

# Customer Lifetime Value Analysis

Auto Insurance Company



# OUR TEAM CREW



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# PROCESS TIMELINE



Data Understanding

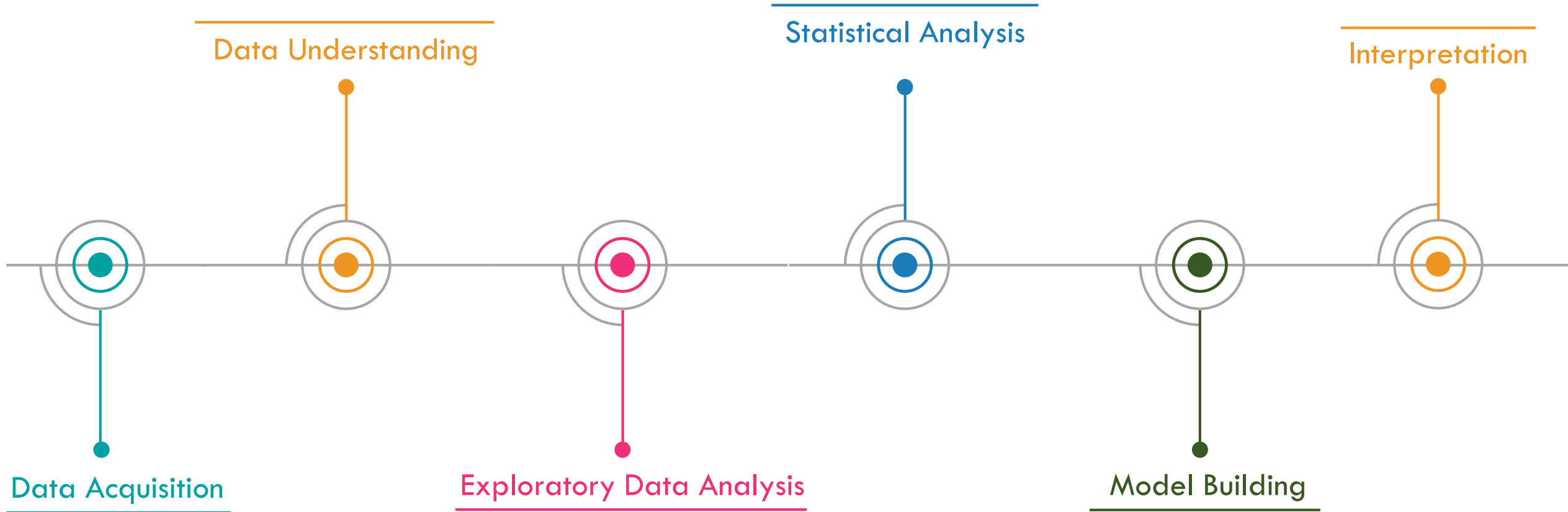
Statistical Analysis

Interpretation

Data Acquisition

Exploratory Data Analysis

Model Building



# PROBLEM STATEMENT



We are expected to create an analytical and modelling framework to predict the lifetime value of each customer based on the quantitative and qualitative features.

## Company's Approach and Insight

- A major non-life insurance company wants to evaluate customer lifetime value based on each customer's demographics and policy information including claim details.
- The CLV is a profitability metric in terms of a value placed by the company on each customer and can be conceived in two dimensions: the customer's present Value and potential future Value.



