Comp309 Assignment 3 Zhanghao Kong 300432074

2.2 Core: Exploring and understanding the Data [40 marks]

• (20 marks) Highlight the findings of your dataset exploration.

Business understanding:

Data mining goal: The overall aim of this assignment is to predict the price of electricity in Wellington.

Core: The model I build needs to be able to predict the price of each instance.

Data understanding:

With the given training set, there are 64 attributes(ID,62 place names and Price). Testing set has 63 attributes excluding the Price attribute. The ID represents the number of each instance. 62 place names represent the generation of a given power station (in kWh) for a given 30-minute time period. The Price attribute represents the price of electricity (measured in \$/MWh).

Data quality: Data quality is quite high because it has many attributes without any missing values. All attributes are related to the goal attribute(except ID for testing). Amount of data for training is big enough.

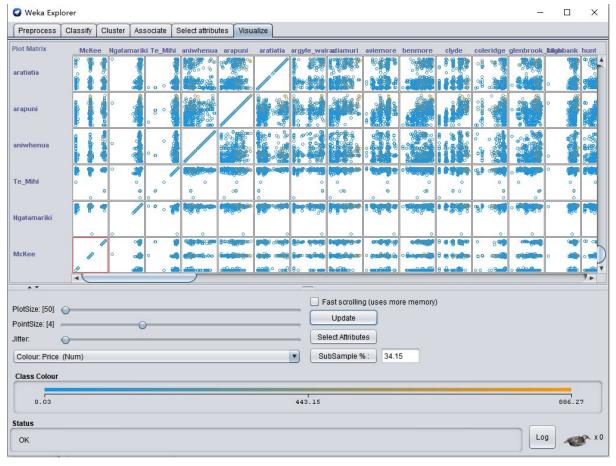
Data completeness: No missing value so have a good data completeness.

Understanding of Features:In the given data sets, attribute "ID" is not related to other attributes, it's just used for marking instances. All other attributes except "price" and "ID" are not related to each other, they just record the name of the power station. But they all have a relation with "price". Below is the Correlation Ranking Filter method in Weka. All attributes in the data set have a less or more effect on "price".

```
Attribute selection output
                 55 waipori
   0.6364989
   0.5917877
                16 huntly 1 4
   0.5639046
                25 maraetai
   0.5501655
                 9 atiamuri
   0.5495757
                54 waipapa
   0.5411297
                59 whakamaru
   0.5402293
                  6 arapuni
   0.5211691
                31 ohakuri
   0.4692934
                42 stratford
   0.4511059
                20 karapiro
   0.4487579
                53 waikaremoana
   0.428585
                36 patea
   0.3850312
                12 clyde
7 aratiatia
   0.3737265
   0.3554082
   0.3461578
                18 huntly_p40
   0.3285118
                62 whirinaki
   0.3278424
                17 huntly_e3p
   0.311822
                50 tokaanu
                15 highbank
   0.3072184
   0.2956748
                 8 argyle_wairau
   0.2903421
                 2 McKee
   0.2873859
                41 southdown
   0.2631358
                40 roxburgh
                44 te_rapa
   0.2546969
                33 ohau_b
   0.2269401
                61 wheao_flaxy
   0.2258973
                35 paerau
   0.2256527
                34 ohau_c
   0.2211247
                60 whareroa
   0.2203966
                32 ohau_a
   0.2019053
                11 benmore
   0.1783755
                  3 Ngatamariki
   0.1693321
                10 aviemore
   0.1668674
                49 tekapo b
   0.1260248
                26 matahina
  0.0776961
0.0735197
                48 tekapo_a
                23 manapouri
   0.0661864
                  5 aniwhenua
                30 ohaaki
27 mokai
   0.0611252
   0.0481803
   0.0457831
                 4 Te_Mihi
   0.0392446
                47 tehuka
   0.0168689
                38 rangipo
  0.0155849
                57 waitaki
  0.0126155
                39 rotokawa
  -0.0000829
                29 ngawha
21 kawerau_new
  -0.0105619
  -0.0380822
  -0.0443963
                22 kinleith
  -0.0460821
                46 te uku
  -0.0542021
                63 white_hill
                56 wairakei
13 coleridge
  -0.0588582
  -0.0819139
  -0.1197273
                58 west_wind
  -0.1481819
                43 te_apiti
51 twf_12
  -0.1707245
                19 kapuni
52 twf_3
  -0.1757987
  -0.1780692
  -0.1981662
                37 poihipi
  -0.2034758
                45 te_rere_hau
  -0.2179371
                28 nap
 -0.231827
```

• (20 marks) Visualisation is an important aspect of this task. Please illustrate at least one important finding of your work.

Here is the screenshot of Weka visualisation.



We can easily find that there is a straight line across the whole graph. This represents a normal distribution in the two-dimensional graph. And I found this line comes through the whole graph, which represents that all attributes are related to attribute "price". This can approve that the data quality is quite good and it's suitable for being a training set.

2.3 Completion: Developing and testing your machine learning system [50 marks]

- (15 marks) Discuss the initial design of your system Data preparation:
- 1. Select Data: I use weka to help rank all attributes.

I choose to use the Correlation Ranking Filter to rank all attributes and the result is below.

I set the attribute "price" as the supervised attribute then run the function.

We can easily find that the only attribute "ID" takes a very few effect on the supervised attribute "price". So I choose to delete it. But when i want to submit the file, i will add it back. For other attributes, they have more or less influence on "price", so I keep them for training.

- 2. Format data: in this data set, all attributes are numeric, so i decide to keep it original state.
 - 3.Build attribute "price" in test data set. Otherwise the date structure is different between training set and testing set, Weka won't work.

