

11463154 劉祐睿 2025/10/14

## 手寫筆記

A Study of Efficient GNSS Coordinate Classification

Strategies for Epidemic Management

Subject: .....

11463154

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Date: .....

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介紹: 本研究想透過 GNSS (全球導航衛星系統) 將各類傳染性疾病 (例: COVID-19, 猴痘) 影響減少, 本研究透過 GNSS 與 KNN (K-Nearest Neighbors) 分類技術預測人群分佈, 可作為疫情時的輔助工具

PIP (Point In Polygon) 技術

判斷一個點是否在多邊形內的方法

- Ray Casting (射線法)

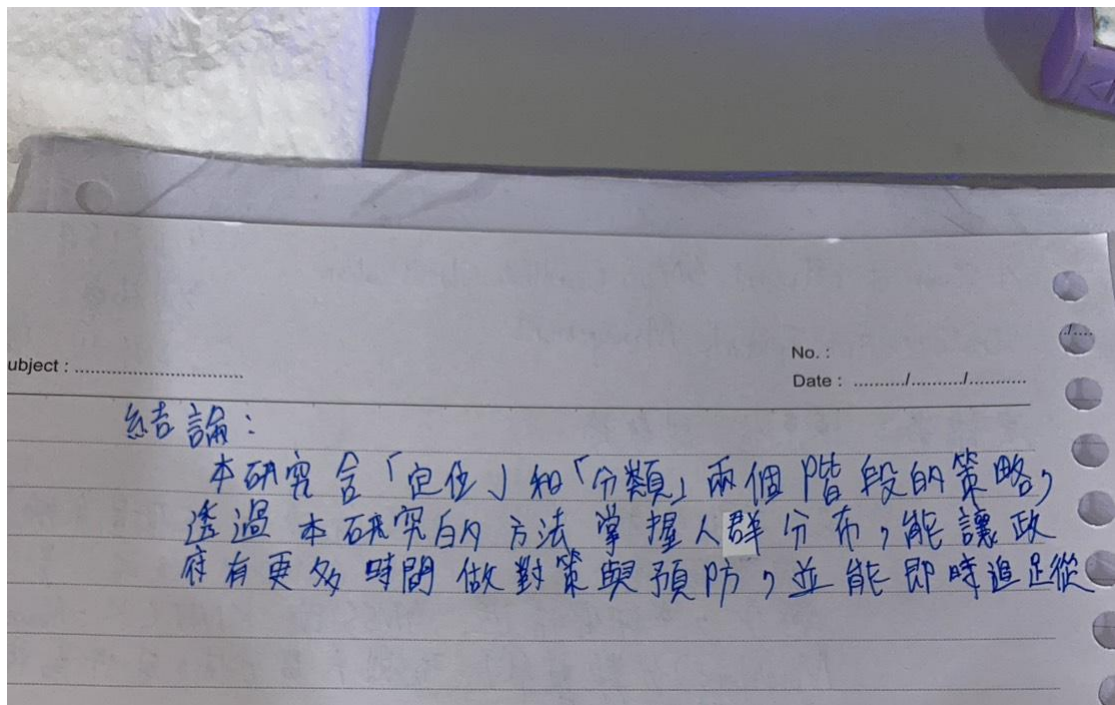
- Winding Number (纏繞數法)

皆用多邊形的邊界判斷點在外或內部

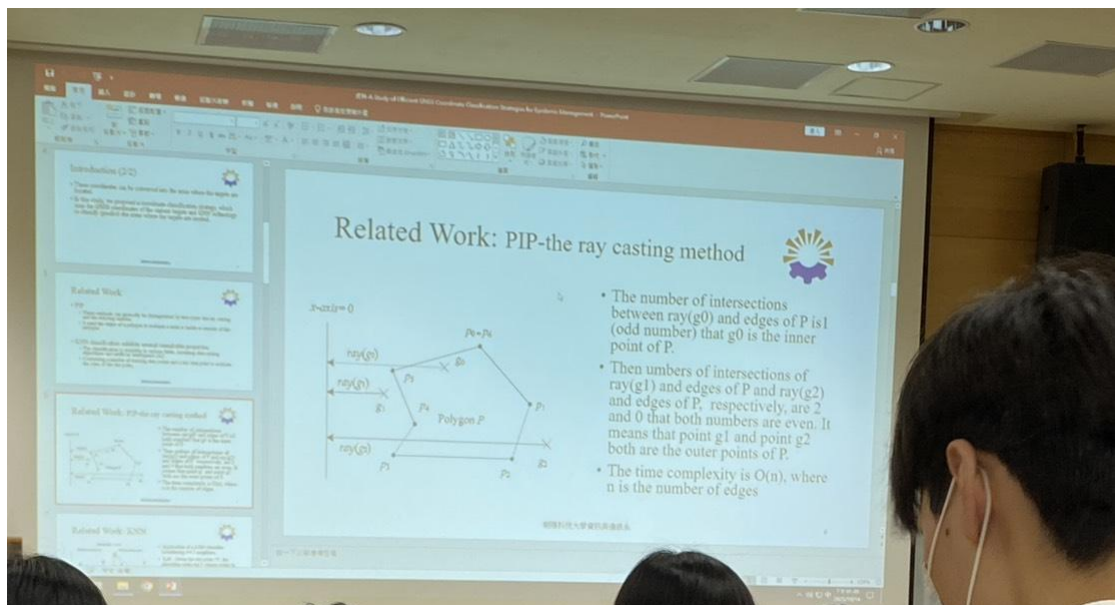
KNN (K-最鄰近) 分類法

由一個或一組測試資料點透過鄰近距離的計算, 來預測測試點的類別

陳忠信副教授補充: 這是以以前一個險書打卡判別打卡地點的題目延伸而來, 來研究題目有前景也有很大研究價值但實際使用於現實社會有隱私方面的問題, 因本題目需透過手機上 GPS 來獲得民眾位置



