In [1]:
```
import numpy as np
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
from sklearn import preprocessing,svm
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LinearRegression
from sklearn.preprocessing import StandardScaler
```

In [2]:
```
df=pd.read_csv(r"C:\Users\sweet\Downloads\insuranceex.csv")
df
```

Out[2]:

	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
4	32	male	28.880	0	no	northwest	3866.85520
...	...	...	...	...	...	...	...
1333	50	male	30.970	3	no	northwest	10600.54830
1334	18	female	31.920	0	no	northeast	2205.98080
1335	18	female	36.850	0	no	southeast	1629.83350
1336	21	female	25.800	0	no	southwest	2007.94500
1337	61	female	29.070	0	yes	northwest	29141.36030

1338 rows × 7 columns

Data Cleaning and Preprocessing

In [3]:
```
df.head()
```

Out[3]:

	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
4	32	male	28.880	0	no	northwest	3866.85520

In [4]:
```
df.tail()
```

Out[4]:

	age	sex	bmi	children	smoker	region	charges
1333	50	male	30.97	3	no	northwest	10600.5483
1334	18	female	31.92	0	no	northeast	2205.9808
1335	18	female	36.85	0	no	southeast	1629.8335
1336	21	female	25.80	0	no	southwest	2007.9450
1337	61	female	29.07	0	yes	northwest	29141.3603

In [5]:
```
df.shape
```

Out[5]:
```
(1338, 7)
```

In [6]:
```
df.describe()
```

Out[6]:

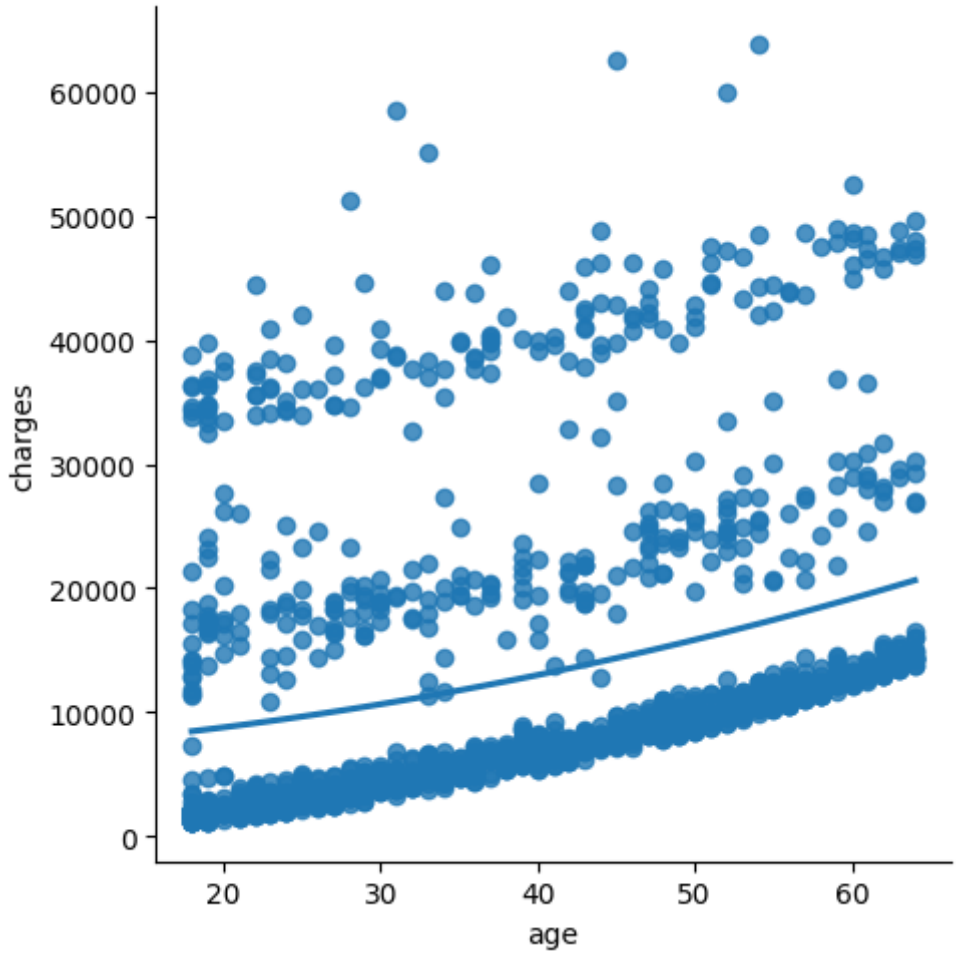
	age	bmi	children	charges
count	1338.000000	1338.000000	1338.000000	1338.000000
mean	39.207025	30.663397	1.094918	13270.422265
std	14.049960	6.098187	1.205493	12110.011237
min	18.000000	15.960000	0.000000	1121.873900
25%	27.000000	26.296250	0.000000	4740.287150
50%	39.000000	30.400000	1.000000	9382.033000
75%	51.000000	34.693750	2.000000	16639.912515
max	64.000000	53.130000	5.000000	63770.428010

