# Household Electric Power Consumption

**Cluster Analysis** 



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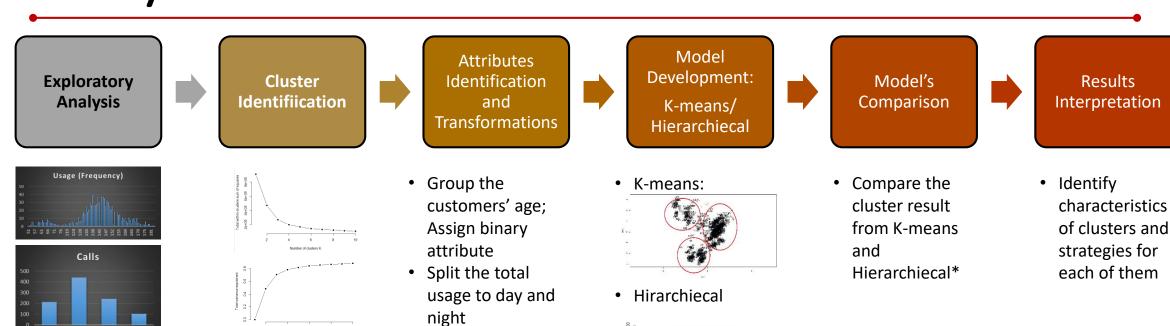








### The methodology that we use to perform the analysis



Standardize the

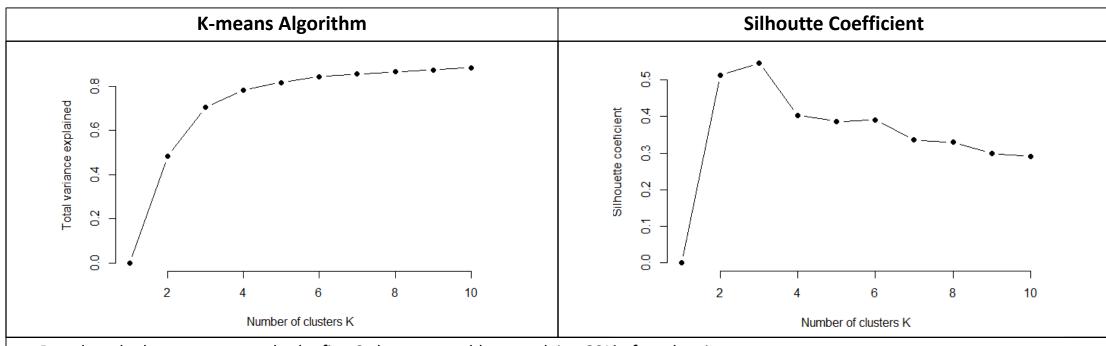
its mean and

data by considering

standard deviation

<sup>\*</sup> K-means is easier to be explained to business people compare to hierarchiecal

## Variance and silhouette coefficient indicate 3 clusters as the most optimal one

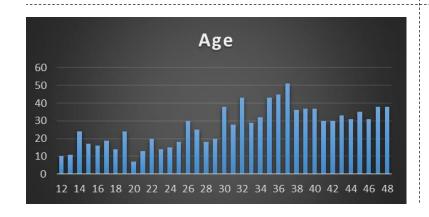


- Based on the k-means approach, the first 3 clusters are able to explain ~80% of total variance
- Moreover, silhouette coefficient also shows 3 clusters has the highest coefficient number
- Given that, the team will use 3 clusters for the analysis

## Pre-processing is the critical process to follow prior to develop the model

#### Assign binary attribute to variable age:

- 2 years or less
- 2 to 3 years
- More than 3 years



#### Complete the data (on the usage):

- Calculate the day usage out of total EPC
- Extract the night usage

Night Usage =
Total Usage - (% Usage x Total Usage)