# DATA DESCRIPTION



CLTV



Total Claim Amount



Employment Status



No. of Policies



State



Income



Gender



Policy/Policy Type



Months Since Last Claim



Response



Location Code



Renew Offer Type



Monthly Premium Auto



Coverage



Marital Status



Sales Channel



Months Since Last Policy Inception



Education



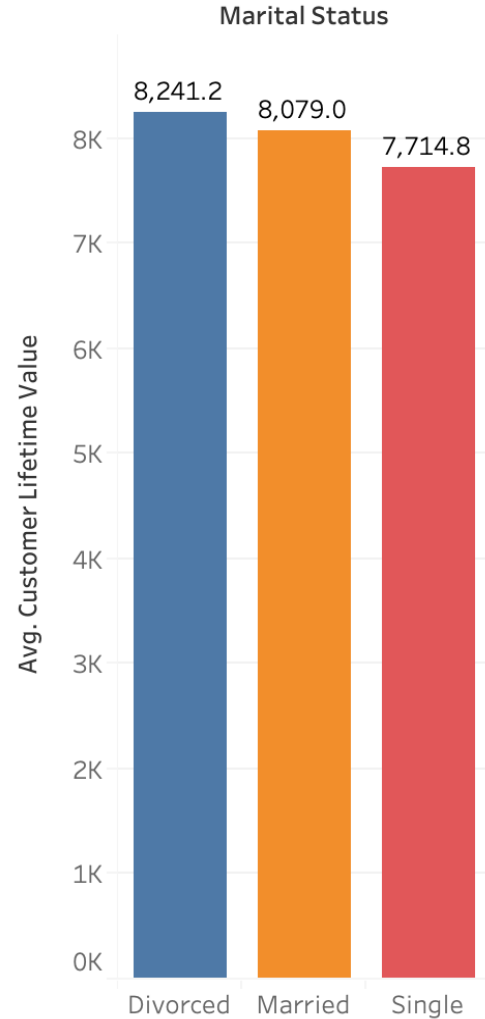