In [7]:
```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1338 entries, 0 to 1337
Data columns (total 7 columns):
 #   Column    Non-Null Count  Dtype  
---  ------    --------------  -----  
 0   age       1338 non-null   int64  
 1   sex       1338 non-null   object 
 2   bmi       1338 non-null   float64
 3   children  1338 non-null   int64  
 4   smoker    1338 non-null   object 
 5   region    1338 non-null   object 
 6   charges   1338 non-null   float64
dtypes: float64(2), int64(2), object(3)
memory usage: 73.3+ KB
```

In [8]:
```
sns.lmplot(x="age",y="charges",data=df,order=2,ci=None)
```

Out[8]:
```
<seaborn.axisgrid.FacetGrid at 0x12592b72250>
```



In [9]:
```
df.fillna(method='ffill',inplace=True)
```

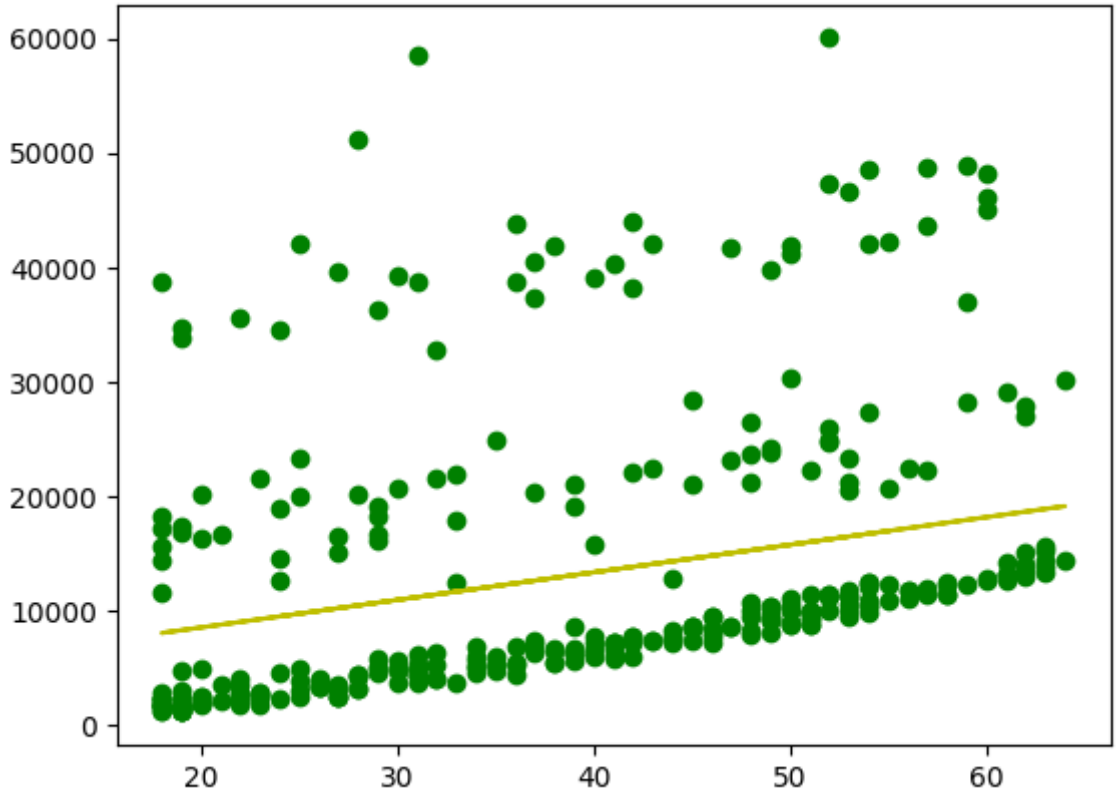
In [10]:
```
x=np.array(df['age']).reshape(-1,1)
y=np.array(df['charges']).reshape(-1,1)
```

