DATS 6202

Machine Learning I



Team 11: Mobile Price Classification

Data Science, Columbian College of Arts & Sciences

Problem Statement

 Bob has started his own mobile company and wants to compete with big players like Apple and Samsung. However, he's unsure how to price his mobile phones competitively. To tackle this, he gathers sales data from various companies to understand the relationship between different features (e.g., RAM, internal memory) and selling prices. Bob needs assistance in analyzing this data and building a model to predict price ranges for his mobile phones based on their specifications. By doing so, Bob aims to make informed pricing decisions and establish his company in the competitive mobile phone market.



Solution

Our aim is to predict the price range of mobile phones based on their specifications. We'll explore different classification models to find the one that best predicts phone prices accurately. By comparing models using metrics like accuracy, we'll identify the most effective one for this task.



Implementation

The Pipeline for Regression and Classification:

- Loading the data
- Splitting the data
- Handling uncommon features
- Handling identifiers
- Handling date time variables

- Handling missing data
- Encoding the data
- Splitting the feature and target
- Scaling the data
- Feature selection & feature engineering



Data Description

• 21 columns

- √ battery_power: The total energy a battery can store in milliampere-hours (mAh).
- ✓ blue: Indicates the presence of Bluetooth (1 if present, 0 if absent).
- ✓ clock_speed: The speed at which the microprocessor executes instructions, measured in gigahertz (GHz).
- √ dual_sim: Indicates whether the device supports dual SIM cards (1 if supported, 0 if not supported).
- √ fc: Front Camera megapixels.
- √ four_g: Indicates the presence of 4G connectivity (1 if present, 0 if absent).
- ✓ int_memory: Internal Memory capacity in gigabytes (GB).
- √ m_dep: Mobile Depth in cm.
- ✓ **mobile_wt**: Weight of mobile phone in grams.
- ✓ n_cores: Number of cores of the processor.



Data Description

• 21 columns

- ✓ pc: Primary Camera megapixels.
- ✓ px_height: Pixel Resolution Height.
- ✓ px_width: Pixel Resolution Width.
- √ ram: Random Access Memory capacity in megabytes (MB).
- ✓ sc_h: Screen Height of mobile in cm.
- ✓ sc_w: Screen Width of mobile in cm.
- √ talk_time: The longest time that a single battery charge will last when you are on a call.
- √ three_g: Indicates the presence of 3G connectivity (1 if present, 0 if absent).
- ✓ touch_screen: Indicates whether the device has a touch screen (1 if present, 0 if absent).
- ✓ wifi: Indicates whether the device supports Wi-Fi (1 if supported, 0 if not supported).
- ✓ price_range: This is the target variable with values indicating the price range of the mobile phone.