Number of open complaints



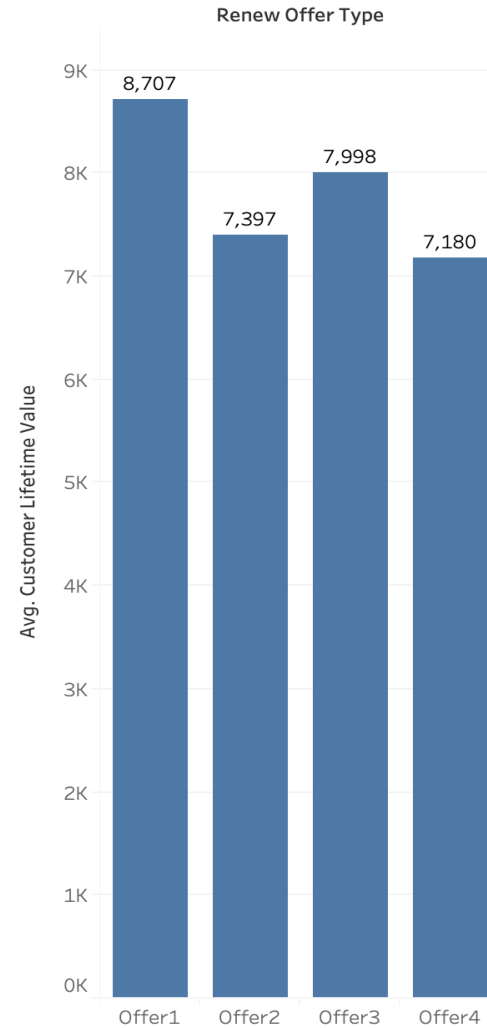
Vehicle Class/Vehicle Size

# Exploratory Data Analysis

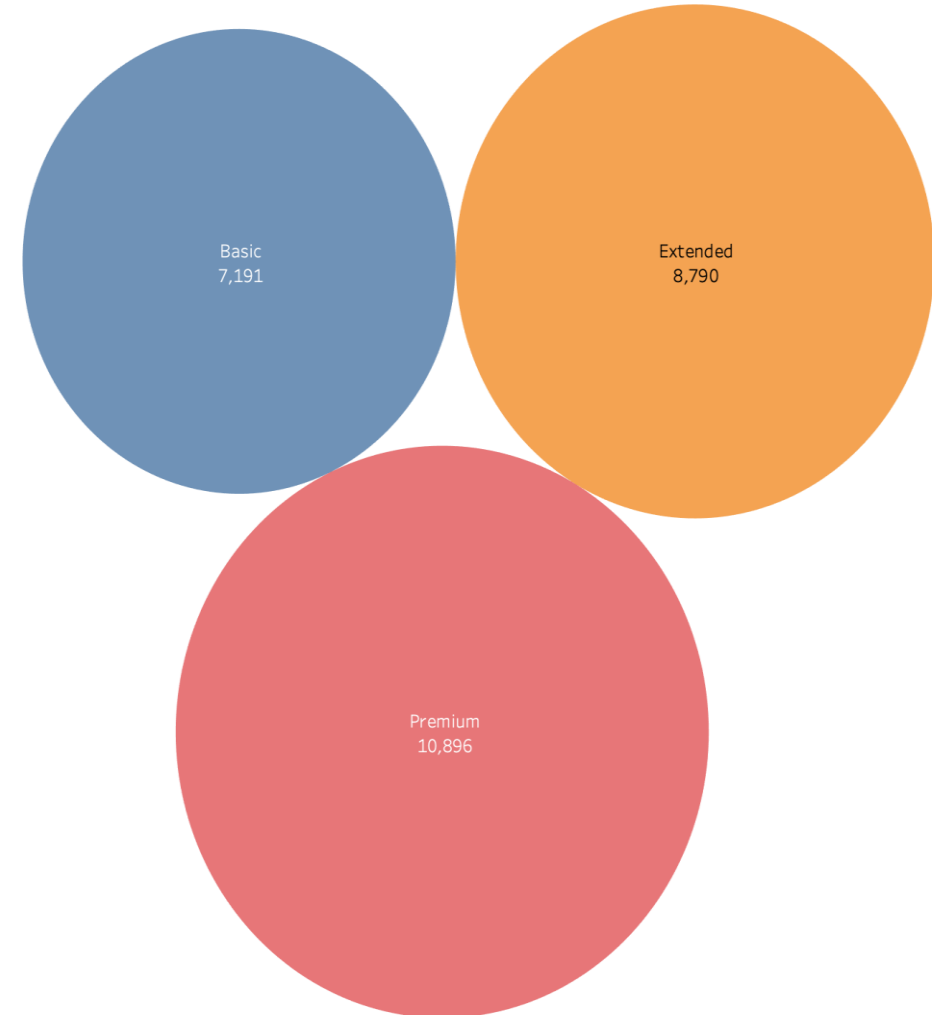
## CLTV vs Marital Status



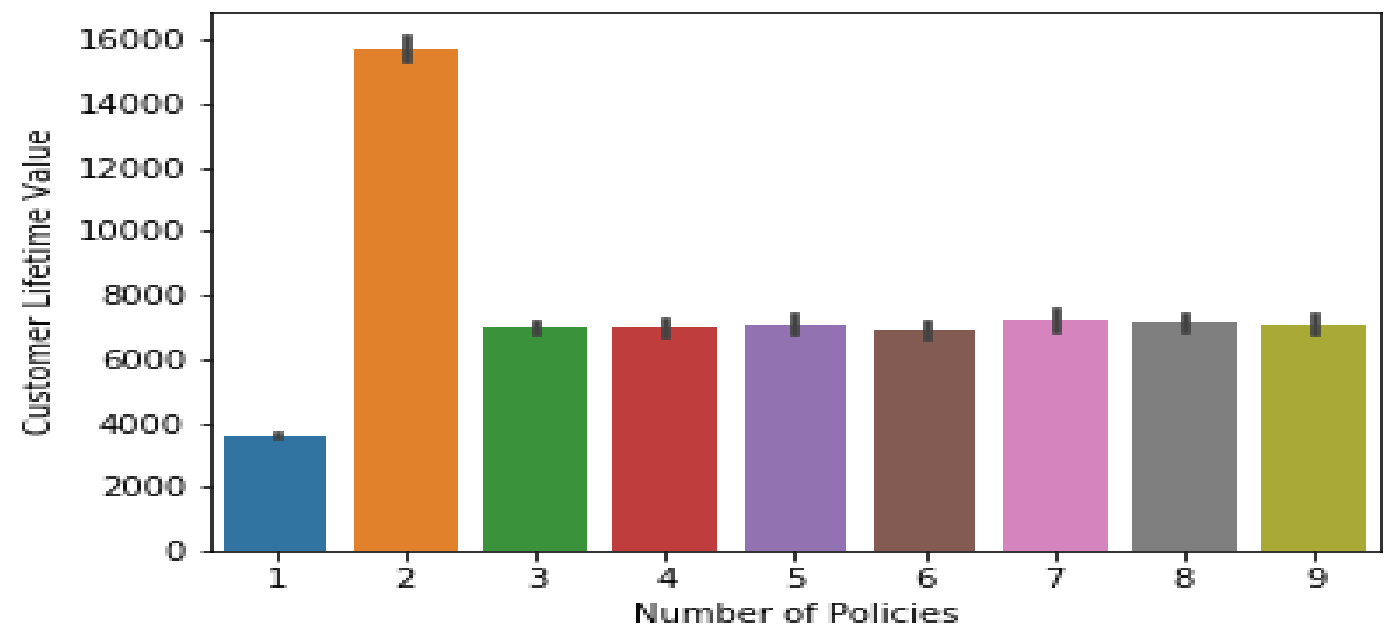
## CLTV vs Renew offer type



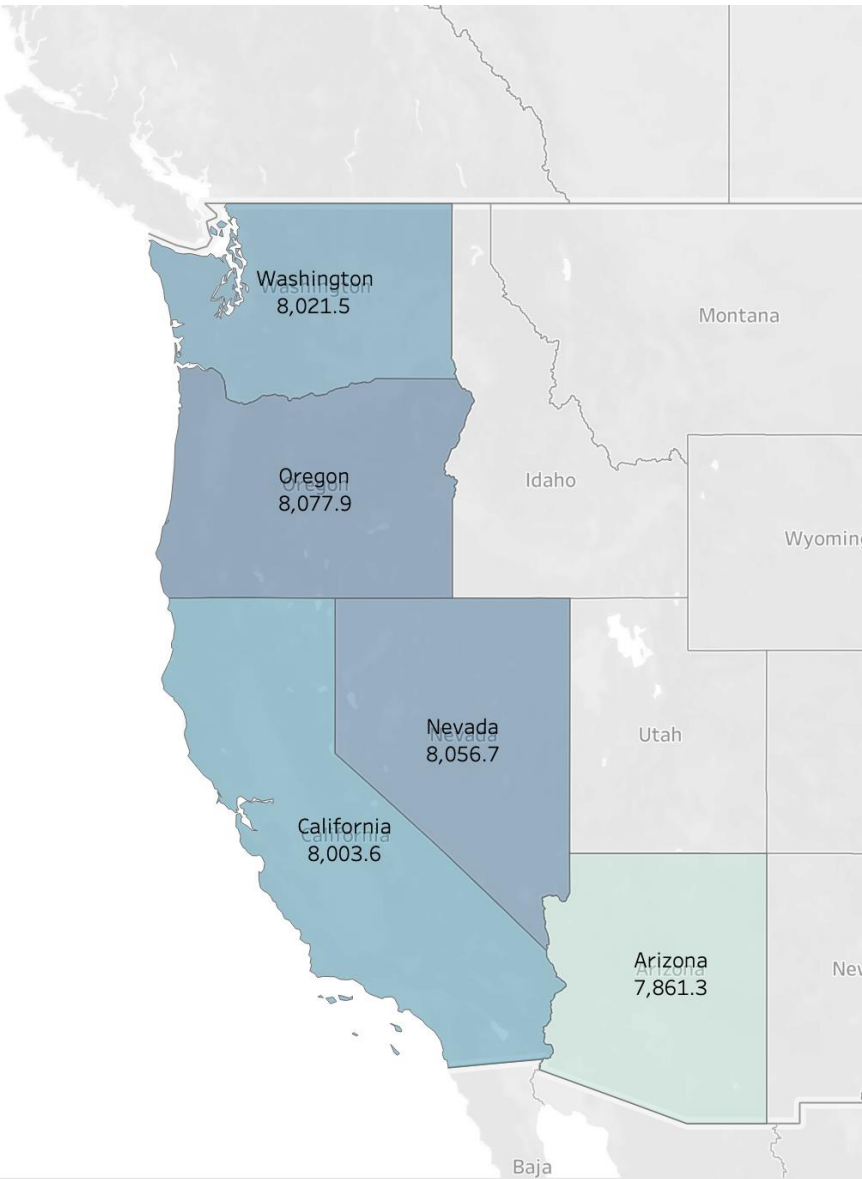
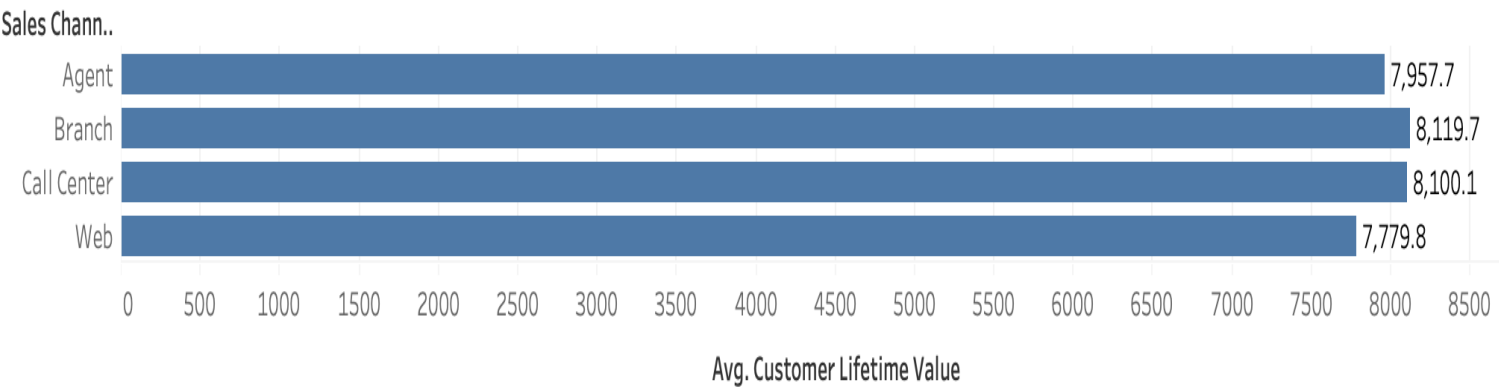
## CLTV vs coverage