In [11]:
```
df.dropna(inplace=True)
```

In [12]:
```
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.25)
regr=LinearRegression()
regr.fit(x_train,y_train)
print(regr.score(x_test,y_test))
```

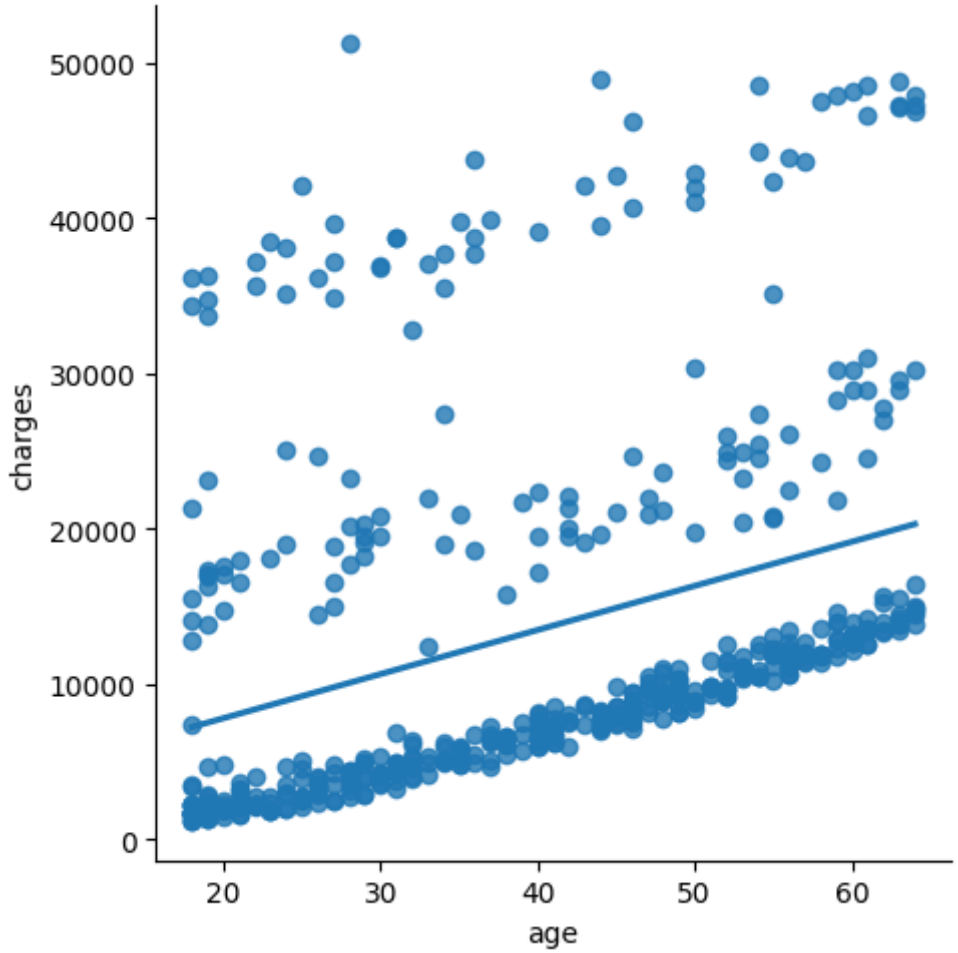
```
0.10968904282635228
```

In [13]:
```
y_pred=regr.predict(x_test)
plt.scatter(x_test,y_test,color='g')
plt.plot(x_test,y_pred,color='y')
plt.show()
```



