



Project Presentation
on
**Predicting The Reality Of
Mobile Price Range**
CHIA-HAO,HSU

Introduction & Background

Introduction(5W1H)

What issue?

How much budget a person should prepare to buy a mobile phone.

Where?

India

When

The resource from 2016

Who(target audience)

People who wants to purchase a new phone based on their own requirements.

Why

Less knowledge or idea about how to measure the budget.

How

The analysis of classification to predict price range.



The price range

Price range label	class	Range	Example
0	Low cost	Up to 8,000	HUAWEI Y9, Asus Zenfone Max Pro M2
1	Medium cost	8,000~15,000	Sony Xperia 10 II, iPhone SE2
2	High cost	15,000~25,000	iPhone 11, Pixel 5, Samsung Galaxy S20 FE 5G
3	Very high cost	At least 25,000	iPhone 12, ASUS ROG Phone 3

Whole Progresses



1 Introduction to the dataset

2 Data Preparation

3 Data Exploring

4 Regression model

5 Conclusion



Introduction to the Dataset

01

Data Sources

≡ kaggle

- Home
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- Loan Data Set
- IMDB Dataset of 50K ...
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Dataset

Mobile Price Classification

classify mobile price range

Abhishek Sharma • updated 3 years ago (Version 1)

Data Tasks Notebooks (1,382) Discussion (10) Activity Metadata

Download (71 KB) New Notebook

Usability 7.1 Tags business, classification

Description

Context

Bob has started his own mobile company. He wants to give tough fight to big companies like Apple,Samsung etc.

He does not know how to estimate price of mobiles his company creates. In this competitive mobile phone market you cannot simply assume things. To solve this problem he collects sales data of mobile phones of various companies.

Bob wants to find out some relation between features of a mobile phone(eg:- RAM,Internal Memory etc) and its selling price. But he is not so good at Machine Learning. So he needs your help to solve this problem.

In this problem you do not have to predict actual price but a price range indicating how high the price is

Data Explorer

181.89 KB

< train.csv (119.53 KB)

Detail Compact Column

21 of 21 columns

Summary

talk_time # three_g # touch_screen # wifi # price_range

longest time that a single battery charge will last Has 3G or not Has touch screen or not Has wifi or not This is the target variable with value of 0(low cost), 1(medium cost), 2(high cost)

URL:<https://www.kaggle.com/iabhishekofficial/mobile-price-classification?select=train.csv>

Introduction of dataset(1)

Variable	Type	Unit	Description
battery_power	int64	mAh	The volume of battery
blue	int64		Support Bluetooth (Yes:1, No:0)
clock_speed	float64	GHz	The speed of microprocessors
dual_sim	int64		Support dual Sim card (Yes:1, No:0)
fc	int64	Mega pixels	The pixels of front camera
four_g	int64		Supprt 4G (Yes:1, No:0)
Int_memory	flost64	Gigabytes	The volume of internal memory

Introduction of dataset(2)

Variable	Type	Unit	Description
m_dep	float64	cm	The depth of mobile phone 手機厚度
Mobile_wt	float64	g	The weight of mobile phone
n_cores	int64	Cores	The number of cores in processor
pc	int64	Mega pixels	The pixels of primary camera
px_height	int64	Pixels	The height of pixels
Px_wideth	float64	Pixels	The width of pixels
ram	float64	Mega bytes	The volume of random access memory

Introduction of dataset(3)



Response

Variable	Type	Unit	Description
sc_h	float64	cm	The height of screen
sc_w	float64	cm	The width of screen
talk_time	float64	hour	The largest time of phone calling(without extra charging)
three_g	int64		Support 3G (Yes:1, No:0)
touch_screen	int64		With touch screen (Yes:1, No:0)
wifi	int64		Support WIFI (Yes:1, No:0)
price_range	int64		0(low cost), 1(medium cost), 2(high cost) and 3(very high cost).



Data Preparation

02

Detecting and dealing with Null Values

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 2000 entries, 0 to 1999
Data columns (total 21 columns):
 #   Column           Non-Null Count  Dtype  
--- 
 0   battery_power    2000 non-null   int64  
 1   blue              2000 non-null   int64  
 2   clock_speed      1957 non-null   float64
 3   dual_sim          2000 non-null   int64  
 4   fc                2000 non-null   int64  
 5   four_g            2000 non-null   int64  
 6   int_memory        1974 non-null   float64
 7   m_dep              1984 non-null   float64
 8   mobile_wt         1972 non-null   float64
 9   n_cores            2000 non-null   int64  
 10  pc                2000 non-null   int64  
 11  px_height         2000 non-null   int64  
 12  px_width          1969 non-null   float64
 13  ram               1978 non-null   float64
 14  sc_h              1982 non-null   float64
 15  sc_w              1982 non-null   float64
 16  talk_time         1986 non-null   float64
 17  three_g            2000 non-null   int64  
 18  touch_screen      2000 non-null   int64  
 19  wifi               2000 non-null   int64  
 20  price_range        2000 non-null   int64  
dtypes: float64(9), int64(12)
memory usage: 328.2 KB
```



