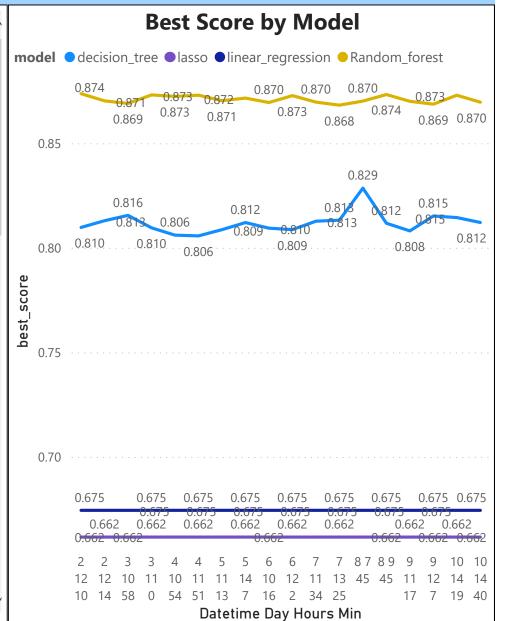


USING MACHINE LEARNING

Zaleha Ali

# **Grid Search (Best Model)**

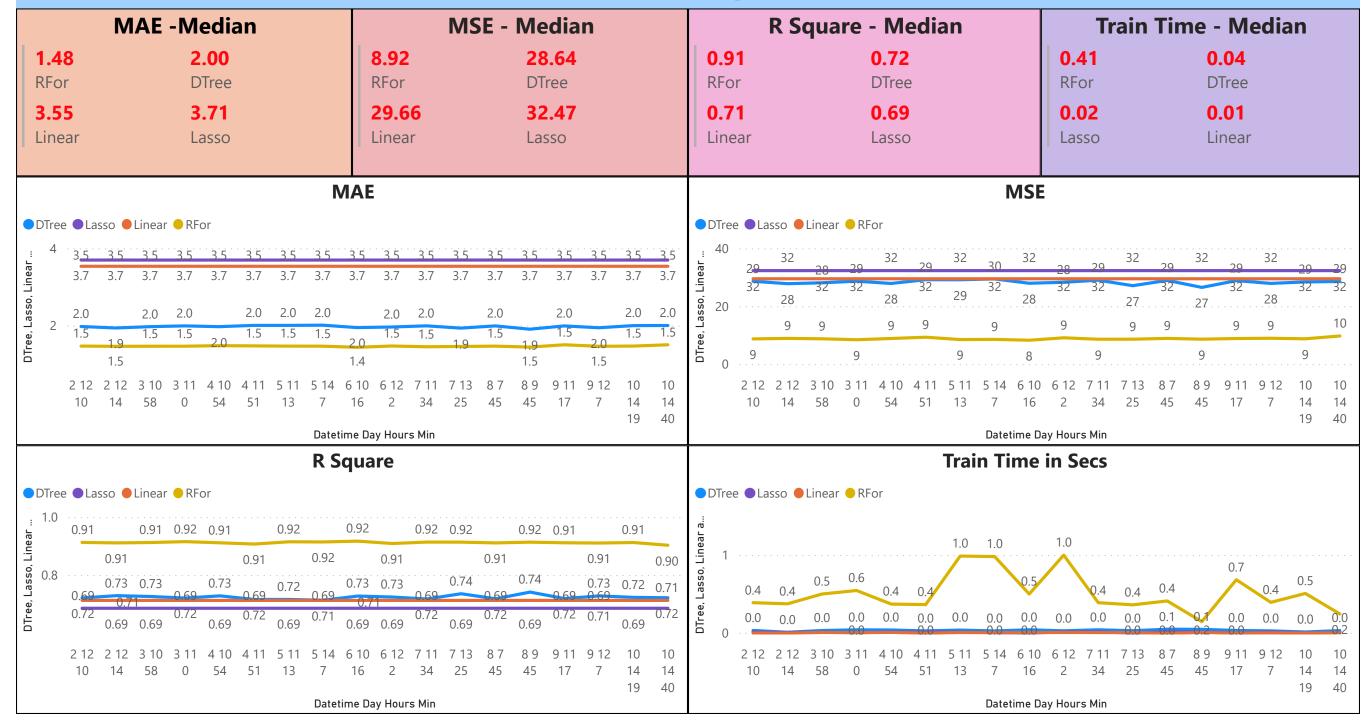
MODEL ~	best_params	best_score	model	Date ^
decision_tree	{'alpha': 1, 'selection': 'cyclic'}	0.66	lasso	5/2/20
lasso	{'criterion': 'friedman_mse', 'max_depth': 10, 'splitter': 'best'}	0.81	decision_tree	5/2/20
☐ linear_regression	{'max_features': 'sqrt', 'n_estimators': 50}	0.87	Random_forest	5/2/20
Random_forest	{'normalize': True}	0.67	linear_regression	5/2/20
random_lorest	{'alpha': 1, 'selection': 'random'}	0.66	lasso	5/2/20
	{'criterion': 'mse', 'max_depth': 10, 'splitter': 'random'}	0.81	decision_tree	5/2/20
<b>Median Best Score</b>	{'max_features': 'sqrt', 'n_estimators': 50}	0.87	Random_forest	5/2/20
(Train Data)	{'normalize': True}	0.67	linear_regression	5/2/20
decision_tree	{'alpha': 1, 'selection': 'cyclic'}	0.66	lasso	5/3/20
	{'criterion': 'friedman_mse', 'max_depth': 10, 'splitter': 'best'}	0.82	decision_tree	5/3/20
0.81	{'max_features': 'sqrt', 'n_estimators': 50}	0.87	Random_forest	5/3/20
Best_score	{'normalize': True}	0.67	linear_regression	5/3/20
	{'alpha': 1, 'selection': 'cyclic'}	0.66	lasso	5/3/20
lasso	{'criterion': 'friedman_mse', 'max_depth': 10, 'splitter': 'best'}	0.81	decision_tree	5/3/20
0.66	{'max_features': 'sqrt', 'n_estimators': 50}	0.87	Random_forest	5/3/20
Best_score	{'normalize': True}	0.67	linear_regression	5/3/20
_	{'alpha': 1, 'selection': 'random'}	0.66	lasso	5/4/20
linear_regression	{'criterion': 'friedman_mse', 'max_depth': 15, 'splitter': 'random'}	0.81	decision_tree	5/4/20
0.67	{'max_features': 'sqrt', 'n_estimators': 60}	0.87	Random_forest	5/4/20
Best_score	{'normalize': True}	0.67	linear_regression	5/4/20
Dest_score	{'alpha': 1, 'selection': 'cyclic'}	0.66	lasso	5/4/20
Random_forest	{'criterion': 'mse', 'max_depth': 15, 'splitter': 'best'}	0.81	decision_tree	5/4/20
	{'max_features': 'sqrt', 'n_estimators': 50}	0.87	Random_forest	5/4/20
0.87	{'normalize': True}	0.67	linear_regression	5/4/20
Best_score	{'alpha': 1, 'selection': 'random'}	0.66	lasso	5/5/20
	{'criterion': 'friedman_mse', 'max_depth': 10, 'splitter': 'best'}	0.81	decision_tree	5/5/2C <sup>&gt;</sup>