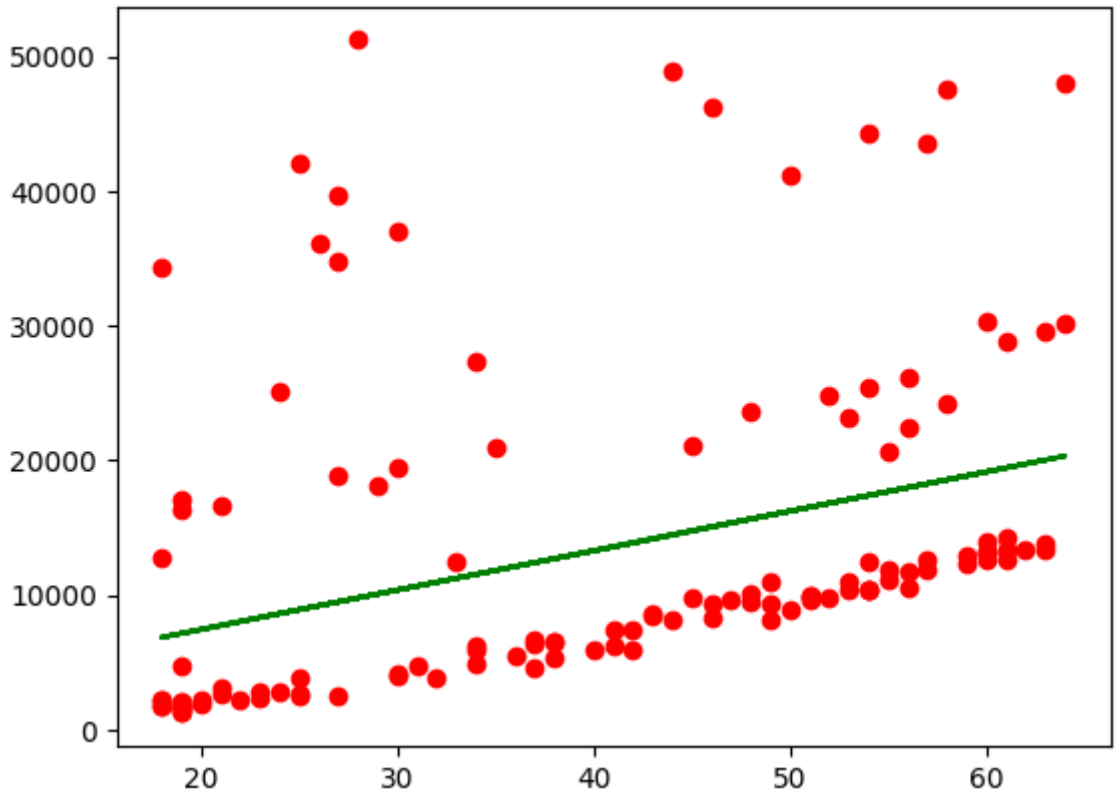
In [14]:
```
df500=df[:][:500]
sns.lmplot(x="age",y="charges",data=df500,order=1,ci=None)
```

Out[14]: <seaborn.axisgrid.FacetGrid at 0x12592c25f90>



In [15]:
```
df500.fillna(method='ffill',inplace=True)
x=np.array(df500['age']).reshape(-1,1)
y=np.array(df500['charges']).reshape(-1,1)
df500.dropna(inplace=True)
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.25)
regr=LinearRegression()
regr.fit(x_train,y_train)
print("Regression:",regr.score(x_test,y_test))
y_pred=regr.predict(x_test)
plt.scatter(x_test,y_test,color='r')
plt.plot(x_test,y_pred,color='g')
plt.show()
```

```
Regression: 0.09329384184108624
```



In [16]:
```
from sklearn.linear_model import LinearRegression
from sklearn.metrics import r2_score
model=LinearRegression()
model.fit(x_train,y_train)
y_pred=model.predict(x_test)
r2=r2_score(y_test,y_pred)
print("R2 Score:",r2)
```

```
R2 Score: 0.09329384184108624
```

In [17]:
```
df.isnull().sum()
```

Out[17]:
```
age         0
sex         0
bmi         0
children    0
smoker      0
region      0
charges     0
dtype: int64
```

Implemention of Ridgeand Lasso Regression model

In [22]:
```
from sklearn.linear_model import Lasso,Ridge
from sklearn.preprocessing import StandardScaler
```