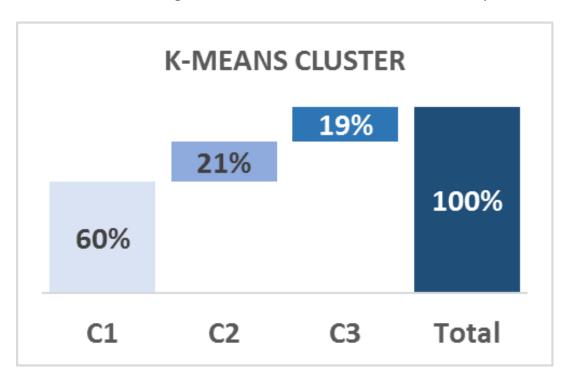
#### Standardize the data:

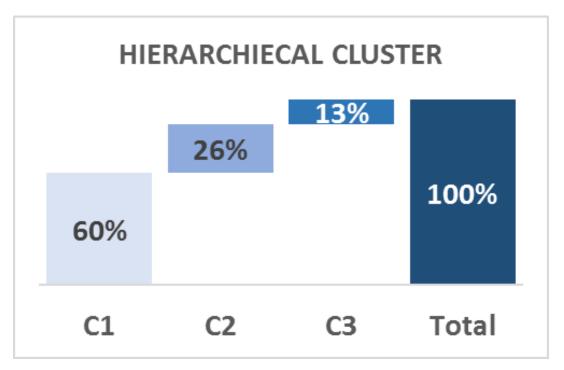
To compare one with another variable

Standardize Value = (Value - Mean)/ Standard Deviation

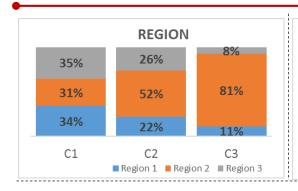
## K-means and hierarchical clustering methodology give similar output

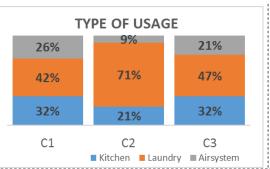
- Based on the clustering methodologies with 3 clusters, the team able to map the distribution of users per cluster
- Both methodologies show that there is a cluster which capture 60% of the users (population)

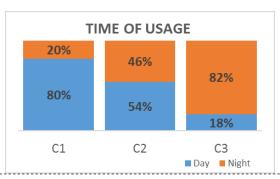


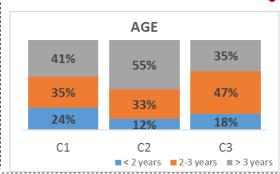


#### Based on the dimensions, we can categorize the clusters to 3 unique characteristics











#### C1 - FAMILY

Housewives who perform cleaning and cooking activities typically on day time

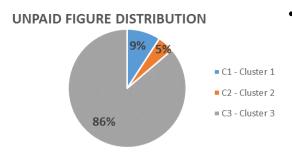
#### **C2 – BOARDING HOUSE**

Boarding house with several rooms rented to tenants, some of whom works on day and some on night

#### **C3 – SINGLE WORKER**

A person who rents or owns a house, but spends most of the day time at work

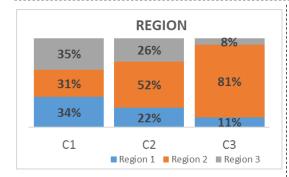
### High level analysis shows that Cluster 3 has a very high unpaid figure



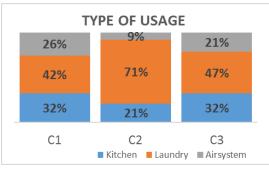
Based on the unpaid figure distribution, we understand that **86%** of unpaid invoices in the last 3 months are belong to Cluster 3; Hence we need to focus on this segment since it will surely impact the company's bottom line



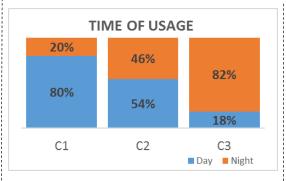
 The average call per customer figure strengthen indication that cluster 3 is the most interesting cluster to analyse, since the company's call center made quite high number of calls per each customer in that segment