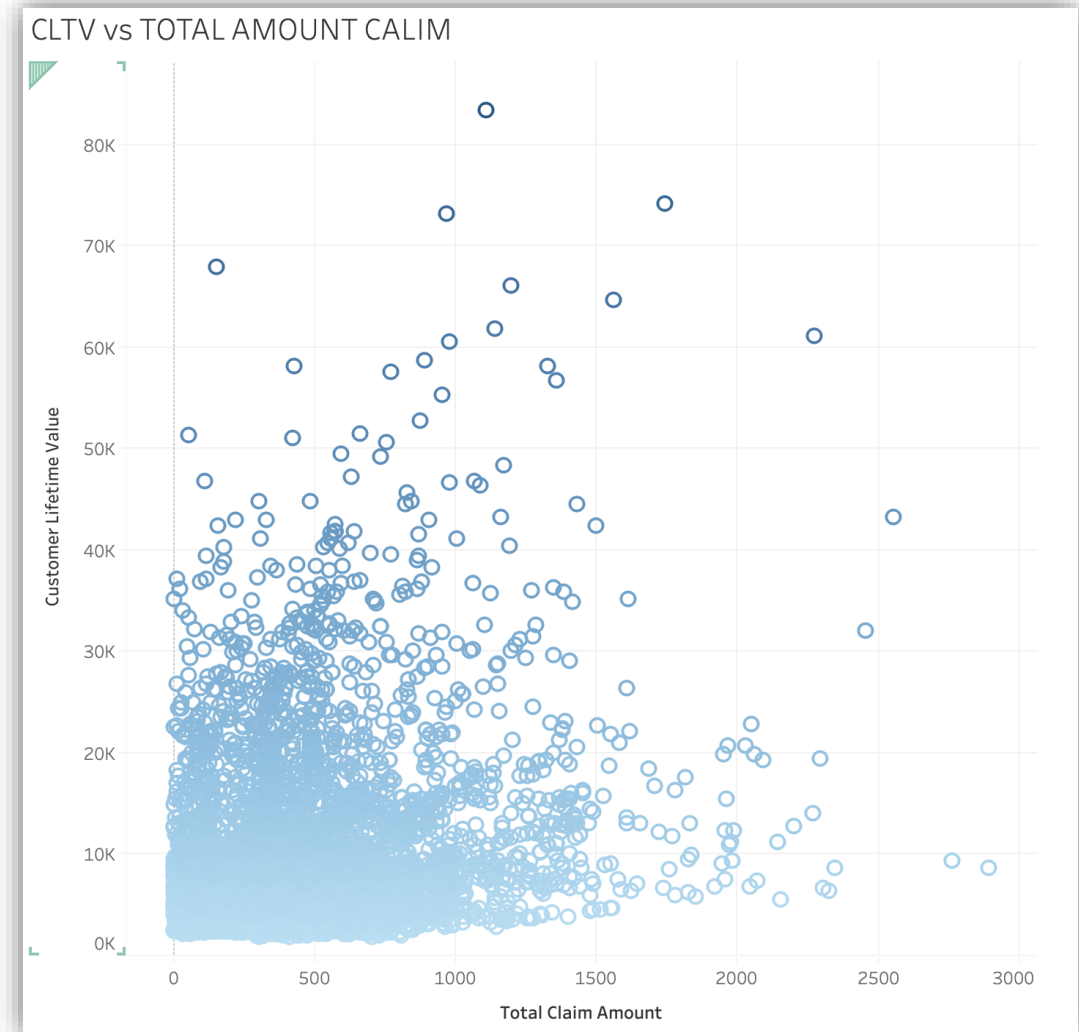
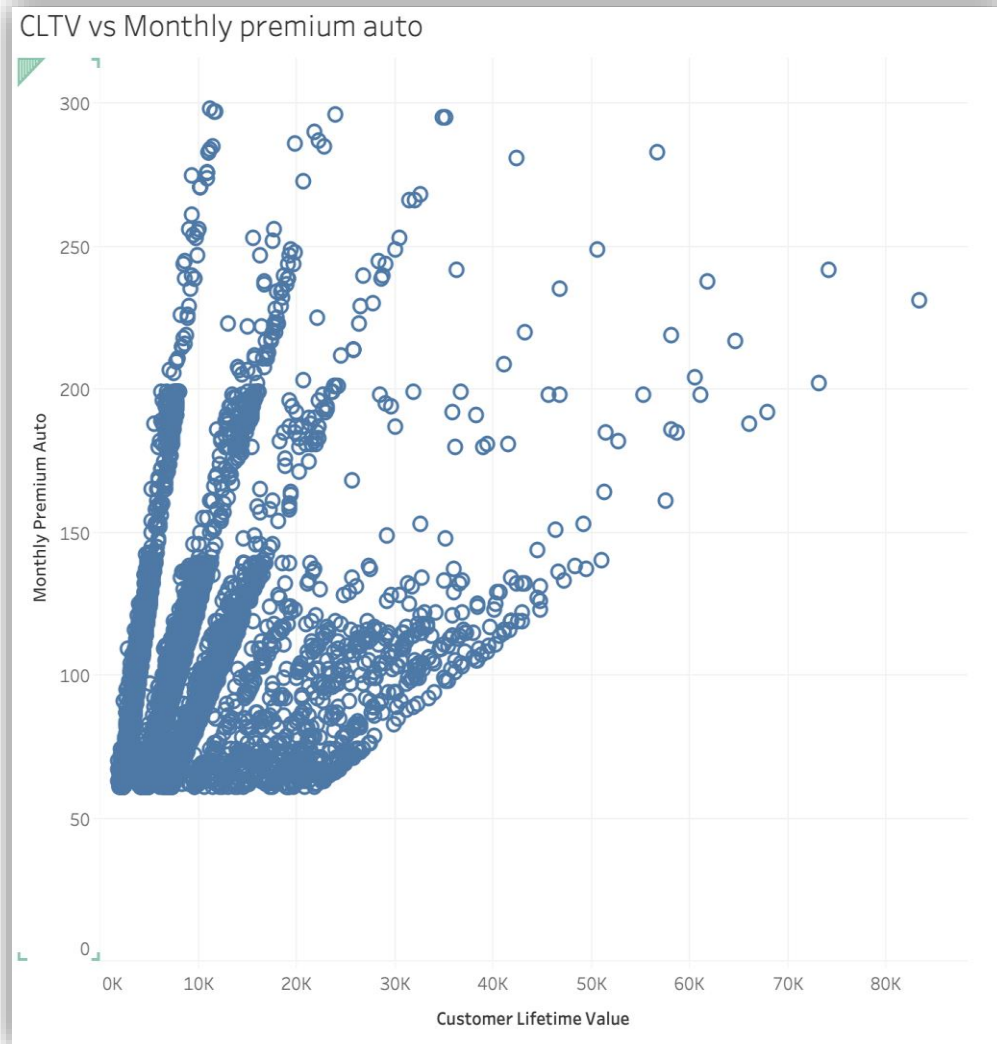
# Exploratory Data Analysis



CLTV vs Sales Channel

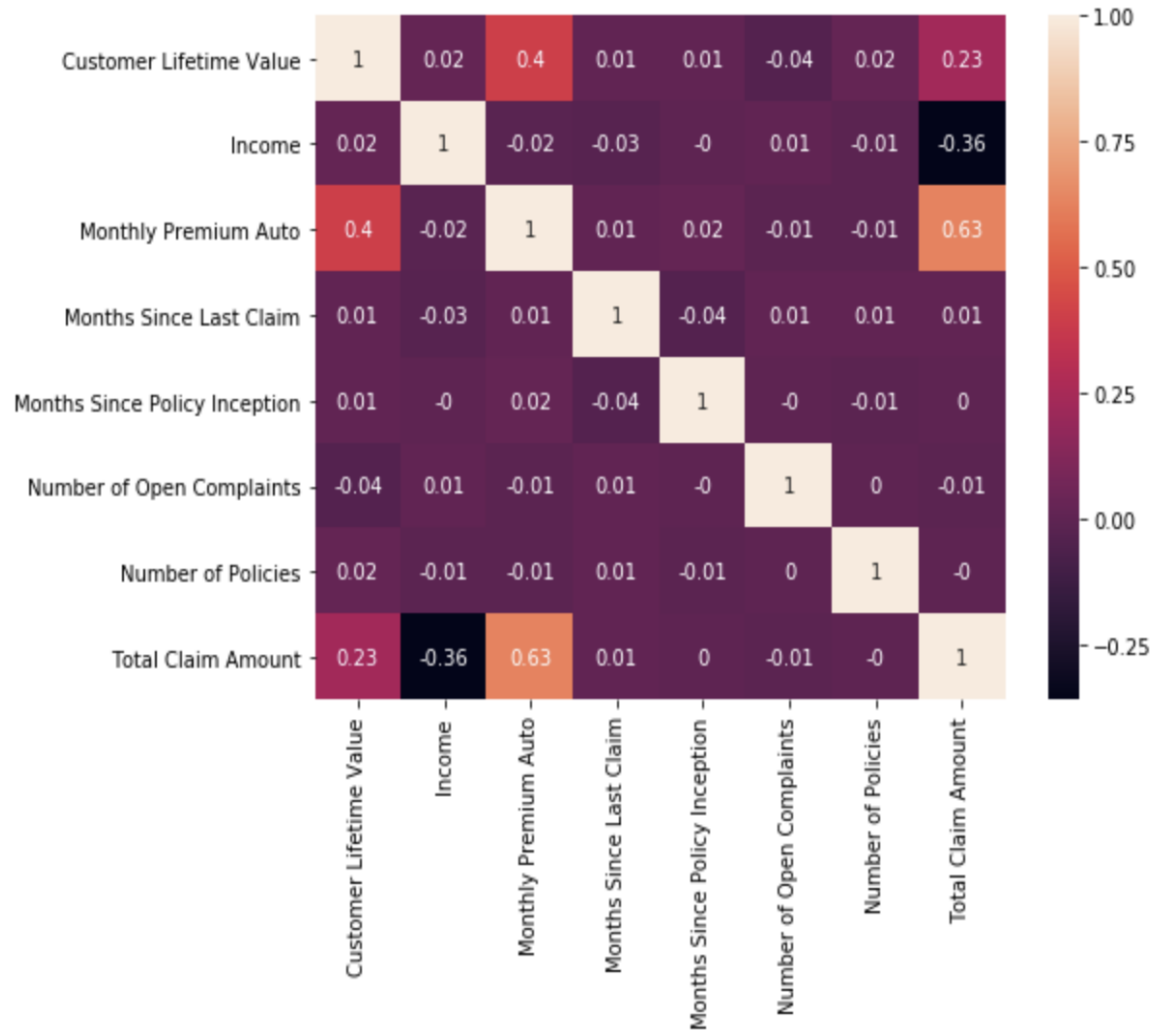


# Exploratory Data Analysis





# Exploratory Data Analysis



## Heatmap

- With respect to our target variable, only Monthly Premium Auto & Total Claim Amount are highly correlated.
- Other than these two variables, there is no correlation amongst the independent Numerical variables