In [23]:
```
convert={"sex":{"male":1,"female":2}}
df=df.replace(convert)
df
```

Out[23]:

	age	sex	bmi	children	smoker	region	charges
0	19	2	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
4	32	1	28.880	0	no	northwest	3866.85520
...	...	...	...	...	...	...	...
1333	50	1	30.970	3	no	northwest	10600.54830
1334	18	2	31.920	0	no	northeast	2205.98080
1335	18	2	36.850	0	no	southeast	1629.83350
1336	21	2	25.800	0	no	southwest	2007.94500
1337	61	2	29.070	0	yes	northwest	29141.36030

1338 rows × 7 columns

In [24]:
```
convert={"smoker":{"yes":1,"no":2}}
df=df.replace(convert)
df
```

Out[24]:

	age	sex	bmi	children	smoker	region	charges
0	19	2	27.900	0	1	southwest	16884.92400
1	18	1	33.770	1	2	southeast	1725.55230
2	28	1	33.000	3	2	southeast	4449.46200
3	33	1	22.705	0	2	northwest	21984.47061
4	32	1	28.880	0	2	northwest	3866.85520
...	...	...	...	...	...	...	...
1333	50	1	30.970	3	2	northwest	10600.54830
1334	18	2	31.920	0	2	northeast	2205.98080
1335	18	2	36.850	0	2	southeast	1629.83350
1336	21	2	25.800	0	2	southwest	2007.94500
1337	61	2	29.070	0	1	northwest	29141.36030

1338 rows × 7 columns

In [25]:
```
convert={"region":{"southeast":3,"southwest":4,"northeast":5,"northwest":6}}
df=df.replace(convert)
df
```

Out[25]:

	age	sex	bmi	children	smoker	region	charges
0	19	2	27.900	0	1	4	16884.92400
1	18	1	33.770	1	2	3	1725.55230
2	28	1	33.000	3	2	3	4449.46200
3	33	1	22.705	0	2	6	21984.47061
4	32	1	28.880	0	2	6	3866.85520
...	...	...	...	...	...	...	...
1333	50	1	30.970	3	2	6	10600.54830
1334	18	2	31.920	0	2	5	2205.98080
1335	18	2	36.850	0	2	3	1629.83350
1336	21	2	25.800	0	2	4	2007.94500
1337	61	2	29.070	0	1	6	29141.36030

1338 rows × 7 columns

In [27]:
```
features = df.columns[0:1]
target = df.columns[-1]
#X and y values
X = df[features].values
y = df[target].values
#splot
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3, random_state=17)
print("The dimension of x_train is {}".format(X_train.shape))
print("The dimension of x_test is {}".format(X_test.shape))
#Scale features
scaler = StandardScaler()
X_train = scaler.fit_transform(X_train)
X_test = scaler.transform(X_test)
```

```
The dimension of x_train is (936, 1)
The dimension of x_test is (402, 1)
```

In [28]:
```
ridgeReg=Ridge(alpha=10)
ridgeReg.fit(X_train,y_train)
train_score_ridge=ridgeReg.score(X_train,y_train)
test_score_ridge=ridgeReg.score(X_test,y_test)
print("\nRidge Model:\n")
print("The train score for ridge model is {}".format(train_score_ridge))
print("The test score for ridge model is {}".format(test_score_ridge))
```

```
Ridge Model:

The train score for ridge model is 0.07446228994221393
The test score for ridge model is 0.10855133360950642
```