```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 2000 entries, 0 to 1999
Data columns (total 21 columns):
 #   Column           Non-Null Count  Dtype  
--- 
 0   battery_power    2000 non-null   int64  
 1   blue              2000 non-null   int64  
 2   clock_speed      2000 non-null   float64
 3   dual_sim          2000 non-null   int64  
 4   fc                2000 non-null   int64  
 5   four_g            2000 non-null   int64  
 6   int_memory        2000 non-null   float64
 7   m_dep              2000 non-null   float64
 8   mobile_wt         2000 non-null   float64
 9   n_cores            2000 non-null   int64  
 10  pc                2000 non-null   int64  
 11  px_height         2000 non-null   int64  
 12  px_width          2000 non-null   float64
 13  ram               2000 non-null   float64
 14  sc_h              2000 non-null   float64
 15  sc_w              2000 non-null   float64
 16  talk_time         2000 non-null   float64
 17  three_g            2000 non-null   int64  
 18  touch_screen      2000 non-null   int64  
 19  wifi               2000 non-null   int64  
 20  price_range        2000 non-null   object 
dtypes: float64(9), int64(11), object(1)
memory usage: 328.2+ KB
```

Detecting and dealing with Null Values

A	B	C
10	100	1000
20	200	2000
30	300	3000
N/A	400	4000
40	N/A	5000
50	500	N/A
60	600	6000