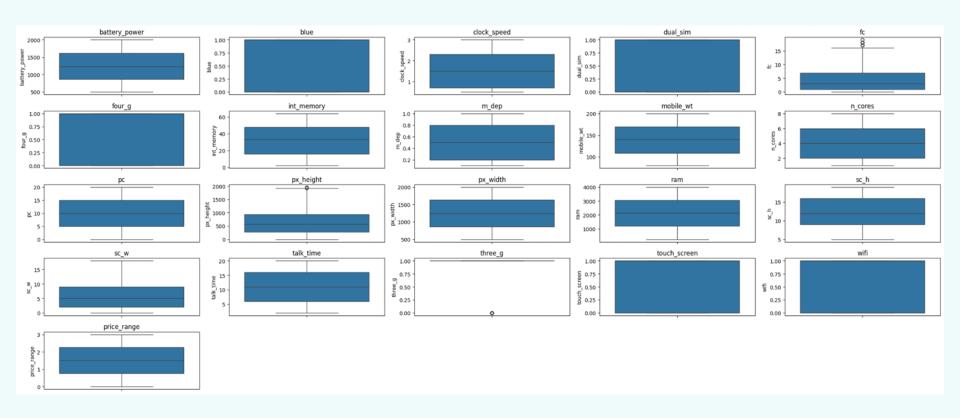
Data Description

• 21 columns

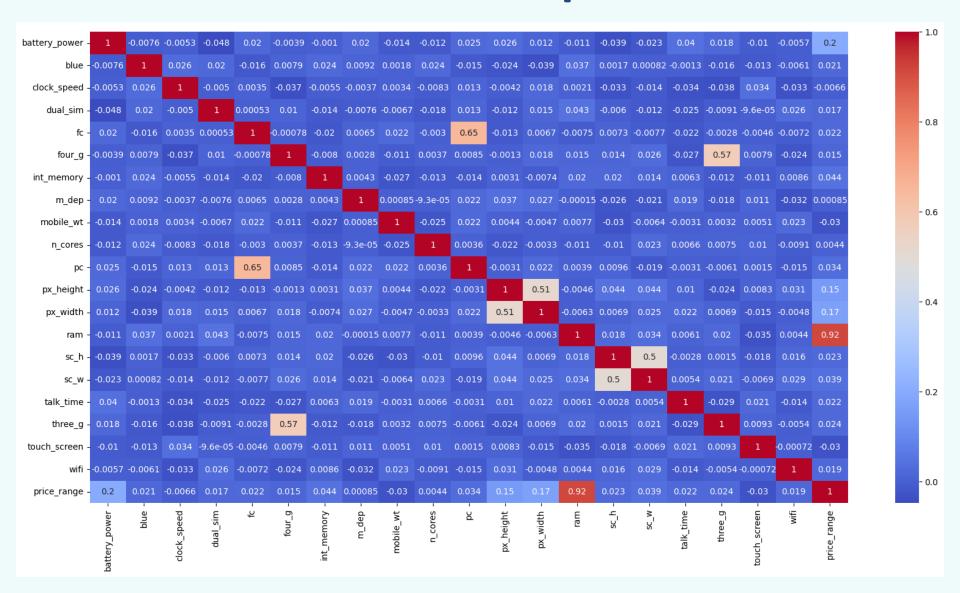
- ✓ pc: Primary Camera megapixels.
- ✓ px_height: Pixel Resolution Height.
- ✓ px_width: Pixel Resolution Width.
- √ ram: Random Access Memory capacity in megabytes (MB).
- ✓ sc_h: Screen Height of mobile in cm.
- ✓ **sc_w**: Screen Width of mobile in cm.
- √ talk_time: The longest time that a single battery charge will last when you are on a call.
- √ three_g: Indicates the presence of 3G connectivity (1 if present, 0 if absent).
- √ touch_screen: Indicates whether the device has a touch screen (1 if present, 0 if absent).
- ✓ wifi: Indicates whether the device supports Wi-Fi (1 if supported, 0 if not supported).
- ✓ price_range: This is the target variable with values indicating the price range of the mobile phone.



EDA



Heat Map





ML Algorithms

- 1. Logistic Regression
- 2. MLPClassifier
- 3. RandomForestClassifier
- 4. Hist GradientBoostingClassifier