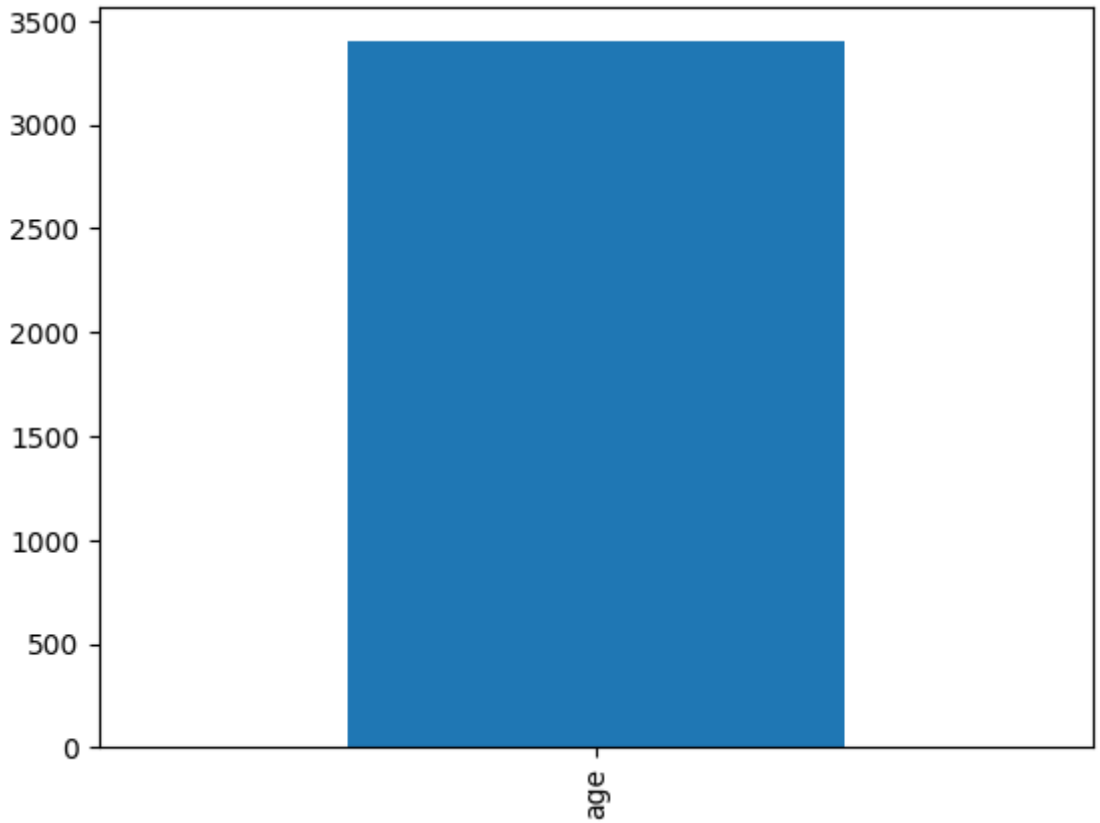
In [29]:
```
lr = LinearRegression()
#Fit model
lr.fit(X_train, y_train)
#predict
#prediction = lr.predict(X_test)
#actual
actual = y_test
train_score_lr = lr.score(X_train, y_train)
test_score_lr = lr.score(X_test, y_test)
print("\nLinear Regression Model:\n")
print("The train score for lr model is {}".format(train_score_lr))
print("The test score for lr model is {}".format(test_score_lr))
```

```
Linear Regression Model:

The train score for lr model is 0.07447061146193878
The test score for lr model is 0.10891203216512224
```

In [31]:
```
print("\nLasso Model: \n")
lasso = Lasso(alpha = 10)
lasso.fit(X_train,y_train)
train_score_ls =lasso.score(X_train,y_train)
test_score_ls =lasso.score(X_test,y_test)
print("The train score for ls model is {}".format(train_score_ls))
print("The test score for ls model is {}".format(test_score_ls))
```

```
Lasso Model:

The train score for ls model is 0.07446997086306062
The test score for ls model is 0.10881427793326703
```

In [33]:
```
pd.Series(lasso.coef_, features).sort_values(ascending = True).plot(kind = "bar")
plt.show()
```



In [38]:
```
from sklearn.linear_model import LassoCV
lasso_cv=LassoCV(alphas=[0.0001,0.001,0.01,0.1,1,10],random_state=0).fit(X_train,y_train)
print(lasso_cv.score(X_train,y_train))
print(lasso_cv.score(X_test,y_test))
```

```
0.07446997086306062
0.10881427793326703
```