```
df['A']= df['A'].fillna( round( df ['A'].mean() ) )
```

```
df['B']= df['B'].fillna( round( df ['B'].mean() ) )
```

```
df['C']= df['C'].fillna( round( df ['C'].mean() ) )
```



When the scale is large, filling null value with mean good at maintaining the distribution in each variable.

A	B	C
10	60	300
20	120	600
30	180	900
35	210	1200
40	190	1500
50	270	1100
60	300	2100



Data Exploring

03

Heat Map Of All Variables

From the correlation coefficient:

$$r (fc, pc) = 0.64$$

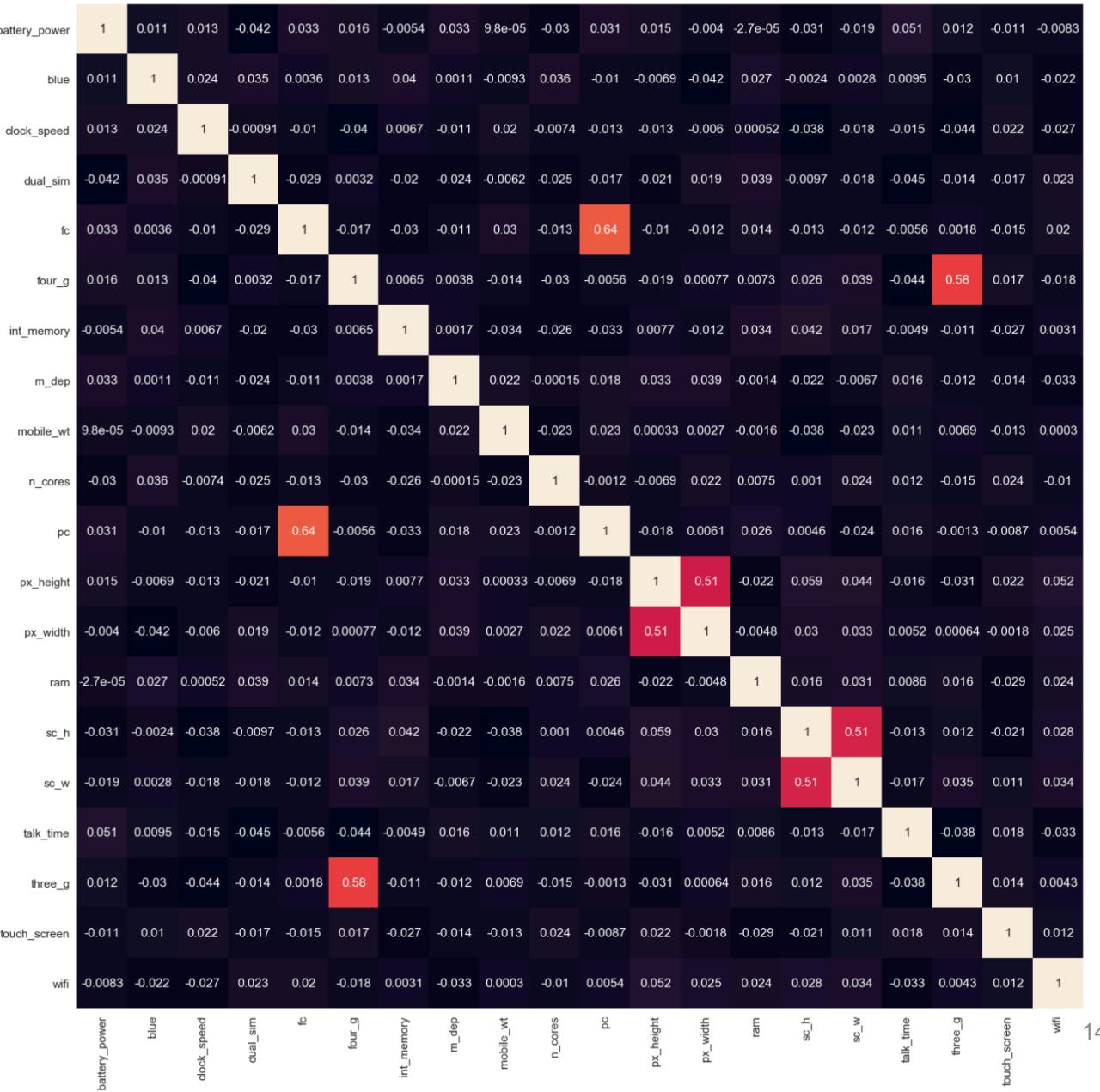