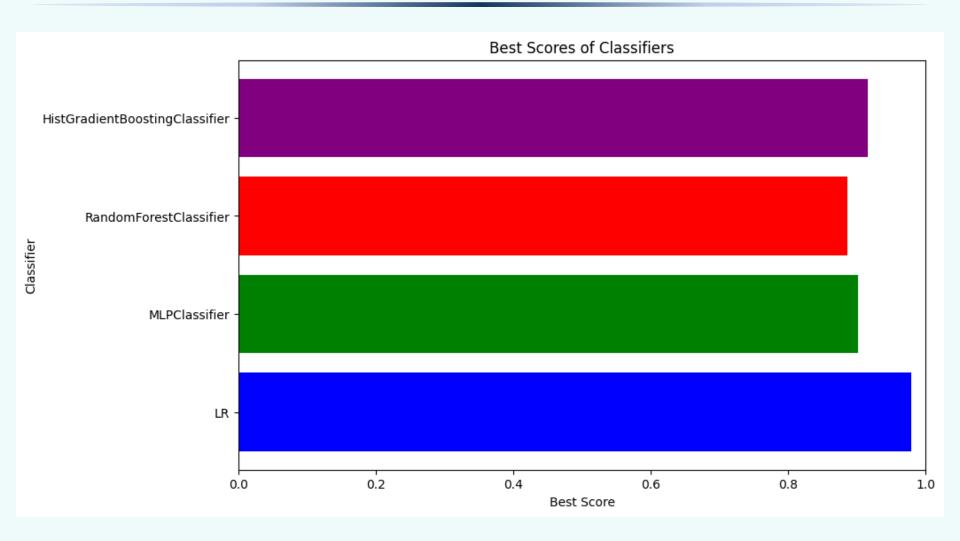


ML Algorithms - Best Score

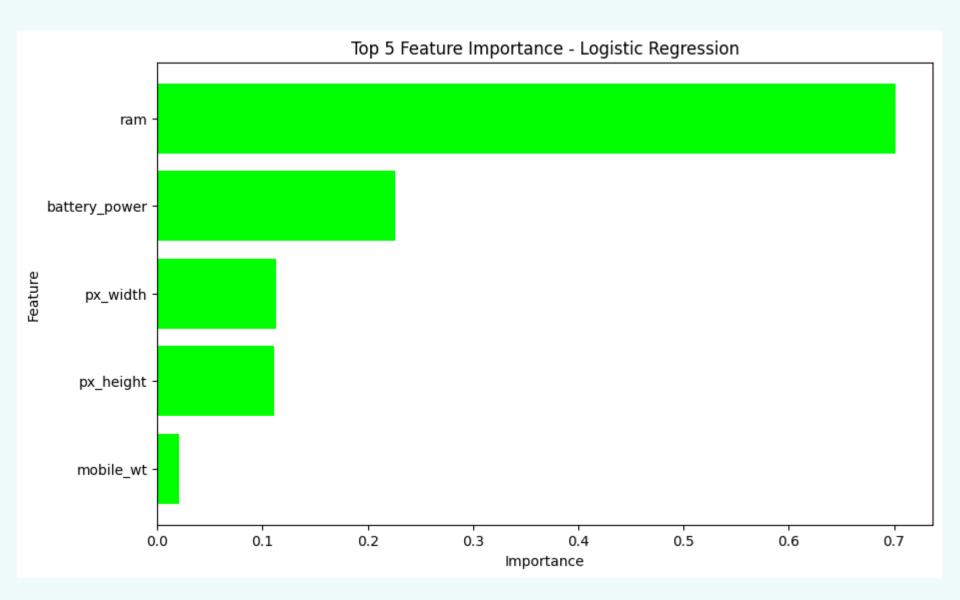
- 1. Logistic Regression **0.979462**
- 2. MLPClassifier 0.901794
- 3. RandomForestClassifier 0.886373
- 4. Hist GradientBoostingClassifier 0.915545



ML Algorithms - Best Score



Feature Importance



Multivariate Visualization

	ram	battery_power	px_height	px_width	mobile_wt
price_range					
0	785.314	1116.902	536.408	1150.270	140.552
1	1679.490	1228.868	666.892	1251.908	140.510
2	2582.816	1228.320	632.284	1234.046	143.614
3	3449.232	1379.984	744.848	1369.838	136.320
4	2138.998	1248.510	627.121	1239.774	139.511



Conclusion

After evaluating various machine learning classification models for predicting phone price ranges based on specifications, we found that logistic regression emerged as the best fit for this task. With an accuracy of **0.979462**, logistic regression outperformed other models in accurately classifying phone prices. Therefore, we recommend utilizing logistic regression for predicting phone price ranges in future applications.



References

- 1. yuxiaohuang/teaching/p2 c2 s5 tree_based_models/case_study
- 2. Kaggle-Mobile Price Classification
- 3. Project layout Referred to table 3 on page 20 of the lecture slides.



Thank You

_