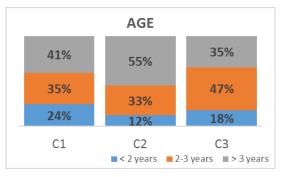
 Unlike C1 that has equal region distribution, C2 and C3 distributed more to Region 2



From the type of usage dimension, we can see that only C2 that has quite different EPC distribution, which heavily consumed on Laundry

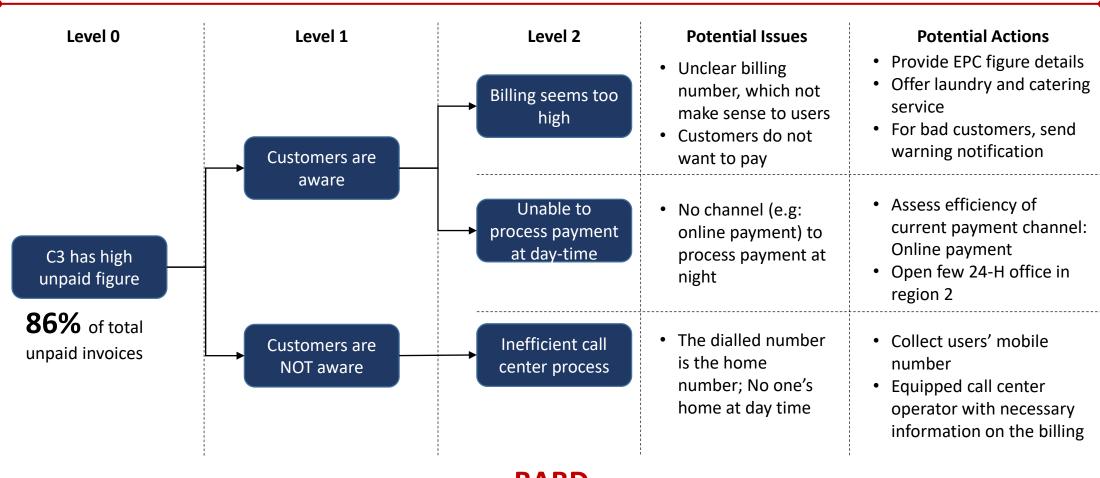


 There is a different time consumption among the clusters, where C1 typically consumes on day while C3 consumes on night

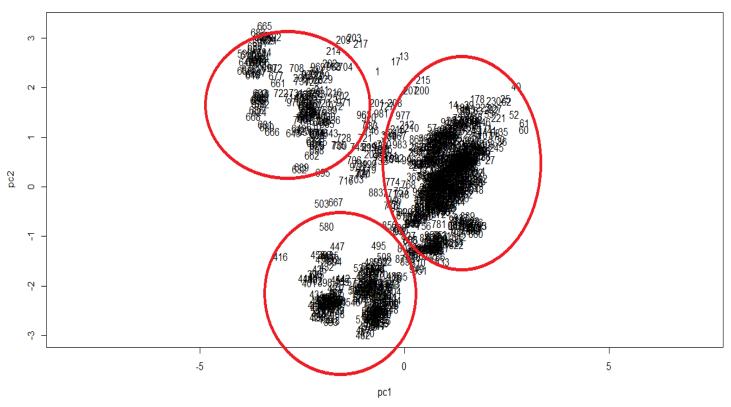


 Simply look at the distribution we can understand the customer's age figure, where across all clusters, old customer capture quite big portion

### Post investigation, the team come up with some potential actions



## Appendix: Principal Component Analysis to map the cluster into two dimensional figure



- Considering that the data have lot of dimensions that probably correlate each other, the team decided to perform Principal Component Analysis
- Based on the PCA, we will able to see the distribution of 3 clusters into two dimensional figure

#### Thank you!

Terima kasih! Merci! 謝謝!