$$r (\text{three_g}, \text{four_g}) = 0.58$$

$$r (\text{px_h}, \text{px_width}) = 0.51$$

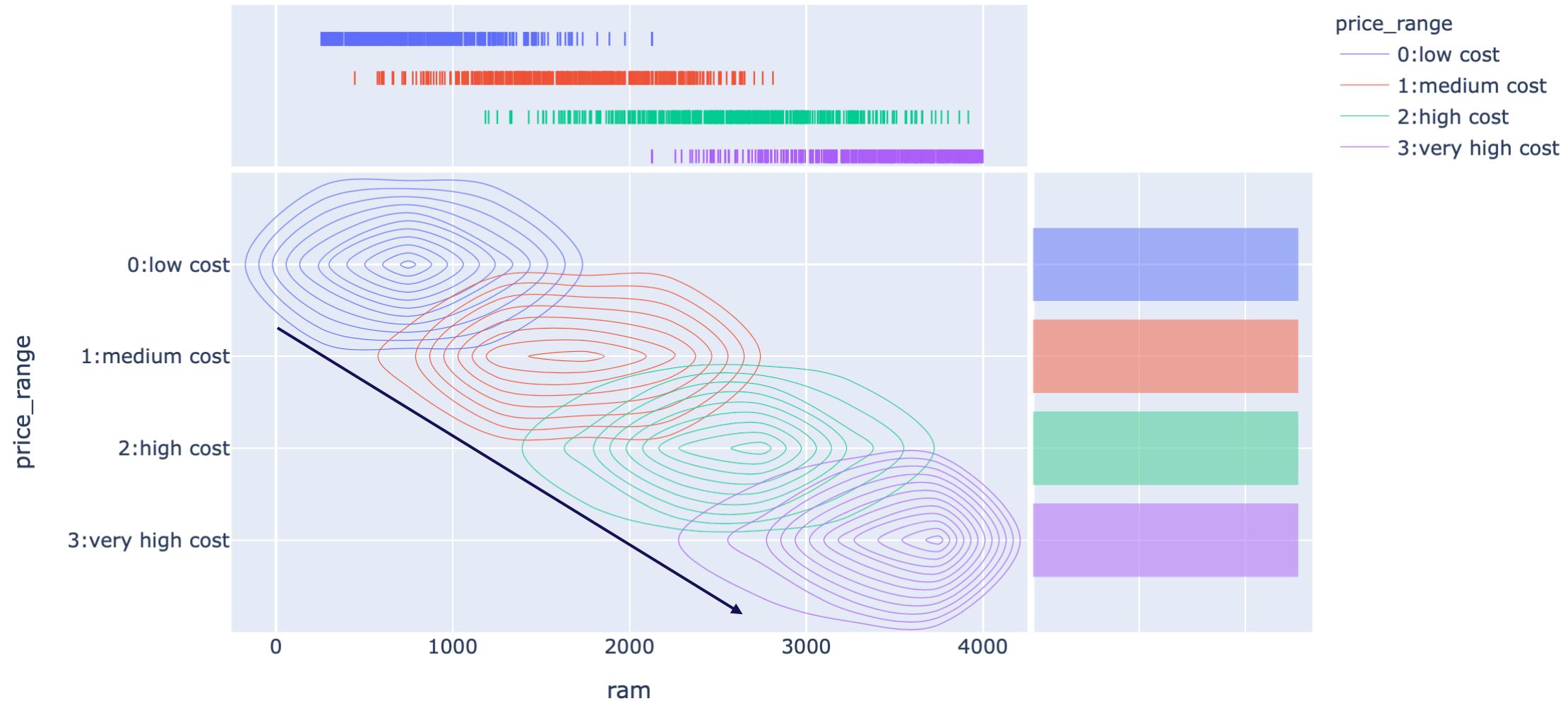
$$r (\text{sc_w}, \text{sc_h}) = 0.51$$

If $r > 0.7$, it has collinearity issue,
we must remove one of variable in the
combination because both have same feature in
the dataset.

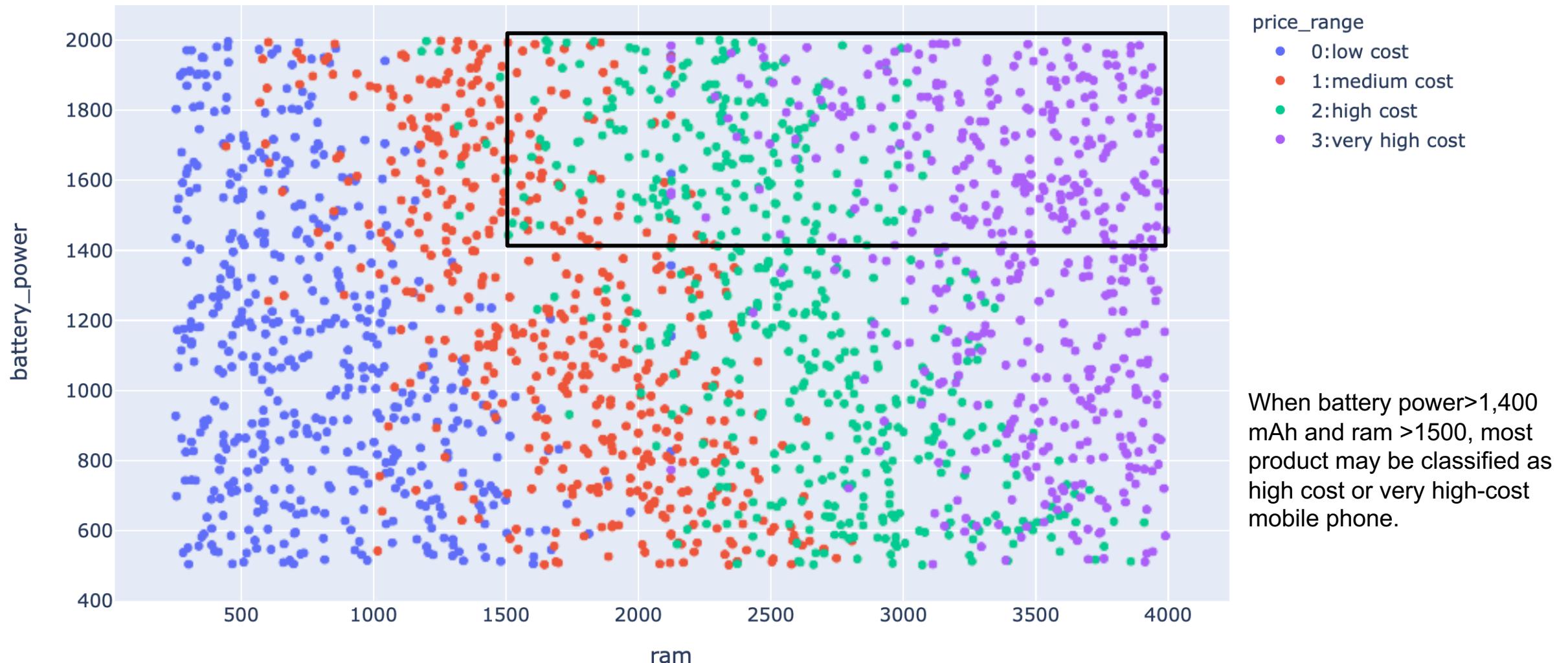
In the heat map,
there is no collinearity issue,
so we **do not need to remove any variable**.



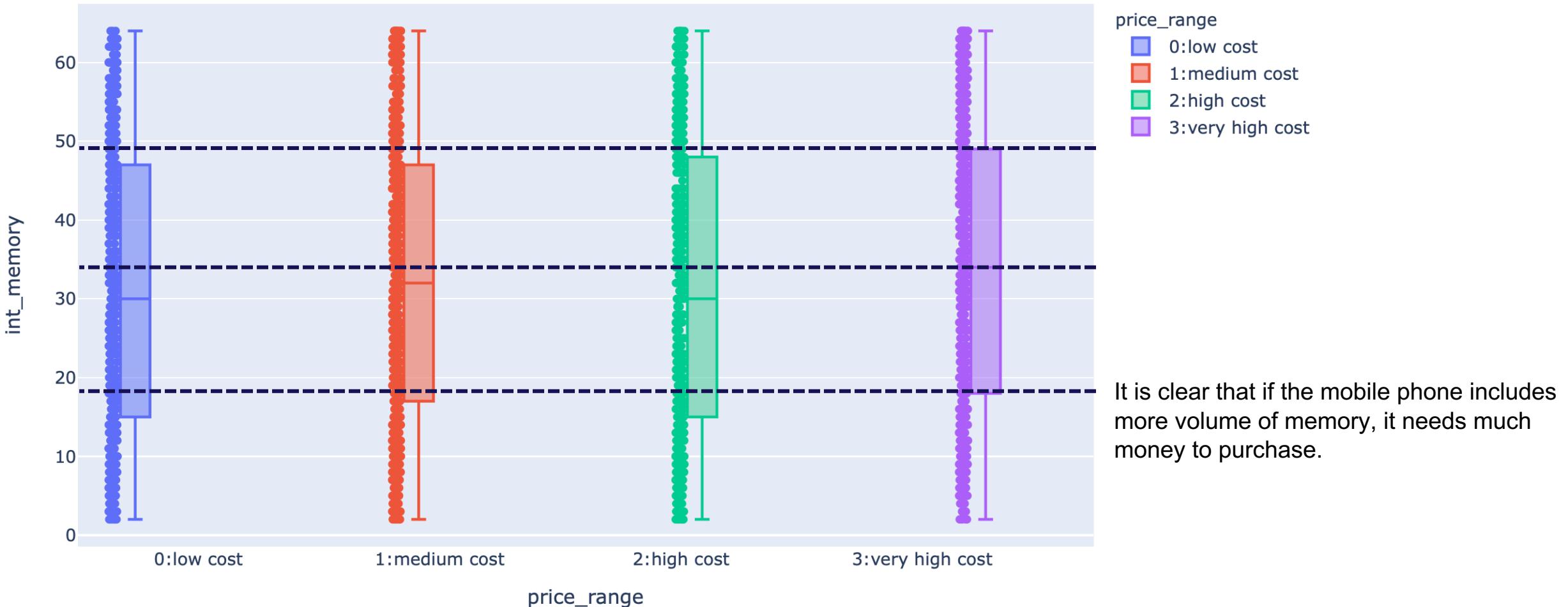
Rule 1- More volume of ram, the much price that customers should pay.



