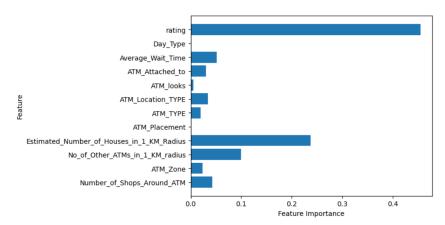
Data Analysis Report

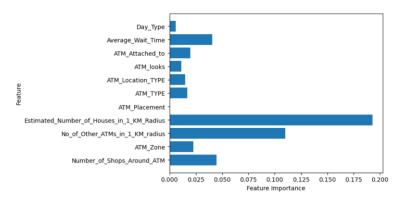
I found that there are two areas where ATM machines can stabilize revenue at around 75k.

After I used the random forest to predict the revenue of the dataset, I observed that among the features related to revenue, rating is the most influential, followed by 'Estimated_Number_of_Houses_in_1_KM_Radiu', 'No_of_Other_ATMs_in_1_KM_radius', 'Average_Wait_Time' and 'Number_of_Shops_Around_ATM'.

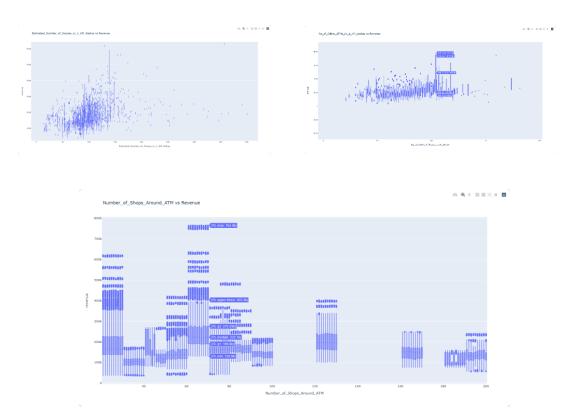


The most influential 'rating' among them may be the rating given by most users to ATM machines. Although higher ratings are more likely to result in higher revenue, which is in line with the actual situation, this is the behavior of most users but not an objective condition that we can control.

Thus I went on to plot the impact of various features on 'rating'. It can be seen that the features impact on 'rating' is the same as that on 'revenue'.



Among them, 'Average_Wait_Time' is also an uncontrollable objective factor with relatively little impact. Without it, The impact of 'Estimated_Number_of_Houses_in_1KM_Radiu', 'No_of_Other_ATMs_in_1_KM_radius', and 'Number_of_Shops_Around_ATM' on 'revenue' is shown in the figure below.



We should first pay more attention to the plot of 'Number_of_Shops_Around_ATM', because it's abnormal with too centralized data. I speculate that the ATMs in the dataset are centrally placed in batches, so there are areas where revenue can be stabilized. So I screened the feature values corresponding to the maximum revenue values in the three box plots and found that there are indeed two regions that can stabilize revenue at around 750k.

	Number_of_Shops_Around_ATM	No_of_Other_ATMs_in_1_KM_radius	Estimated_Number_of_Houses_in_1_KM_Radius	ATM_Attached_to	revenue
1182	69	160	15623	Petrol Bunk	749600
2642	70	160	15623	Petrol Bunk	747800
4102	68	160	15623	Petrol Bunk	752500
5562	69	160	15623	Petrol Bunk	752500
7022	70	160	15623	Petrol Bunk	746700

144262	66	160	15623	Petrol Bunk	748400
145722	62	160	15623	Petrol Bunk	746700
147182	68	160	15623	Petrol Bunk	748800
148642	63	160	15623	Petrol Bunk	752700
150102	63	160	15623	Petrol Bunk	752600
03 rows	s × 5 columns Number_of_Shops_Around_ATM	No_of_Other_ATMs_in_1_KM_radius	Estimated_Number_of_Houses_in_1_KM_Radius	ATM_Attached_to	revenue
103 row:		No of Other ATMs in 1 KM radius	Estimated Number of Houses in 1 KM Padius	ATM Attached to	rovonuo
03 rows		No_of_Other_ATMs_in_1_KM_radius	Estimated_Number_of_Houses_in_1_KM_Radius 21535	ATM_Attached_to Building	revenue
	Number_of_Shops_Around_ATM				
691	Number_of_Shops_Around_ATM 70	104	21535	Building	761700
691 2151	Number_of_Shops_Around_ATM 70 67	104 104	21535 21535	Building Building	761700 764700
691 2151 3611	Number_of_Shops_Around_ATM 70 67 63	104 104 104	21535 21535 21535	Building Building Building	761700 764700 757900
691 2151 3611 5071	Number_of_Shops_Around_ATM	104 104 104 104	21535 21535 21535 21535	Building Building Building Building	761700 764700 757900 764600
691 2151 3611 5071 6531	Number_of_Shops_Around_ATM	104 104 104 104 104	21535 21535 21535 21535 21535	Building Building Building Building Building	761700 764700 757900 764600 762100
691 2151 3611 5071 6531 	Number_of_Shops_Around_ATM 70 67 63 67 61	104 104 104 104 104	21535 21535 21535 21535 21535	Building Building Building Building Building	761700 764700 757900 764600 762100
691 2151 3611 5071 6531 143771	Number_of_Shops_Around_ATM 70 67 63 67 61 68	104 104 104 104 104 	21535 21535 21535 21535 21535 21535	Building Building Building Building Building	761700 764700 757900 764600 762100
691 2151 3611 5071 6531	Number_of_Shops_Around_ATM 70 67 63 67 61 68 70	104 104 104 104 104 104 104 104	21535 21535 21535 21535 21535 21535 21535	Building Building Building Building Building Building	761700 764700 757900 764600 762100 756200 758800