# Base Model: OLS

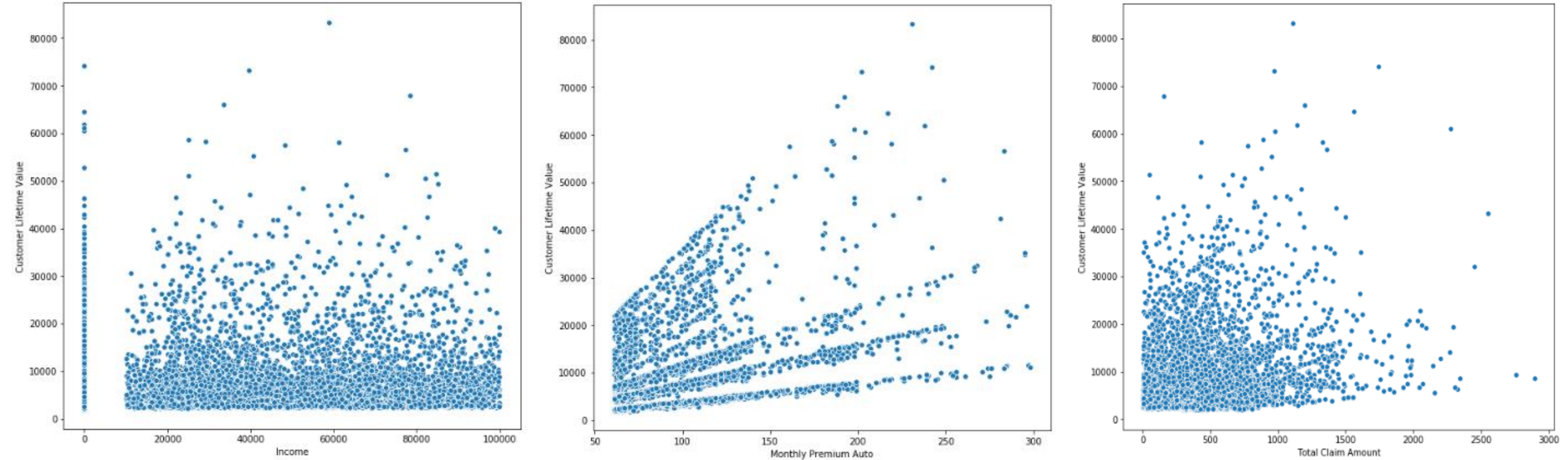
OLS Regression Results

<b>Dep. Variable:</b>	Customer Lifetime Value	<b>R-squared:</b>	0.166
<b>Model:</b>	OLS	<b>Adj. R-squared:</b>	0.164
<b>Method:</b>	Least Squares	<b>F-statistic:</b>	78.82
<b>Date:</b>	Wed, 13 Nov 2019	<b>Prob (F-statistic):</b>	0.00
<b>Time:</b>	12:16:53	<b>Log-Likelihood:</b>	-92831.
<b>No. Observations:</b>	9134	<b>AIC:</b>	1.857e+05
<b>Df Residuals:</b>	9110	<b>BIC:</b>	1.859e+05
<b>Df Model:</b>	23		
<b>Covariance Type:</b>	nonrobust		

	coef	std err	t	P> t	[0.025	0.975]
<b>const</b>	872.5455	499.361	1.747	0.081	-106.315	1851.406
<b>Customer</b>	-0.0767	0.025	-3.073	0.002	-0.126	-0.028
<b>State</b>	30.3958	51.107	0.595	0.552	-69.786	130.578
<b>Response</b>	-434.8693	192.448	-2.260	0.024	-812.111	-57.627
<b>Coverage</b>	-186.4730	113.665	-1.641	0.101	-409.281	36.335
<b>Education</b>	93.7398	47.771	1.962	0.050	0.097	187.382
<b>Effective To Date</b>	-0.1302	3.883	-0.034	0.973	-7.742	7.481
<b>EmploymentStatus</b>	-104.9096	73.576	-1.426	0.154	-249.134	39.315
<b>Gender</b>	-137.3192	132.608	-1.036	0.300	-397.261	122.623
<b>Income</b>	0.0022	0.003	0.662	0.508	-0.004	0.009
<b>Location Code</b>	109.7248	116.551	0.941	0.347	-118.742	338.191
<b>Marital Status</b>	-237.4657	110.640	-2.146	0.032	-454.344	-20.587
<b>Monthly Premium Auto</b>	82.7472	2.923	28.304	0.000	77.017	88.478
<b>Months Since Last Claim</b>	7.0304	6.544	1.074	0.283	-5.798	19.859
<b>Months Since Policy Inception</b>	-0.8386	2.373	-0.353	0.724	-5.489	3.812
<b>Number of Open Complaints</b>	-250.4702	72.355	-3.462	0.001	-392.303	-108.638
<b>Number of Policies</b>	63.8915	27.655	2.310	0.021	9.681	118.102
<b>Policy Type</b>	441.5321	290.952	1.518	0.129	-128.798	1011.863
<b>Policy</b>	-48.2763	86.230	-0.560	0.576	-217.306	120.753
<b>Renew Offer Type</b>	-348.8194	68.467	-5.095	0.000	-483.030	-214.609
<b>Sales Channel</b>	22.4563	62.272	0.361	0.718	-99.610	144.523
<b>Total Claim Amount</b>	-0.5882	0.362	-1.626	0.104	-1.298	0.121
<b>Vehicle Class</b>	45.9054	32.409	1.416	0.157	-17.623	109.434
<b>Vehicle Size</b>	187.3131	123.322	1.519	0.129	-54.426	429.052