Rule 2- High battery power & ram spend much money to have.

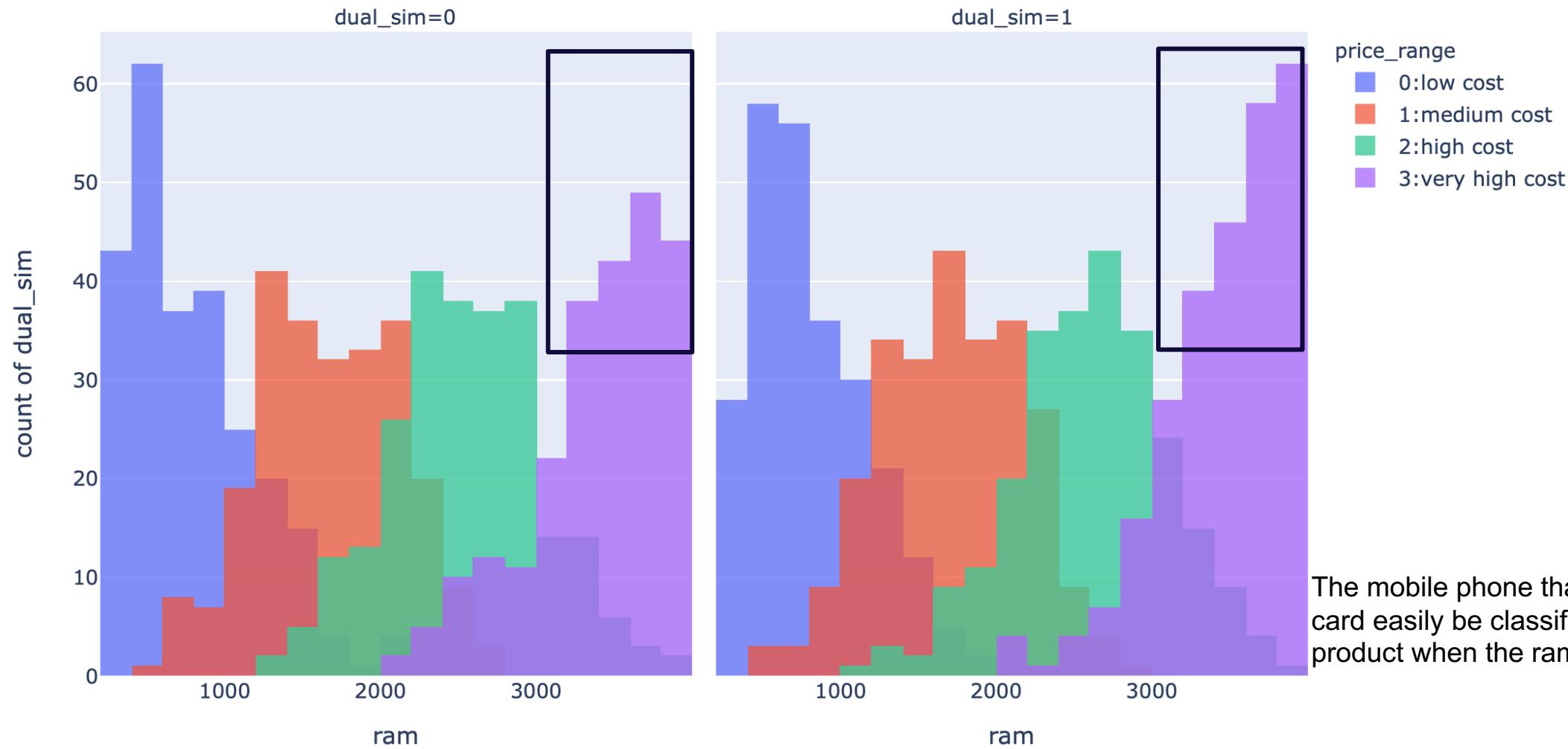


Rule 3- More volume of memory, much budget to prepare.



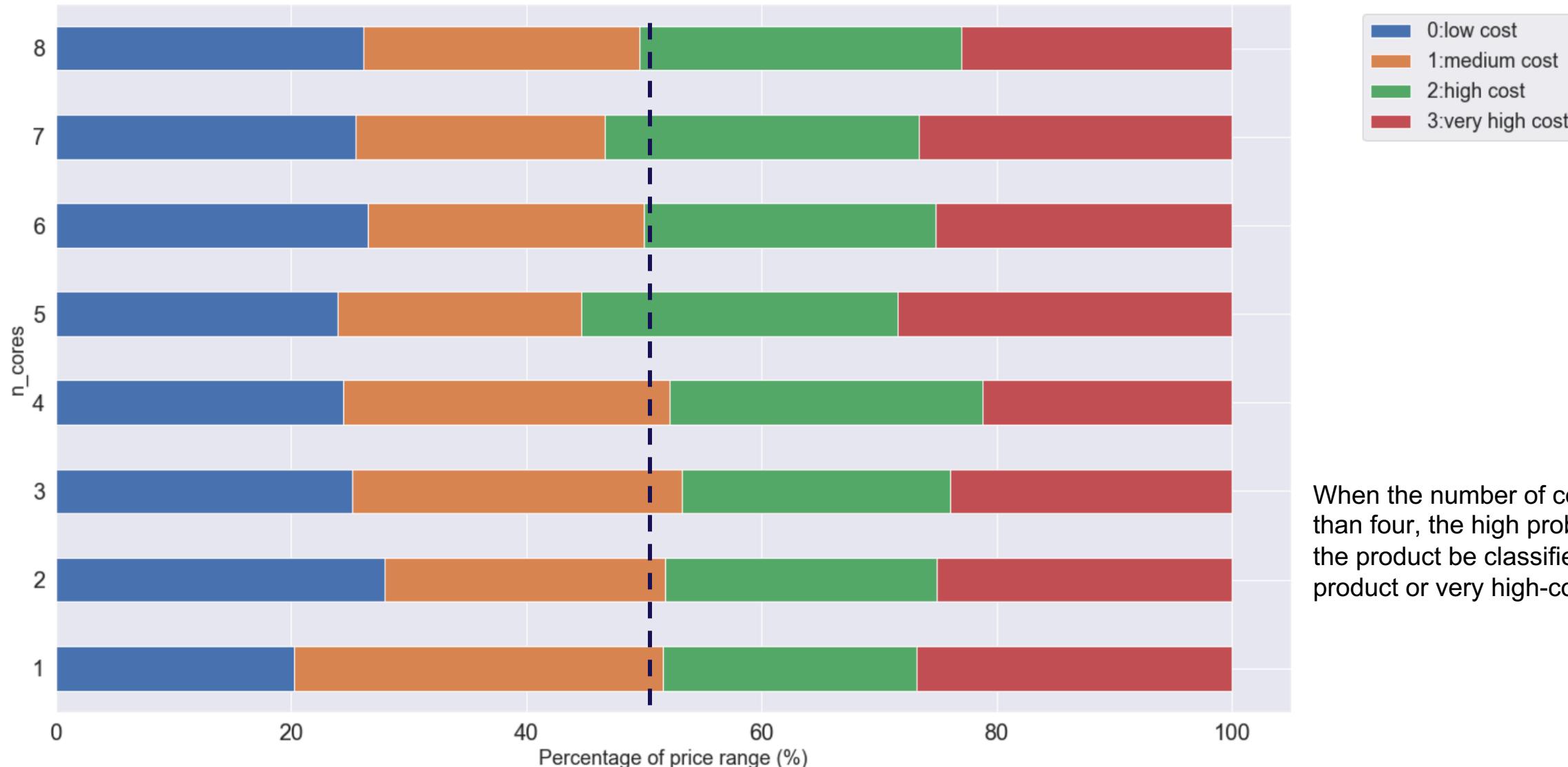
Rule 4-

In high volume of ram, supporting dual sim card, the product may increase the price.



The mobile phone that supporting dual sim card easily be classified as very high-cost product when the ram is higher than 3,000.

Rule 5- More number of cores, the higher price for the mobile phone.





Classification Model

04

Regression Model

Predictors