## 上課紀錄

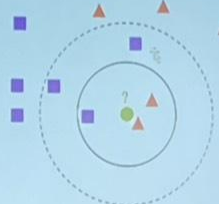


## PIP 範例與介紹



## Related Work: Traditional KNN

- In the KNN classification, three steps are involved for a test data point as follows:
  - Step 1. It evaluates the Euclidean distance between the sample and the test point, with a time complexity of  $O(nTD)$ , where  $nTD$  is the data set size.
  - Step 2. It sorts the training dataset based on Euclidean distances with an  $O(nTD^2)$  time complexity.
  - Step 3. It uses the majority classification rule to predict the class of the test point, with a time complexity of  $O(k)$ , where  $k$  is the number of neighbors in the KNN classification.
- For example  $k=3$

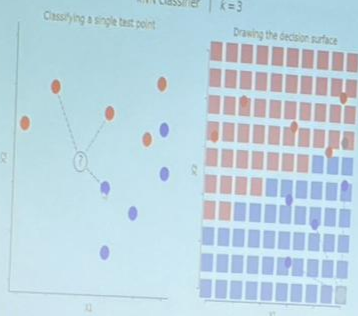


Source: [https://en.wikipedia.org/wiki/K-nearest\\_neighbors\\_algorithm](https://en.wikipedia.org/wiki/K-nearest_neighbors_algorithm)

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## Related Work: KNN

k-NN classifier |  $k=3$



- Application of a  $k$ -NN classifier considering  $k=3$  neighbors.
- Left - Given the test point "2", the algorithm seeks the 3 closest points in the training set, and adopts the majority vote to classify it as "class red".
- By iteratively repeating the prediction over the whole feature space ( $X_1, X_2$ ), one can depict the "decision surface".

Source: [https://en.wikipedia.org/wiki/K-nearest\\_neighbors\\_algorithm](https://en.wikipedia.org/wiki/K-nearest_neighbors_algorithm)

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## KNN 範例與介紹

## Experiment: Class distributions



- The class distributions of Type-1 and Type-2. Type-1 has 12 classes and Type-2 has 256 classes.



(a) Type-1



(b) Type-2

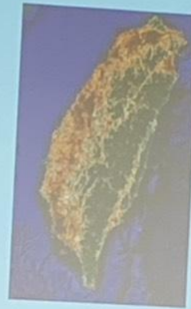
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## Related Work: One-Area and Cell/Rectangle Based PIP



. Distribution of points in hot spot one with 6024 points.

- (a) Distribution of spots and (b) distribution of highways.



(a)



(b)

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實驗以台灣城市為例

# Experiment: Classification Time



Table 2. Classification time based on different sizes in Type-1, where size is the average number of training data points per class.

Classification \ Size	2	4	8	16	32	64	128
knn	$5.24 \times 10^{-6}$	$1.42 \times 10^{-5}$	$4.36 \times 10^{-5}$	$1.52 \times 10^{-4}$	$5.60 \times 10^{-4}$	$2.30 \times 10^{-3}$	$9.43 \times 10^{-3}$
wknn	$6.66 \times 10^{-6}$	$1.57 \times 10^{-5}$	$5.17 \times 10^{-5}$	$1.61 \times 10^{-4}$	$5.95 \times 10^{-4}$	$2.41 \times 10^{-3}$	$9.90 \times 10^{-3}$
awknn	$3.08 \times 10^{-6}$	$5.42 \times 10^{-6}$	$6.25 \times 10^{-6}$	$1.27 \times 10^{-5}$	$1.96 \times 10^{-5}$	$5.07 \times 10^{-5}$	$1.79 \times 10^{-4}$

Note: unit—seconds. Page 13

Table 3. Classification time based on different size in Type-2.

Classification \ Size	2	4	8	16	32	64	128
knn	$3.17 \times 10^{-3}$	$1.34 \times 10^{-2}$	$6.38 \times 10^{-2}$	$2.74 \times 10^{-1}$	$1.12 \times 10^0$	$4.91 \times 10^0$	$2.29 \times 10^1$
wknn	$3.21 \times 10^{-3}$	$1.37 \times 10^{-2}$	$6.48 \times 10^{-2}$	$2.79 \times 10^{-1}$	$1.14 \times 10^0$	$5.02 \times 10^0$	$2.31 \times 10^1$
awknn	$5.51 \times 10^{-5}$	$2.47 \times 10^{-4}$	$5.39 \times 10^{-4}$	$2.09 \times 10^{-3}$	$7.55 \times 10^{-3}$	$2.96 \times 10^{-2}$	$4.07 \times 10^{-2}$

Note: unit—seconds. Page 13

## 實驗結果