

Mobile_price_classification_with_ML

December 23, 2025

1 Mobile Price Classification with Machine Learning

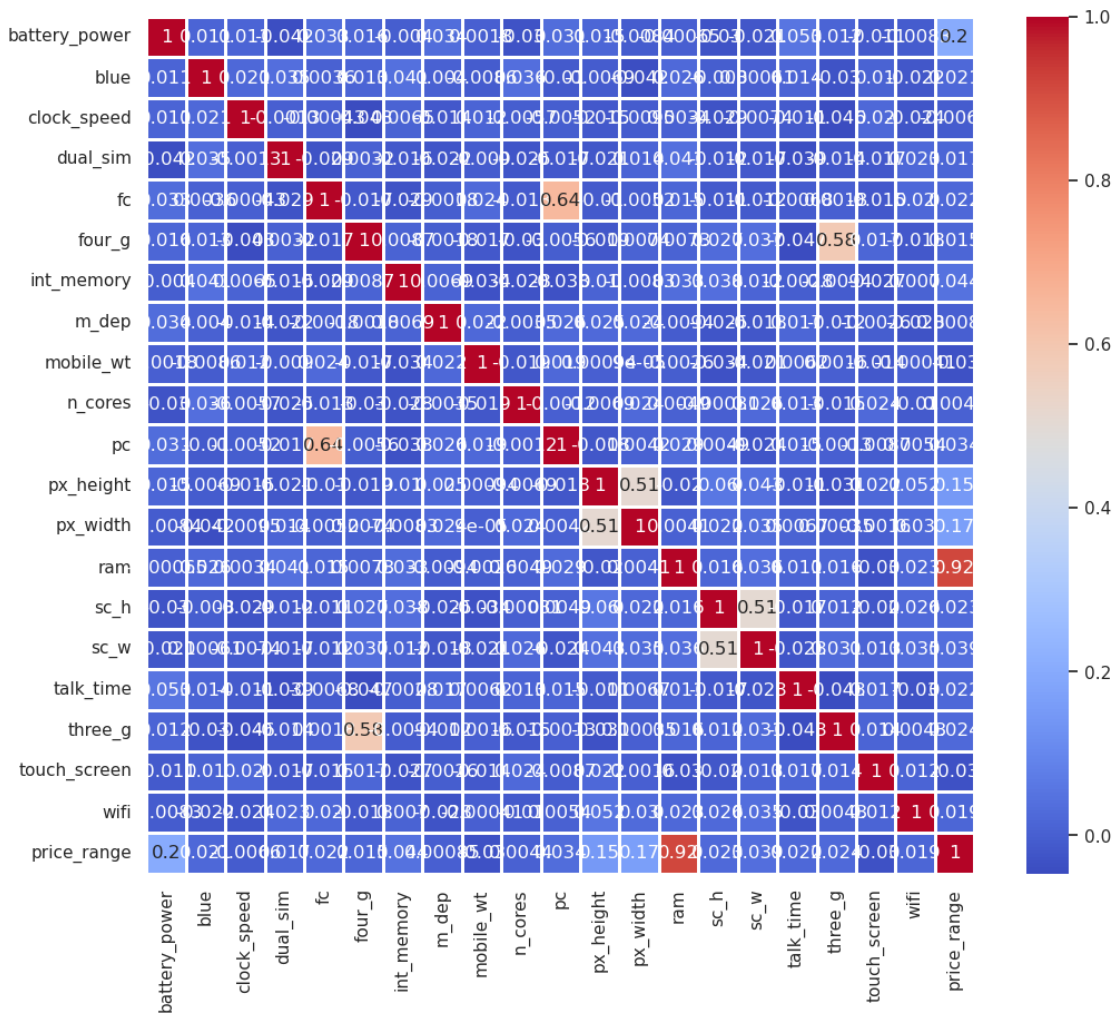
```
[2]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
from sklearn.linear_model import LogisticRegression
from sklearn.metrics import accuracy_score
sns.set()
```

```
[7]: data = pd.read_csv("mobile_prices.csv")
print(data.head())
```