⋮

Battery Power



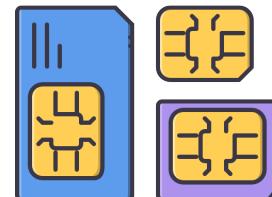
⋮

Blue tooth



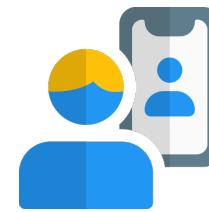
⋮

Clock speed



⋮

Dual SIM card



⋮

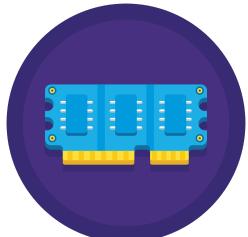
Pixels of
Front camera

Regression Model

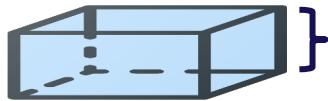
Predictors



4G



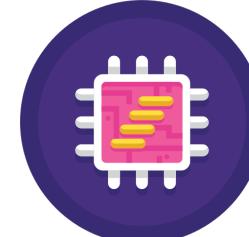
Volume of memory



Mobile depth



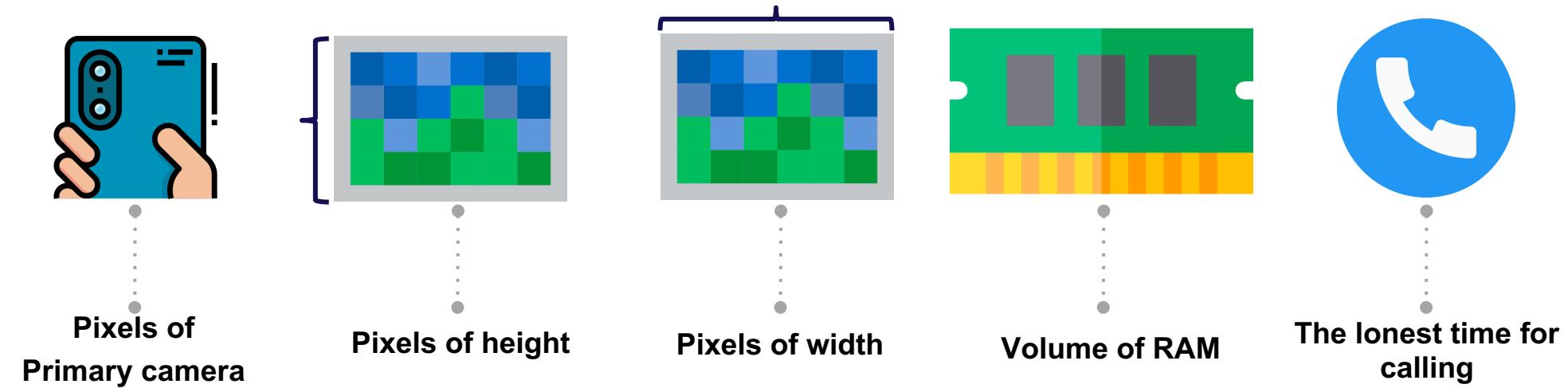
Mobile weight



Number of cores

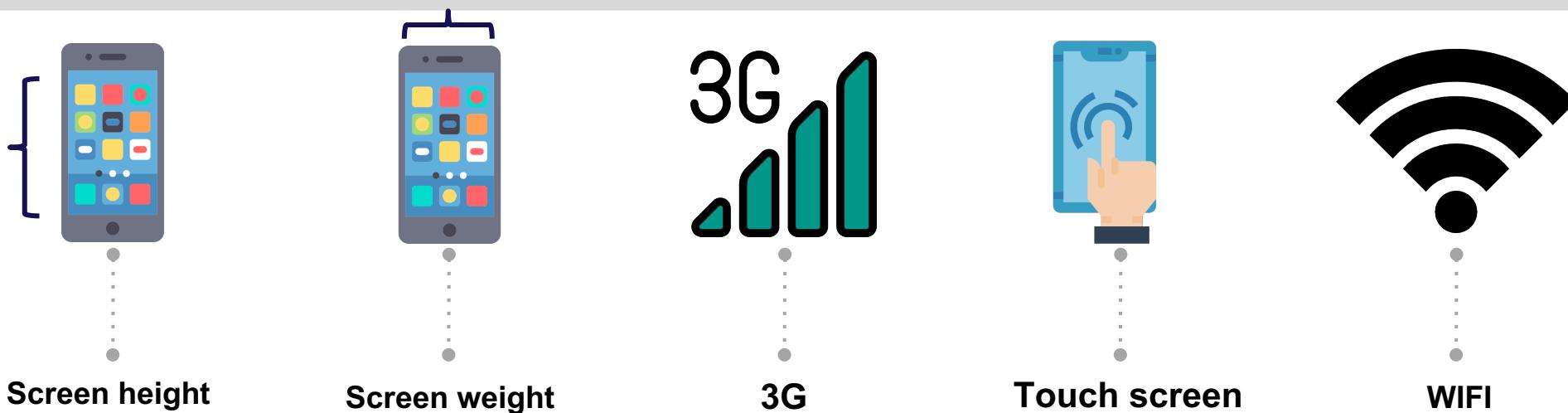
Regression Model

Predictors



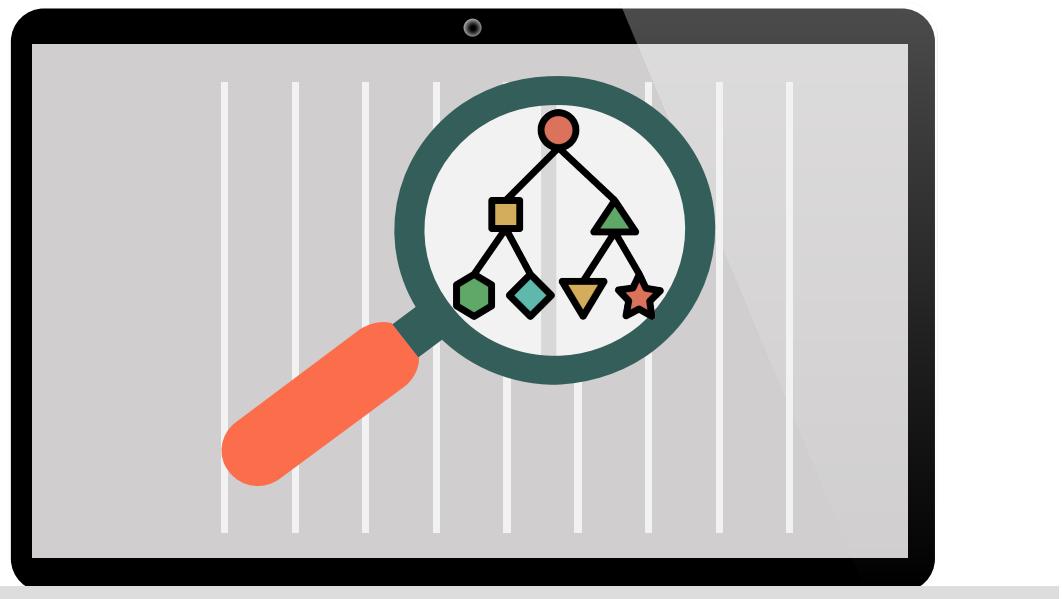