BG	ВН	BI
hirinaki	white_hil	Price
0	10215.2	0
0	22490.5	0
0	0	0
0	12070.13	0
0	24409.3	0
0	9356.141	0
0	273.4061	0
0	7520.01	0
0	24439.86	0
0	962.6453	0
0	17220.11	0
0	2714.621	0

Method Choosing:

5 hours ago by Adam Kong

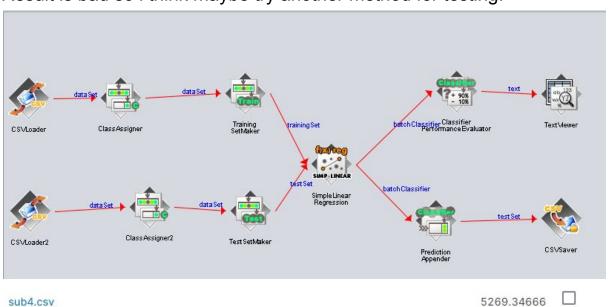
by using SimpleLinearRegression.

I choose Regression because when attempting to determine the strength and character of the relationship between one dependent variable (usually denoted by Y) and a series of other variables (known as independent variables). In this data set, one dependent variable is "price" and others are independent variables.

In my pineline, I choose Linear Regression as a method to training and testing the data set. Here is pipeline:

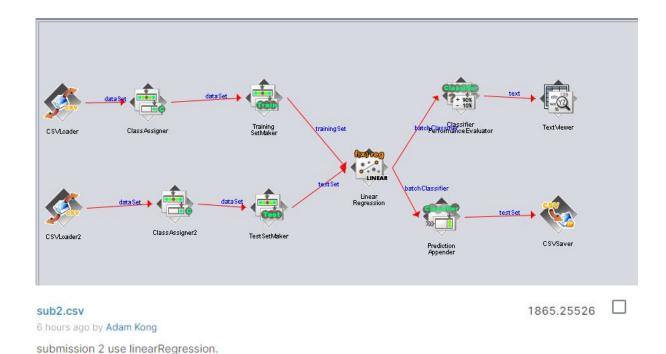
As shown in my pipeline, I didn't make any change to the data set because the data set is perfect which has high quality and good completeness. Each independent attribute has a relation with the dependent attribute("price"). So I use them directly in my pipeline.

Result is bad so i think maybe try another method for testing.



• (15 marks) Discuss the initial design of your system

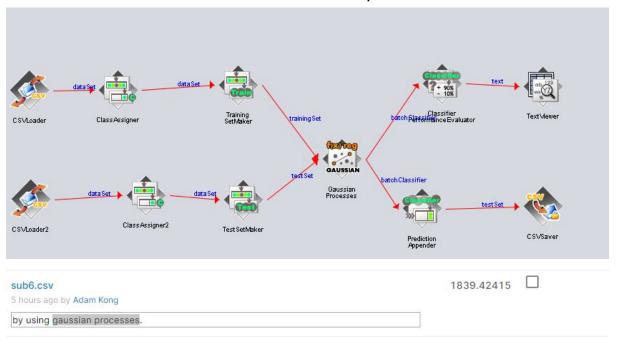
After trying Simple Linear Regression,I realized simple linear regression is suitable for one independent variable and one dependent variable.But in this data set, it contains many independent variables. So I chose Linear Regression as the next method for testing.



And with this pipeline, I got a much better result.

However the result is still not good enough and there are still some methods I didn't use.

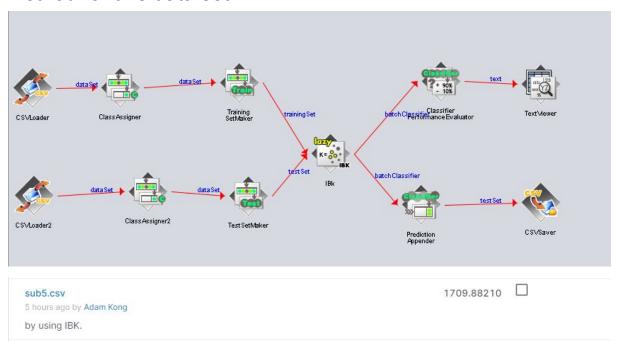
By visualisation, I found the graph has a normal distribution. The linear combination of random variables in the Gaussian process obeys normal distribution, and each finite dimensional distribution is joint normal distribution. So I used the Gaussian process as the next method.



It improved a little bit but almost the same with linear regression.

• (10 marks). Use your judgement to choose the best system you have developed.

Finally, I chose ibk because in this dataset, there are almost no noisy or irrelevant features. So ibk can be a suitable method for this data set.



But for the KNN algorithm, a good k value can make the result more accurate. So I changed the k value many times and got K = 3 as the most suitable value for this data set.



I chose ibk in my final system because it fit the data set best. It can be used for regression and have a k value which can reduce errors in testing. Besides, this data set has an outstanding quality which all attributes are relevant with goal attribute("price") and no missing value. So it could be the perfect method used in my system.

2.4 Challenge: Reflecting on your findings [10 marks]

The model i finally chose is based on ibk, which is easy to explain: For this data set we need to figure out the relationship between "price" and other attributes. Regression is quite suitable to analyze this kind of problem. Ibk can detect most close neighbours with the "price" by k value. It can help calculation reduce errors and noise. Good results come naturally.

Based on my findings, I suggest that the government should promote and support new energy power generation. With the increase of electricity consumption, the demand for electricity is increasing. Promoting new energy power generation (wind power and hydropower) can make the electricity price cheaper and pollution-free. Wind and hydropower in New Zealand are easier to obtain, so they should be promoted to lay a foundation for reducing electricity costs.