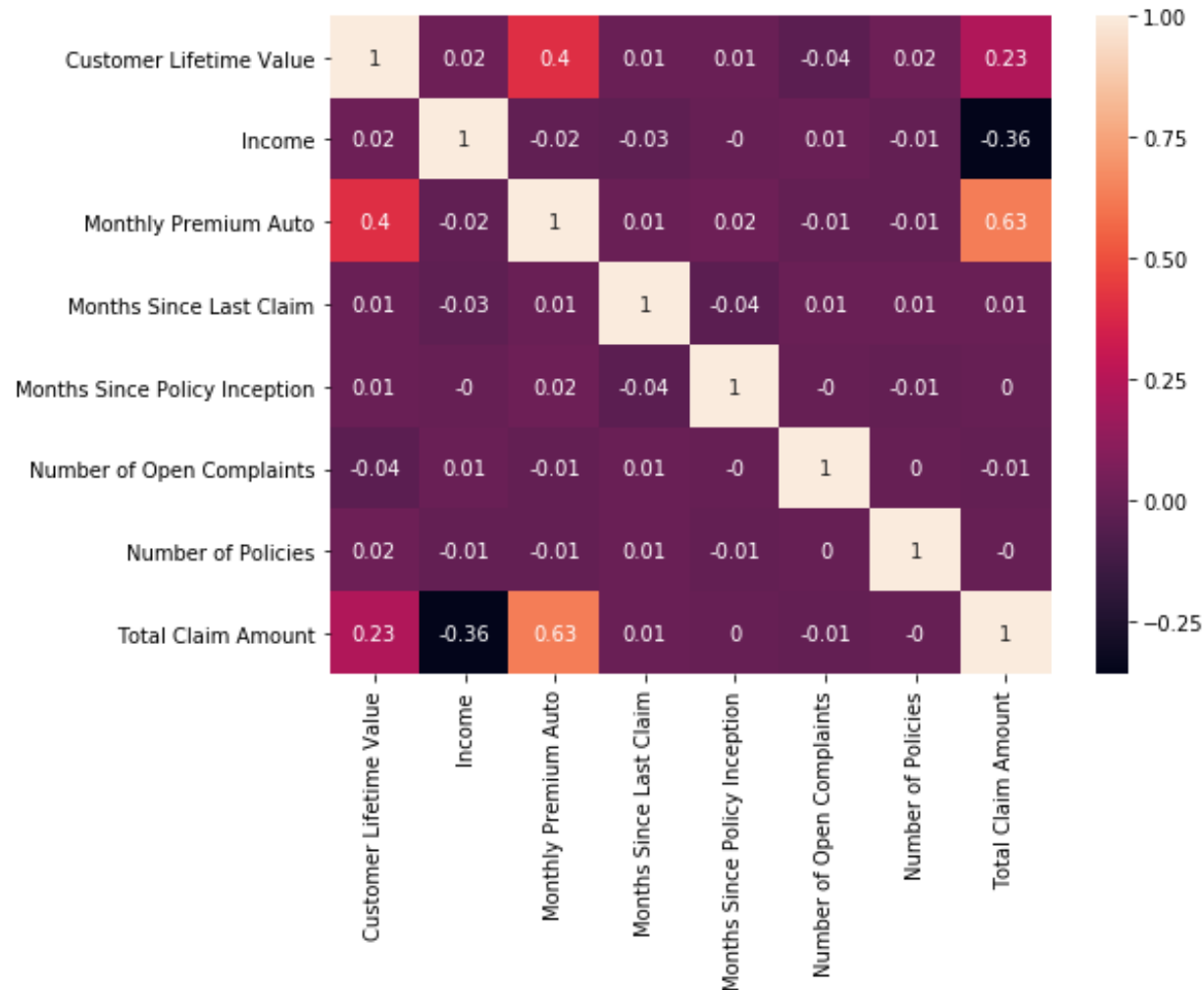
# Assumptions for Linear Regression

1. Linear relationship exists between the dependent and predictor variables



# Assumptions for Linear Regression

## 2. Multicollinearity : Independent variables should not be correlated



Iteration : 1

VIF Factor

12.406466

6.664016

3.267690

2.829624

2.784210

2.296918

1.165267

features

Monthly Premium Auto

Total Claim Amount

Months Since Policy Inception

Income

Months Since Last Claim

Number of Policies

Number of Open Complaints

Iteration : 2

VIF Factor

3.065106

2.667965

2.571419

2.250677

2.128979

1.163048

features

Months Since Policy Inception

Months Since Last Claim

Total Claim Amount

Number of Policies

Income

Number of Open Complaints

# Assumptions for Linear Regression

3. Auto Correlation : There should be no correlation between the residuals

Durbin-Watson Test :

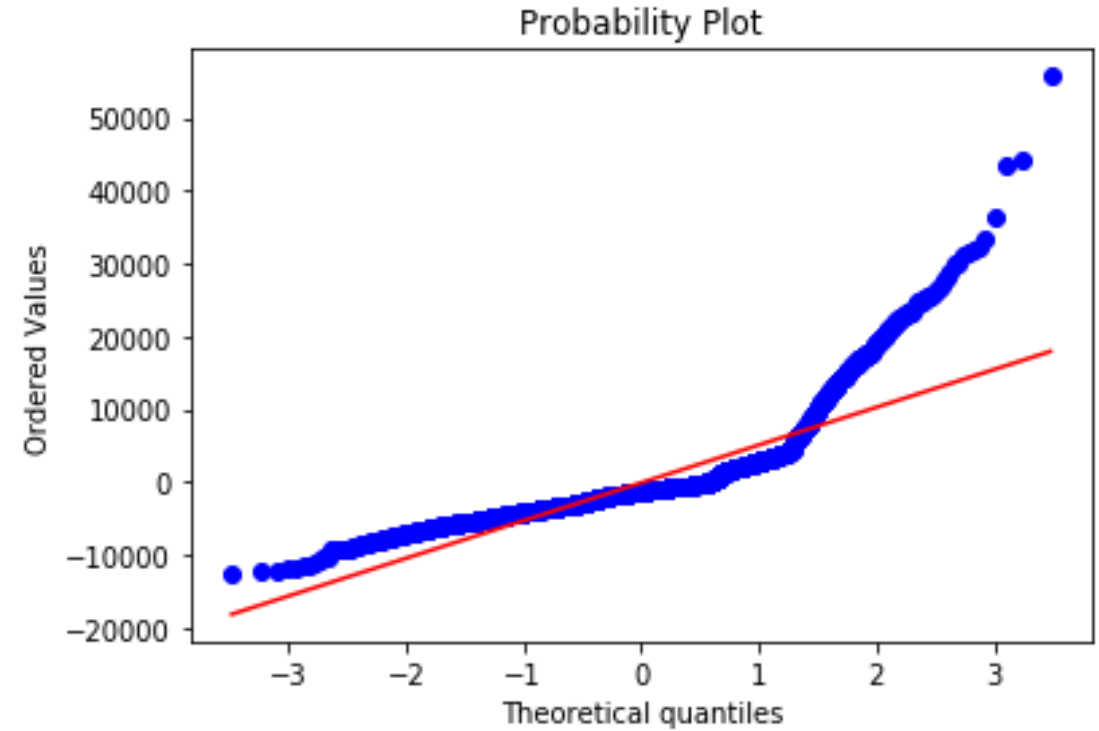
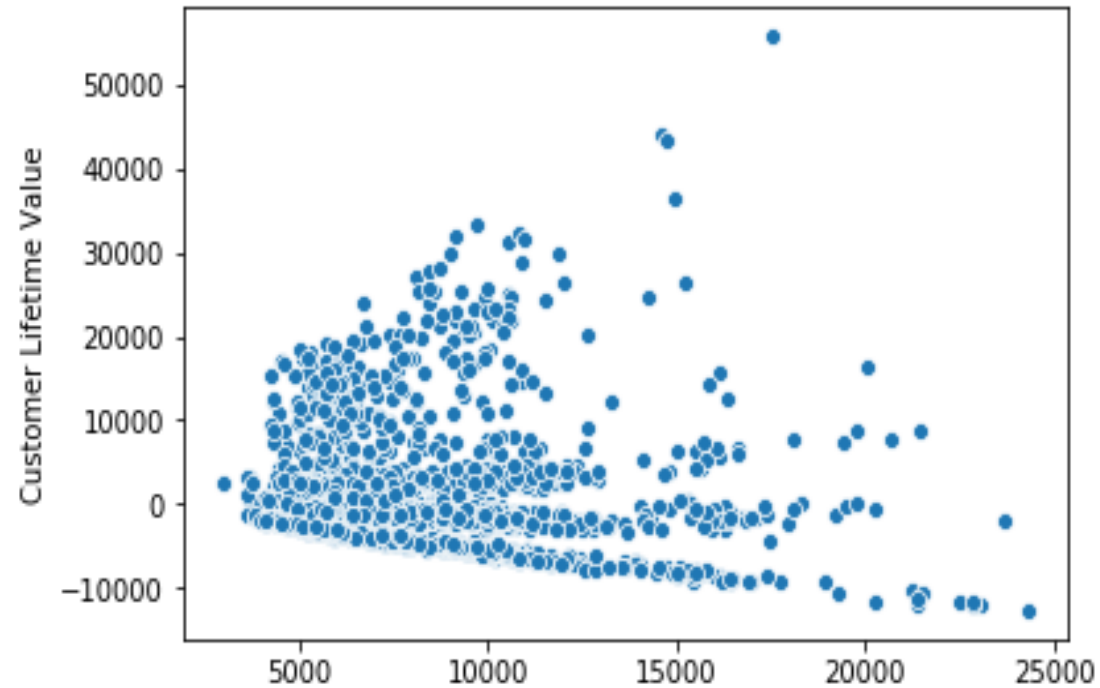
- Measure the relevance of features by their correlation with dependent variable
- Uses statistical methods for evaluation