	battery_power	blue	clock_speed	dual_sim	fc	four_g	int_memory	m_dep	
mobile_wt	n_cores	pc	px_height	px_width	ram	sc_h	sc_w	talk_time	
three_g	touch_screen	wifi	price_range						
0	842	0	2.2	0	1	0	7	0.6	
188	2	2	20	756	2549	9	7	19	0
0	1	1							
1	1021	1	0.5	1	0	1	53	0.7	
136	3	6	905	1988	2631	17	3	7	1
1	0	2							
2	563	1	0.5	1	2	1	41	0.9	
145	5	6	1263	1716	2603	11	2	9	1
1	0	2							
3	615	1	2.5	0	0	0	10	0.8	
131	6	9	1216	1786	2769	16	8	11	1
0	0	2							
4	1821	1	1.2	0	13	1	44	0.6	
141	2	14	1208	1212	1411	8	2	15	1
1	0	1							

```
[8]: plt.figure(figsize=(12,10))
sns.heatmap(data.corr(), annot=True,
            cmap="coolwarm",linecolor="white",linewidths=1)
```

[8]: <Axes: >



1.1 Data Preparation

```
[10]: x = data.iloc[:, :-1].values
      y = data.iloc[:, -1].values
      x = StandardScaler().fit_transform(x)
      x_train, x_test, y_train, y_test = train_test_split(x, y, test_size=0.20,
      ↪random_state=0)
```

1.2 Mobile Price Classification Model

```
[14]: from sklearn.linear_model import LogisticRegression
      lreg = LogisticRegression()
      lreg.fit(x_train, y_train)
```

```
y_pred = lreg.predict(x_test)
```

```
[16]: accuracy = accuracy_score(y_test, y_pred) * 100  
print("Accuracy of the Logistics Regression Model: ",accuracy)
```

Accuracy of the Logistics Regression Model: 95.5

```
[17]: print(y_pred)
```

```
[3 0 2 2 3 0 0 3 3 1 1 3 0 2 3 0 3 2 2 1 0 0 3 1 2 2 3 1 3 1 1 0 2 0 2 3 0  
0 3 3 3 1 3 3 1 3 0 1 3 1 1 3 0 3 0 2 2 2 0 3 3 1 3 2 1 2 3 2 2 2 3 2 1 0  
1 3 2 2 1 2 3 3 3 0 0 0 2 1 2 3 1 2 2 1 0 3 3 3 0 3 1 1 3 1 3 2 2 3 2 3 3  
0 0 1 3 3 0 0 1 0 0 3 2 2 1 2 1 1 0 2 1 3 3 3 3 3 2 0 1 1 2 1 3 0 3 0 0  
2 0 1 1 1 1 3 0 0 3 1 3 2 1 3 1 2 3 3 2 1 0 3 1 2 3 3 0 2 2 3 1 2 1 0 1 2  
2 2 0 3 3 1 1 0 2 3 0 1 2 2 0 3 3 3 1 2 3 3 3 0 0 0 2 3 3 0 0 1 3 2 3 3 3  
0 0 2 3 3 1 0 2 0 0 0 3 2 1 2 2 1 1 0 2 3 3 0 0 1 3 3 1 3 0 3 1 1 0 2 3 3  
2 0 0 1 2 3 2 2 3 2 1 0 3 3 2 1 3 2 2 2 1 0 2 2 1 0 0 2 2 2 3 0 1 3 0 2 2  
3 0 2 0 1 1 3 0 0 2 3 1 2 0 2 0 3 0 3 3 2 3 1 2 2 1 1 1 0 1 0 3 1 0 3 0 0  
1 3 0 3 1 1 0 1 3 0 2 1 1 2 1 1 0 2 0 0 3 1 2 3 2 2 0 3 2 2 1 3 2 3 3 3 0  
2 0 3 0 1 1 2 3 1 3 1 2 0 1 2 3 0 0 1 3 0 3 0 2 2 1 1 0 2 0]
```

```
[18]: (unique, counts) = np.unique(y_pred, return_counts=True)  
price_range = np.array((unique, counts)).T  
print(price_range)
```

```
[[ 0 95]  
 [ 1 90]  
 [ 2 97]  
 [ 3 118]]
```

```
[ ]:
```