

## **Capstone RECAP**

► Use Mobile Price Classification dataset to build models to predict the price range indicating how high the price is of mobile phone base on mobile specifications

Data: Mobile Price Classification



## **Capstone RECAP**

- Mobile phone price change depending on the specifications of the phone, but there are many special examples. Some phones' specifications not as good as other phones, but the price is not lower than other phones.
- Random access memory in megabytes and battery power have a greater impact on the price.



## **Capstone RECAP**

Models:

Nature of the Problem: Multiclass classification problem

Logistic Regression: Accuracy for the training set is 86.75%, for the test set is 83.25%

Neural Network: Accuracy for the training set is 100%, for the test set is 92%

Random Forest: Accuracy for the training set is 98.75%, for the test set is 86.75%



#### Work Extension

- A problem is that the price data we obtained is not a definite value range.
- This makes our data and results less usable.
- Because over time, the price of mobile phones is changing. Even with the same configuration, the price in 2010 and the price in 2020 will be very different.
- I proposed a method to simulate a specific price range.



#### **Obtain Price Data**

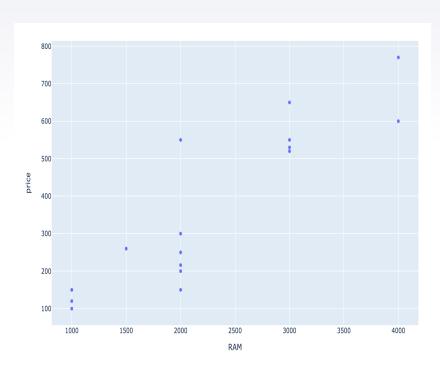
- Source: Wikipedia
- Randomly select 16 phones for the list and google the release price of that year and the RAM(in megabytes).
- I retrieved the Data of the Year 2015 and 2016

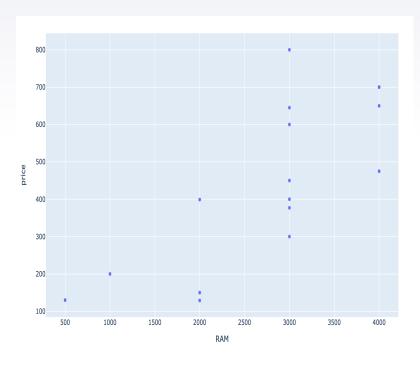
	Name	RAM	price
0	Acer Liquid Z630	2000	216
1	Alcatel One Touch Idol 3	2000	250
2	Droid Turbo 2	3000	520
3	Honor 5X	2000	200
4	HTC Butterfly 3	3000	530
5	iPhone 6S	2000	550
6	Lava Pixel V1	2000	150
7	Lenovo A6000	1000	150
8	LG G4	3000	550
9	Microsoft Lumia 430	1000	120

	Name	RAM	price
0	Alcatel Idol 4	3000	377
1	BlackBerry DTEK50	3000	300
2	Cat S60	3000	600
3	HP Elite x3	4000	700
4	iPhone SE	2000	399
5	Pixel	4000	650
6	Samsung Galaxy A8	3000	800
7	Sony Xperia X	3000	645
8	LG G5	4000	700
9	Redmi 3	2000	150



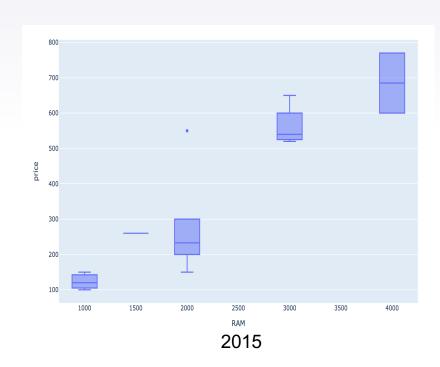
## Scatter Plot the Data

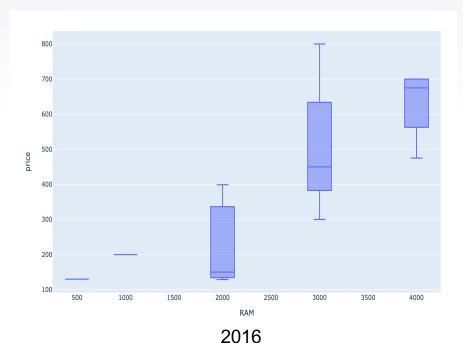




2015 2016

# Box plot the Data





#### Simulation

- Based on the Box Plot, people can easily separate four price ranges.
- We also can use first quartile(Q1) and third quartile(Q3) to decide the demarcation point.

Price Tag	2015	2016
Low	<200	<200
Median	200 ~ 500	200 ~ 350
High	500 ~ 600	350 ~ 600
Very High	600<	600<



# Other Finding

- Price can also be affected by Brand and Country.
- For example:

Freedom 251 sale in India

**Price**: ₹251(the equivalent of \$3.54 as of 2020)



Compatible 2.5G, 3G, HSUPA

networks

First February 18, 2016; 4 years ago

released

Availability India

by region

Type Smartphone
Form factor Slate

**Mass** 130 g

Operating system

CPU 1.3 GHz Quad-Core Processor

Android 5.1 Lollipop

Memory 1 GB RAM

Storage 8 GB

Removable Up to 32 GB MicroSD

storage

Battery 1450 mAh Lithium-ion battery

Data inputs Touchscreen, Accelerometer,

Magnetometer, Proximity sensor, Ambient light sensor

lay 102 mm (4.0 in) qHD IPS

Display (960x540 pixels)

Rear camera 3.2 megapixel

Front camera 0.3 megapixel

Connectivity Wi-Fi, Bluetooth 3.0, USB 2.0,

DLNA. UMA

Other Wi-Fi Hot Spot, FM radio





Xiaomi Mi 2 \$300





iPhone 5 \$650

V.S.

# Future Work

Use Web scraping technique to obtain the data with other important features, like brand, country, new design to make my models more accuracy under actual conditions



# THANKS!

#### **Any questions?**

You can find me at:

- siyuw1@umbc.edu
- https://github.com/ciciwang1/DAT A606





### Credits

- Presentation template by Slide Carnival
- Dataset by Abhishek Sharma

