$h(x)$ 中不等於 -999 的個數：521 (這些點是由 $R(x)$ 給的)