Regression Model

Predictors



Classification Model

Response –Price range



01

DecisionTreeClassifier (Decision Tree)

02

SVC (Support Vector Machines)

03

KNeighborsClassifier (Nearest Neighbors)

04

LogisticRegression

05

RandomForestClassifier (Ensemble Learning)

06

XGBClassifier (Ensemble Learning)

Classification Model

Split the dataset

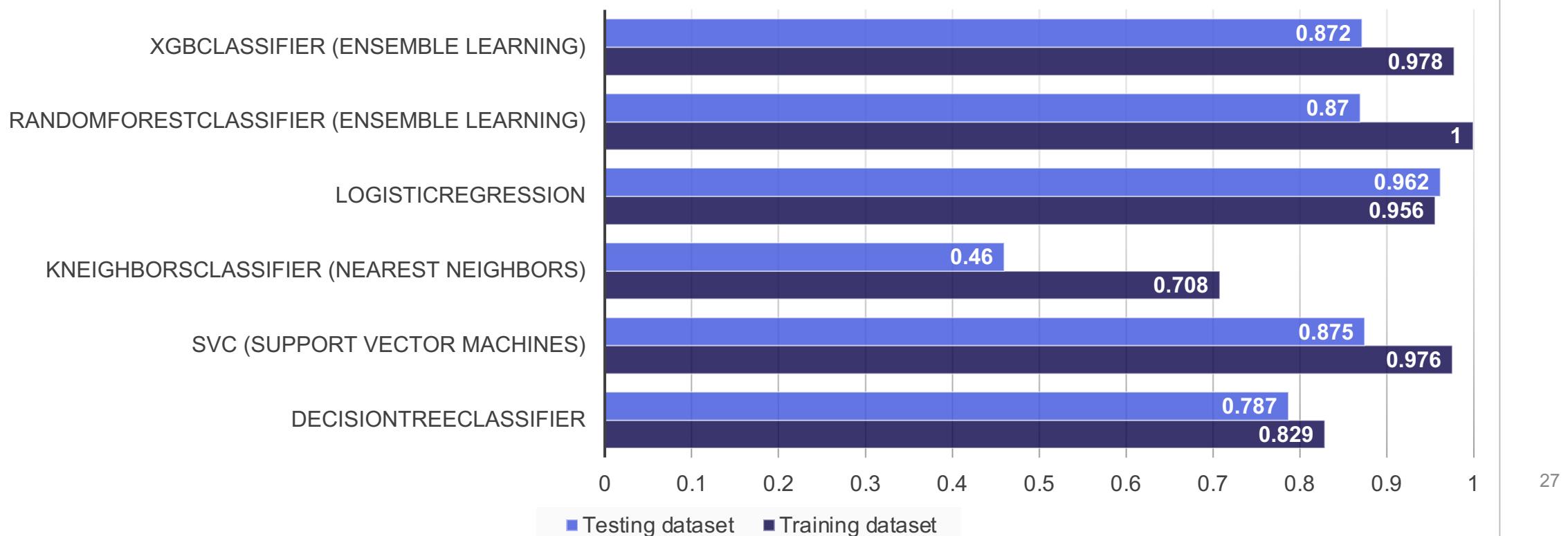
Training dataset : 80%, Testing dataset : 20%.



Classification Model

Accuracy

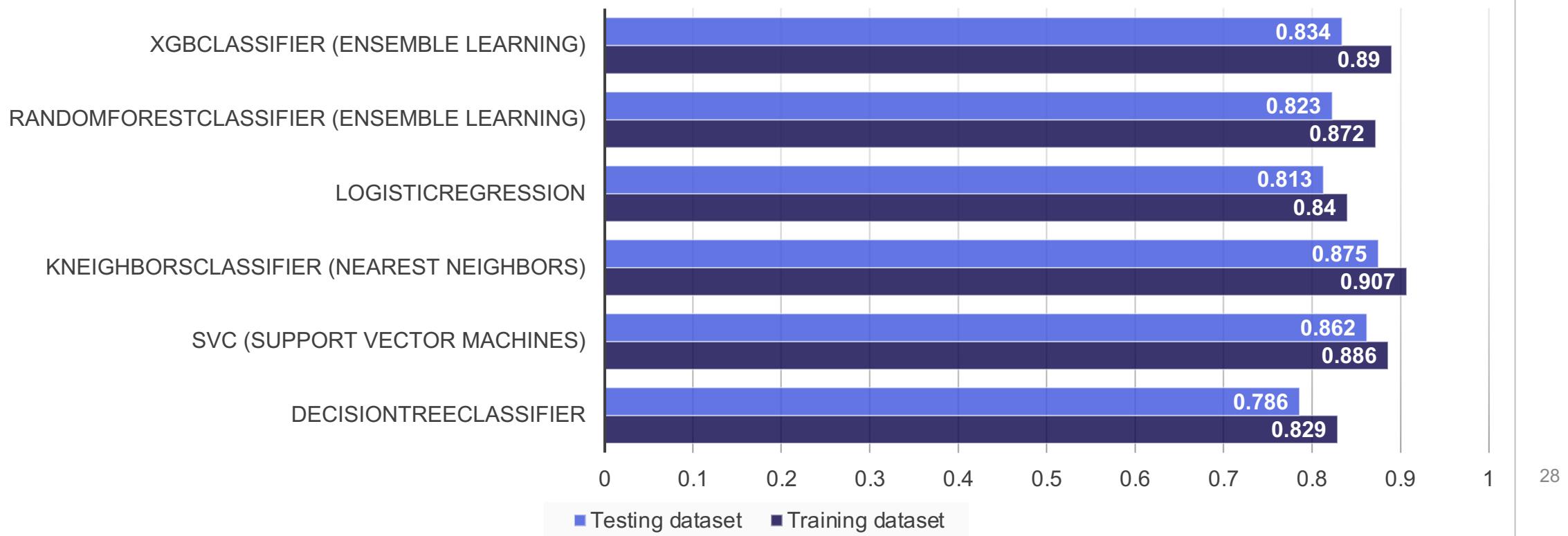
Accuracy



Regression Model-Cross validation

Accuracy

Accuracy





Conclusion

05

Conclusion

Flow Chart



What issue?

We cannot know how much budget we should prepare.



Measuring the budget

How much money to buy a mobile phone.

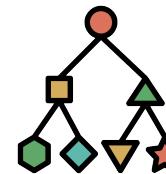
Import recent dataset



Visualizing data

Find some interesting features in our dataset.

Present on reality



Predicting price range

Accuracy 87.5%

We have 87.5% of the observed variation can be explained in this model.



**Thanks for your
watching and listening.**

CHIA-HAO,HSU