In [1]: In [2]:	<pre>import pandas as pd df=pd.read_csv('insurance.csv')# ### Prediction on Test Data</pre>
In [3]: Out[3]:	df.head() age sex bmi children smoker region charges 0 19 female 27.900 0 yes southwest 16884.92400
	1 18 male 33.770 1 no southeast 1725.55230 2 28 male 33.000 3 no southeast 4449.46200 3 33 male 22.705 0 no northwest 21984.47061
In [4]: Out[4]:	4 32 male 28.880 0 no northwest 3866.85520 df.tail() age sex bmi children smoker region charges
	133 50 male 30.97 3 no northwest 10600.5483 133 18 female 31.92 0 no northeast 2205.9808 133 18 female 36.85 0 no southeast 1629.8335 133 21 female 25.80 0 no southwest 2007.9450
	1336 21 female 25.80 0 no southwest 2007.9450 1337 61 female 29.07 0 yes northwest 29141.3603 #### Finding No. of rows & Columns
In [5]: Out[5]:	df.shape
	### Checking null value in Dataset
In [6]: Out[6]:	
	2 False False False False False False False 3 False False False False False False False 4 False False False False False False False
	1333FalseFalseFalseFalseFalseFalse1334FalseFalseFalseFalseFalseFalse1335FalseFalseFalseFalseFalseFalseFalse
	1336FalseFalseFalseFalseFalseFalse1337FalseFalseFalseFalseFalseFalse1338rows × 7 columns
In [7]: Out[7]:	Sex 0
	<pre>bmi 0 children 0 smoker 0 region 0 charges 0 dtype: int64</pre>
In [8]:	### Getting Overall Statistics of Dataset
Out[8]:	age bmi children charges count 1338.000000 1338.000000 1338.000000 1338.000000 mean 39.207025 30.663397 1.094918 13270.422265
	std 14.049960 6.098187 1.205493 12110.011237 min 18.000000 15.960000 0.000000 1121.873900 25% 27.000000 26.296250 0.000000 4740.287150 50% 39.00000 30.400000 1.000000 9382.033000
	75% 51.000000 34.693750 2.000000 16639.912515 max 64.000000 53.130000 5.000000 63770.428010
In [9]: Out[9]:	df.describe(include='all') # statistics for caterogical and numerical column both age sex bmi children smoker region charges count 1338.000000 1338 1338.000000 1338.000000 1338 1338 1338.000000 unique NaN NaN NaN NaN
	top NaN male NaN no southeast NaN freq NaN 676 NaN NaN 1064 364 NaN mean 39.207025 NaN 30.663397 1.094918 NaN NaN 13270.422265
	std 14.049960 NaN 6.098187 1.205493 NaN NaN 12110.011237 min 18.000000 NaN 15.960000 0.000000 NaN NaN 1121.873900 25% 27.000000 NaN 26.296250 0.000000 NaN NaN 4740.287150 50% 39.000000 NaN 30.400000 1.000000 NaN NaN 9382.033000
In [10]:	75% 51.000000 NaN 34.693750 2.000000 NaN NaN 16639.912515 max 64.000000 NaN 53.130000 5.000000 NaN NaN 63770.428010 df.head()
Out[10]:	age sex bmi children smoker region charges 0 19 female 27.900 0 yes southwest 16884.92400 1 18 male 33.770 1 no southeast 1725.55230
	2 8 male 33.000 3 no southeast 4449.46200 3 33 male 22.705 0 no northwest 21984.47061 4 32 male 28.880 0 no northwest 3866.85520
In [11]:	### Converting Columns from string to Integer df['sex'].unique()
	<pre>df['sex']= df['sex'].map({'female':0, 'male':1})</pre>
Out[13]:	<pre>df['sex'] 0 0 1 1 2 1 3 1 4 1</pre>
	1333 1 1334 0 1335 0 1336 0
In [14]: Out[14]:	
	0 19 0 27.900 0 yes southwest 16884.92400 1 18 1 33.770 1 no southeast 1725.55230 2 28 1 33.000 3 no southeast 4449.46200 3 33 1 22.705 0 no northwest 21984.47061
In [15]:	4 32 1 28.880
In [16]:	<pre>array(['yes', 'no'], dtype=object) df['smoker']= df['smoker'].map({'yes':1, 'no':0}) df['smoker']</pre>
Out[17]:	
	1333 0 1334 0 1335 0 1336 0 1337 1
In [18]: Out[18]:	Name: smoker, Length: 1338, dtype: int64 df['region'].unique()
<pre>In [19]: In [20]: Out[20]:</pre>	<pre>df['region']=df['region'].map({'southwest':1, 'southeast':2, 'northwest':3, 'northeast':4}) df['region'] 0 1 1 2</pre>
v]:	1 2 2 2 3 3 3 4 3 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5
	1334
In [21]: Out[21]:	df.head() age sex bmi children smoker region charges 0 19 0 27.900 0 1 1 16884.92400 1 18 1 33.770 1 0 2 1725.55230
	2 28 1 33.000 3 0 2 4449.46200 3 33 1 22.705 0 0 3 21984.47061 4 32 1 28.880 0 0 3 3866.85520
In [22]: Out[22]:	<pre>df.columns Index(['age', 'sex', 'bmi', 'children', 'smoker', 'region', 'charges'], dtype='object')</pre>