Ljung Box Test :

- Measure the relevance of features by their correlation with dependent variable
- Uses statistical methods for evaluation

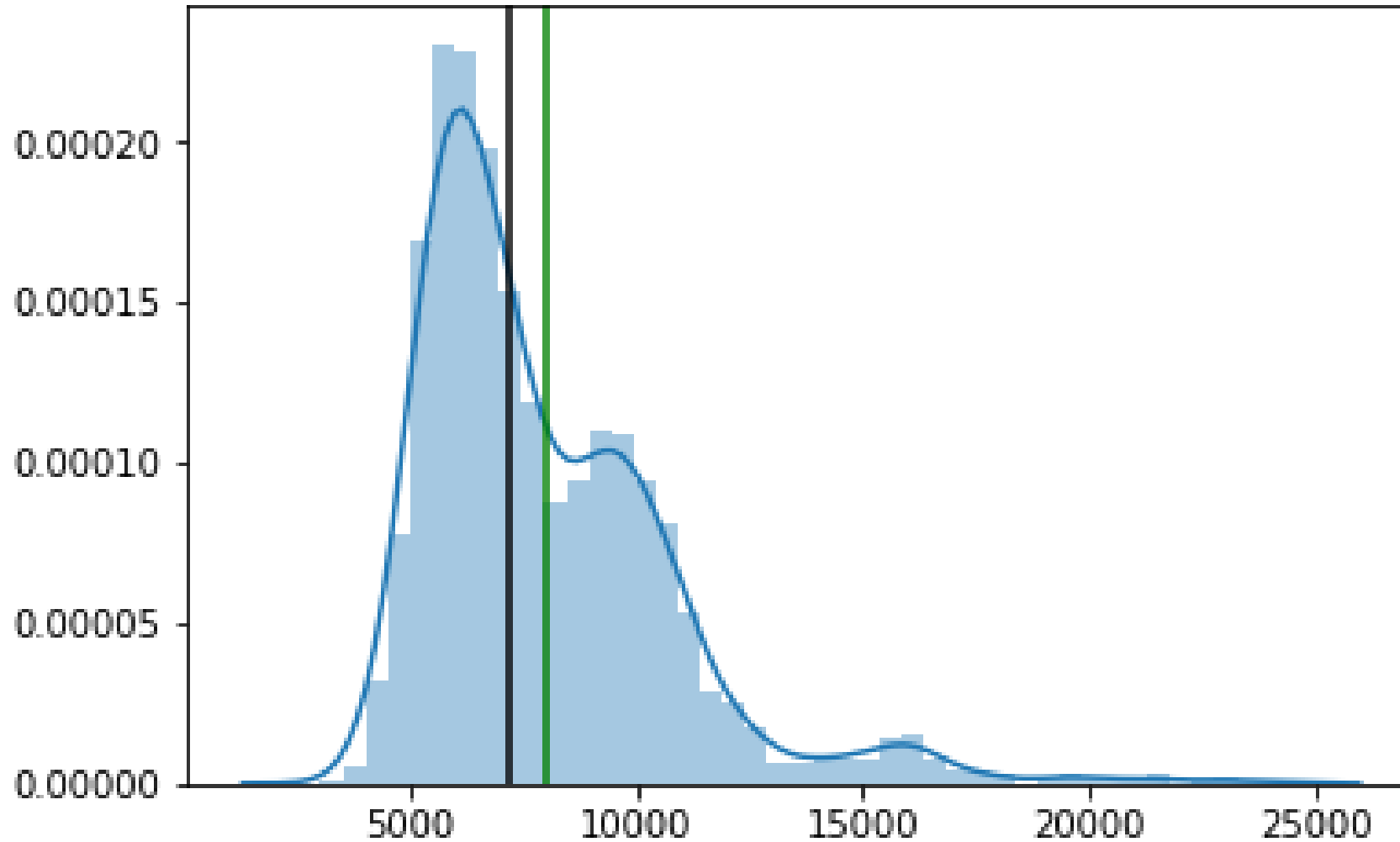
# Assumptions for Linear Regression

## 4. Homoskedasticity : Residuals must have constant variance



# Assumptions for Linear Regression

## 5. Residuals must be normally distributed

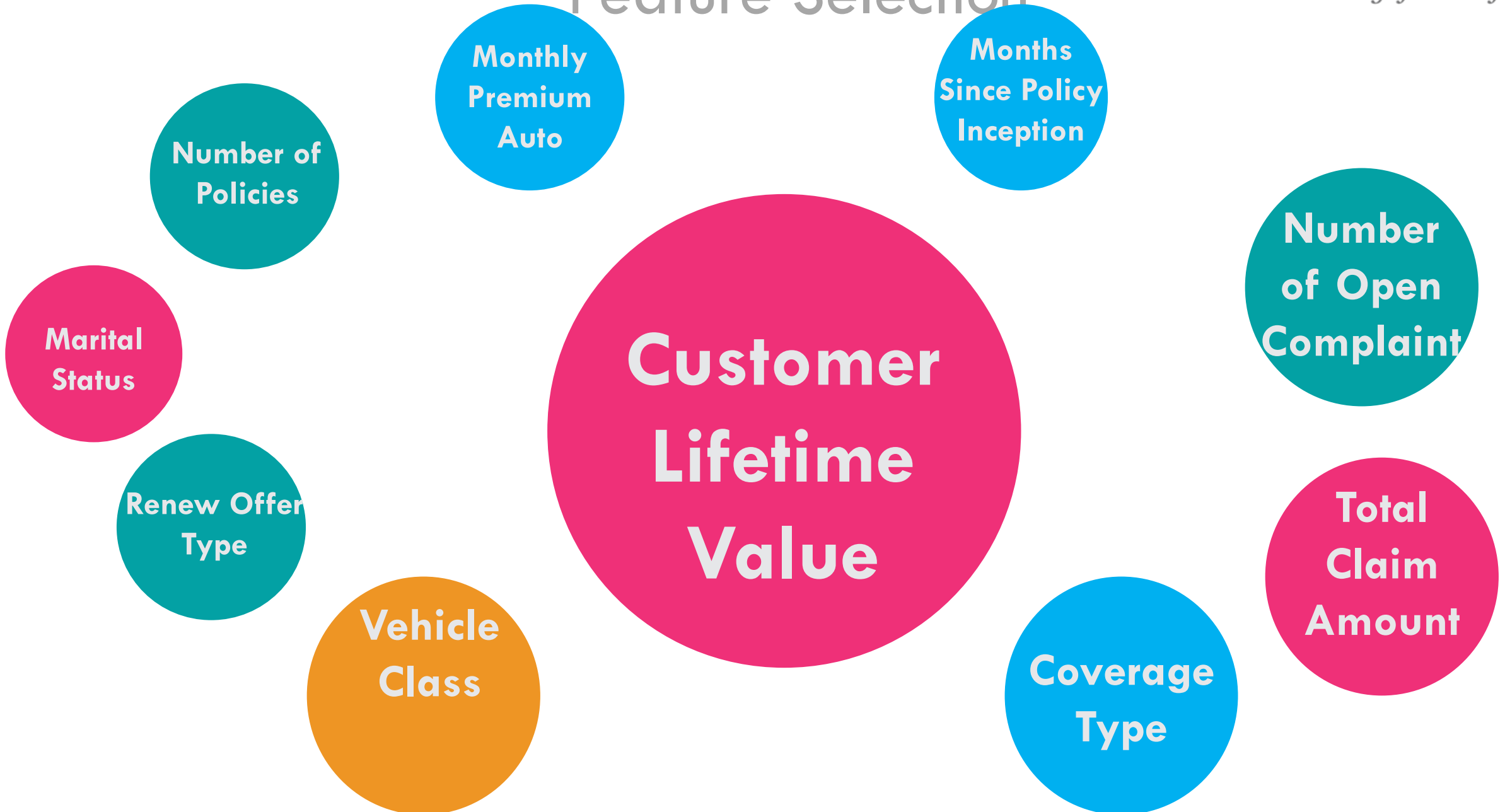


# Statistical Analysis

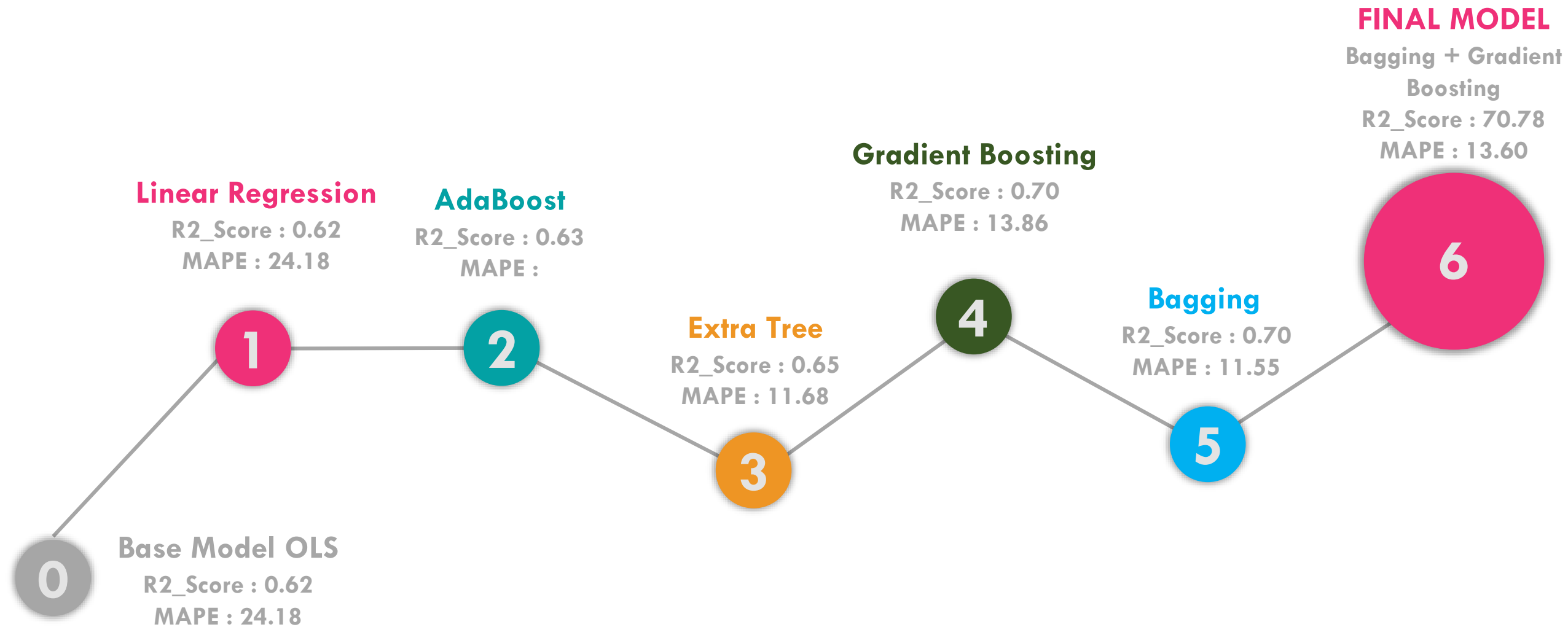
Statistical Test	Feature	T-Statistic	P-Value
ANOVA	State	0.273	0.896
ANOVA/ttest_ind	Customer Response	0.728	0.393
ANOVA	Coverage	133.675	6.01e-58
ANOVA	Education	2.422	0.046
ANOVA	Employment Status	3.809	0.004
ANOVA/ttest_ind	Gender	1.691	0.193
ANOVA	Location Code	0.108	0.897
ANOVA	Marital Status	3.317	0.036
ANOVA	Policy	1.183	0.304
ANOVA	Renew Offer Type	25.832	1.23e-16
ANOVA	Sales Channel	0.880	0.450
ANOVA	Vehicle Class	267.158	2.08e-267
ANOVA	Vehicle Size	2.382	0.092



## Feature Selection



# Model Comparison



# Magic Model

10

**KFold Validation using  
Bagging regressor with base  
estimator Gradient Boosting.**

**R2\_Score : 0.86**

**MAPE :**



**THANK YOU FOR YOUR TIME**