# **Median R Square (Test Data)**

C	0.72	0.69	0.71	0.91
	Decision Tree	Lasso	Linear Regression	Random Forest

# **Parameters - MAE, MSE, R Square and Train Time**



# **Random Predictions**

Label	<b>~</b>	
0		
□ 1	- 1	
_ 2		
3		
4		
5		

_									_
	Brand_Name ▲	Label	Fuel_Type	Label	Transmission	Label	Owner_Type	Label	^
	Ambassador	0	CNG	0	Automatic	0	First	0	
	Audi	1	Diesel	1	Manual	1	Fourth & Above	1	
	Bentley	3	LPG	3			Third	3	
	BMW	2	Electric	2			Second	2	~
	Chevrolet	4	Petrol	4					

# Median Prediction Value 2.93 Prediction\_1

**2.34** Prediction\_2



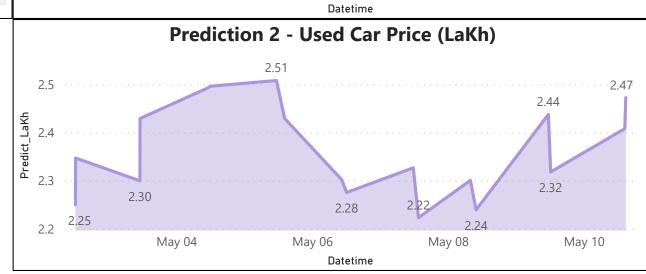
# **Prediction 1 - Using Labels**

Brand_Name	Year	Engine_CC	Fuel_Type	Km_Driven	Mileage_kmk	Owner_Type	Power_bhp	Predict_L^
					g			
17	2015	1000	3	900000	15	0	40	
17	2015	1000	3	900000	15	0	40	;
17	2015	1000	3	900000	15	0	40	
17	2015	1000	3	900000	15	0	40	
17	2015	1000	3	900000	15	0	40	;
17	2015	1000	3	900000	15	0	40	
17	2015	1000	3	900000	15	0	40	
17	2015	1000	3	900000	15	0	40	
17	2015	1000	3	900000	15	0	40	> '

# Prediction 1 - Used Car Price (LaKh) 3.2 3.25 3.20 3.20 3.20 3.20 3.20 3.21 3.00 3.21 3.00 3.22 3.00 3.01 3.02 3.00 3.02 3.00

## **Prediction 2 - Using Values**

Brand_Name	Year	Engine_CC	Fuel_Type	Km_Driven	Mileage_kmkg	Owner_Type	Power_bhp^
Maruti	2013	1000	LPG	100000	15	First	40
Maruti	2013	1000	LPG	100000	15	First	40
Maruti	2013	1000	LPG	100000	15	First	40
Maruti	2013	1000	LPG	100000	15	First	40
Maruti	2013	1000	LPG	100000	15	First	40
Maruti	2013	1000	LPG	100000	15	First	40
Maruti	2013	1000	LPG	100000	15	First	40
Maruti	2013	1000	LPG	100000	15	First	40 >



# **Actual Vs Predicted (Random Forest)**