In [40]:
```
plt.figure(figsize=(10,10))
#add plot for ridge regression
plt.plot(features,ridgeReg.coef_,alpha=0.7,linestyle='none',marker='*',markersize=5,color='red',label='Ridge;$\alpha=1
#add plot for lasso regression
plt.plot(lasso_cv.coef_,alpha=0.6,linestyle='none',marker='d',markersize=6,color='blue',label='Ridge;$\alpha=grid$')
#add plot for linear model
plt.plot(features,lr.coef_,alpha=0.4,linestyle='none',marker='o',markersize=7,color='green',label='Linear Regression')
#rotate axis
plt.xticks(rotation=90)
plt.legend()
plt.title("comparison plot of Ridge,Lasso and Linear regression model")
plt.show()
```

```
---------------------------------------------------------------------------
AttributeError                            Traceback (most recent call last)
Cell In[40], line 3
      1 plt.figure(figsize=(10,10))
      2 #add plot for ridge regression
----> 3 plt.plot(features,ridgeReg.coef_,alpha=0.7,linestyle='none',marker='*',markersize=5,color='red',label='Rid
ge;$\alpha=10$',border=7)
      4 #add plot for lasso regression
      5 plt.plot(lasso_cv.coef_,alpha=0.6,linestyle='none',marker='d',markersize=6,color='blue',label='Ridge;$\alp
ha=grid$')

File ~\AppData\Local\Programs\Python\Python311\Lib\site-packages\matplotlib\pyplot.py:2812, in plot(scalex, scale
y, data, *args, **kwargs)
   2810 @_copy_docstring_and_deprecators(Axes.plot)
   2811 def plot(*args, scalex=True, scaley=True, data=None, **kwargs):
-> 2812     return gca().plot(
   2813         *args, scalex=scalex, scaley=scaley,
   2814         **({"data": data} if data is not None else {}), **kwargs)
```

In [42]:
```
from sklearn.linear_model import RidgeCV
#Ridge Cross validation
ridge_cv = RidgeCV(alphas = [0.0001, 0.001,0.01, 0.1, 1, 10]).fit(X_train, y_train)
#score
print(ridge_cv.score(X_train, y_train))
print(ridge_cv.score(X_test, y_test))
```

```
0.07446228994221393
0.10855133360950775
```

Elastic net regression

In [43]:
```
from sklearn.linear_model import ElasticNet
regr=ElasticNet()
regr.fit(X,y)
print(regr.coef_)
print(regr.intercept_)
regr.score(X,y)
```

```
[257.0684655]
3191.532406056682
```

Out[43]:
```
0.08940532368214038
```

In [44]:
```
y_pred_elastic=regr.predict(X_train)
```

In [45]:
```
mean_squared_error=np.mean((y_pred_elastic-y_train)**2)
print("Mean Squared Error on test set",mean_squared_error)
```

```
Mean Squared Error on test set 267460995.25217086
```

Logistic Regression

In [46]:
```
import pandas as pd
import numpy as np
from sklearn.linear_model import LogisticRegression
from sklearn.preprocessing import StandardScaler
```

In [47]:
```
df=pd.read_csv(r"C:\Users\sweet\Downloads\insuranceex.csv")
df
```

Out[47]:

	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
4	32	male	28.880	0	no	northwest	3866.85520
...	...	...	...	...	...	...	...
1333	50	male	30.970	3	no	northwest	10600.54830
1334	18	female	31.920	0	no	northeast	2205.98080
1335	18	female	36.850	0	no	southeast	1629.83350
1336	21	female	25.800	0	no	southwest	2007.94500
1337	61	female	29.070	0	yes	northwest	29141.36030

1338 rows × 7 columns

In [48]:
```
pd.set_option('display.max_rows',10000000000)
pd.set_option('display.max_columns',10000000000)
pd.set_option('display.width',95)
```

In [49]:
```
print('This DataFrame has %d Rows and %d columns'%(df.shape))
```

```
This DataFrame has 1338 Rows and 7 columns
```

In [50]:
```
convert={"smoker":{"yes":1,"no":2}}
df=df.replace(convert)
df
```

Out[50]:

	age	sex	bmi	children	smoker	region	charges
0	19	female	27.900	0	1	southwest	16884.924000
1	18	male	33.770	1	2	southeast	1725.552300
2	28	male	33.000	3	2	southeast	4449.462000
3	33	male	22.705	0	2	northwest	21984.470610
4	32	male	28.880	0	2	northwest	3866.855200
5	31	female	25.740	0	2	southeast	3756.621600
6	46	female	33.440	1	2	southeast	8240.589600
7	37	female	27.740	3	2	northwest	7281.505600
8	37	male	29.830	2	2	northeast	6406.410700
9	60	female	25.840	0	2	northwest	28923.136920
10	25	male	26.220	0	2	northeast	2721.320800
11	62	female	26.290	0	1	southeast	27808.725100