In [23]:	### Storing Features Matrix in X and response in Y x = df.drop(['charges'],axis=1)
In [24]: Out[24]:	age sex bmi children smoker region 0 19 0 27.900 0 1 1
	1 18 1 33.770 1 0 2 2 28 1 33.000 3 0 2 3 33 1 22.705 0 0 3 4 32 1 28.880 0 0 3
	1333 50 1 30.970 3 0 3 1334 18 0 31.920 0 0 4
	1335 18 0 36.850 0 0 0 2 1336 21 0 25.800 0 0 1 1337 61 0 29.070 0 1 3 1338 rows × 6 columns
	Y= df['charges']
Out[26]:	1 1725.55230 2 4449.46200 3 21984.47061 4 3866.85520
	1333 10600.54830 1334 2205.98080 1335 1629.83350 1336 2007.94500 1337 29141.36030
To [07].	Name: charges, Length: 1338, dtype: float64 from sklearn.model_selection import train_test_split
In [27]:	<pre>X_train, X_test, Y_train, Y_test = train_test_split(X, Y, test_size=0.2, random_state=42)</pre>
	<pre>X_train,X_test,Y_train,Y_test = train_test_split(X,Y,test_size=0.2, random_state=42) from sklearn.linear_model import LinearRegression from sklearn.svm import SVR from sklearn.ensemble import RandomForestRegressor from sklearn.ensemble import GradientBoostingRegressor</pre>
In [28]:	<pre>from sklearn.linear_model import LinearRegression from sklearn.svm import SVR from sklearn.ensemble import RandomForestRegressor from sklearn.ensemble import GradientBoostingRegressor #### Model Trainning Ir=LinearRegression() Ir.fit(X_train,Y_train)</pre>
In [28]:	<pre>from sklearn.linear_model import LinearRegression from sklearn.svm import SVR from sklearn.ensemble import RandomForestRegressor from sklearn.ensemble import GradientBoostingRegressor ### Model Trainning lr=LinearRegression()</pre>
<pre>In [28]: In [29]: In [30]:</pre>	from sklearn.linear_model import LinearRegression from sklearn.ensemble import SVR from sklearn.ensemble import RandomForestRegressor from sklearn.ensemble import GradientBoostingRegressor #### Model Trainning Ir=LinearRegression() Ir.fit(X_train,Y_train) sv=SVR() sv.fit(X_train,Y_train) rf= RandomForestRegressor() rf.fit(X_train,Y_train) gb=GradientBoostingRegressor()
<pre>In [28]: In [29]: In [30]:</pre>	from sklearn.linear_model import LinearRegression from sklearn.ensemble import RandomForestRegressor from sklearn.ensemble import GradientBoostingRegressor ### Model Trainning Ir=LinearRegression() Ir.fit(X_train, Y_train) sv=SvR() sv.fit(X_train, Y_train) fr= RandomForestRegressor() ff.fit(X_train, Y_train) gb=GradientBoostingRegressor() gb.fit(X_train, Y_train) GradientBoostingRegressor()
<pre>In [28]: In [29]: In [30]: In [31]: In [32]: In [33]:</pre>	from sklearn.linear_model import SNR from sklearn.essemble import RandomForestRegressor from sklearn.essemble import RandomForestRegressor ### Model Trainning Ir=LinearRegression() Ir.fit(X_train,Y_train) sv=SNR() sv_fit(X_train,Y_train) fr= RandomForestRegressor() fr= RandomForestRegressor() fr= RandomForestRegressor() fr= RandomForestRegressor() ### Prediction on Test Data V_predi = Ir.predict(X_test) V_pred2 = sv_predict(X_test) V_pred4 = gb_predict(X_test) V_pred4 = gb_predict(X_test) V_pred4 = gb_predict(X_test) V_pred4 = gb_predict(X_test) V_pred6 = fr_predict(X_test) V_pred6 = fr_predict(X_test) V_pred6 = fr_predict(X_test) V_pred6 = gb_predict(X_test) V_pred7 = gb_predict(X_test) V_pred8 = g
<pre>In [28]: In [29]: In [30]: In [31]:</pre>	### Model Trainning ### Model Trainning ### Prediction on Test Data ***EndertheostingRegressor() ### Prediction on Test Data ***EndertheostingRegressor() ### predict(X.test) **EndertheostingRegressor() ### predict(X.test) **EndertheostingRegressor() ### predict(X.test) **EndertheostingRegressor() #### predict(X.test) ###
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In [28]: In [29]: In [30]: In [31]: In [32]: In [33]: Out [33]: Out [35]: In [36]: In [40]: In [41]: In [42]: In [43]: In [43]: In [44]:	## Model Train on Test Data ## Model Train on Test Data ## Find Control on Test Data ## Prediction on Test Data
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