In [51]:
```
convert={"sex":{"male":8,"female":9}}
df=df.replace(convert)
df
```

Out[51]:

	age	sex	bmi	children	smoker	region	charges
0	19	9	27.900	0	1	southwest	16884.924000
1	18	8	33.770	1	2	southeast	1725.552300
2	28	8	33.000	3	2	southeast	4449.462000
3	33	8	22.705	0	2	northwest	21984.470610
4	32	8	28.880	0	2	northwest	3866.855200
5	31	9	25.740	0	2	southeast	3756.621600
6	46	9	33.440	1	2	southeast	8240.589600
7	37	9	27.740	3	2	northwest	7281.505600
8	37	8	29.830	2	2	northeast	6406.410700
9	60	9	25.840	0	2	northwest	28923.136920
10	25	8	26.220	0	2	northeast	2721.320800
11	62	9	26.290	0	1	southeast	27808.725100

In [52]:
```
features_matrix=df.iloc[:,0:4]
```

In [53]:
```
target_vector=df.iloc[:,-3]
```

In [55]:
```
print('The Features Matrix Has %d Rows And %d Column(s)'%(features_matrix.shape))
```

```
The Features Matrix Has 1338 Rows And 4 Column(s)
```

In [57]:
```
print('The Target Matrix Has %d Rows And %d Column(s)'%(np.array(target_vector).reshape(-1,1).shape))
```

```
The Target Matrix Has 1338 Rows And 1 Column(s)
```

In [58]:
```
features_matrix_standardized=StandardScaler().fit_transform(features_matrix)
```

In [60]:
```
algorithm=LogisticRegression(penalty='l2',dual=False,tol=1e-4,C=1.0,fit_intercept=True,intercept_scaling=1,class_weigh
```

In [62]:
```
Logistic_Regression_Model=algorithm.fit(features_matrix_standardized,target_vector)
```

In [63]:
```
observation=[[1,0,0.99539,-0.05889,]]
```

In [65]:
```
predictions=Logistic_Regression_Model.predict(observation)
print('The Model Predicted The Observation To Belong To Class %s'%(predictions))
```

```
The Model Predicted The Observation To Belong To Class [2]
```

In [67]:
```
print('The algorithm was trained to predict one of the two classes: %s'%(algorithm.classes_))
```

```
The algorithm was trained to predict one of the two classes: [1 2]
```

In [85]:
```
print(" " "The model says the probability of the observation we passed belonging to class[0] Is %s" " "%(algorithm.pre
```

```
 The model says the probability of the observation we passed belonging to class[0] Is 0.1942921563693959
```

In [86]:
```
print(" " "The model says the probability of the observation we passed belonging to class['1'] Is %s" " "%(algorithm.p
```

```
 The model says the probability of the observation we passed belonging to class['1'] Is 0.1942921563693959
```

In [87]:
```
x=np.array(df['age']).reshape(-1,1)
y=np.array(df['smoker']).reshape(-1,1)
```

In [88]:
```
lerg=LogisticRegression()
lerg.fit(x,y)
print(lerg.score(x,y))
```

```
0.7952167414050823

C:\Users\sweet\AppData\Local\Programs\Python\Python311\Lib\site-packages\sklearn\utils\validation.py:1143: DataConve
rsionWarning: A column-vector y was passed when a 1d array was expected. Please change the shape of y to (n_samples,
), for example using ravel().
  y = column_or_1d(y, warn=True)
```

decision tree regression

In [89]:
```
import numpy as np
import pandas as pd
import seaborn as sns
from sklearn.model_selection import train_test_split
from sklearn.tree import DecisionTreeClassifier
```

In [90]:
```
df=pd.read_csv(r"C:\Users\sweet\Downloads\insuranceex.csv")
df
```

Out[90]:

	age	sex	bmi	children	smoker	region	charges
0	19	female	27.900	0	yes	southwest	16884.924000
1	18	male	33.770	1	no	southeast	1725.552300
2	28	male	33.000	3	no	southeast	4449.462000
3	33	male	22.705	0	no	northwest	21984.470610
4	32	male	28.880	0	no	northwest	3866.855200
5	31	female	25.740	0	no	southeast	3756.621600
6	46	female	33.440	1	no	southeast	8240.589600
7	37	female	27.740	3	no	northwest	7281.505600
8	37	male	29.830	2	no	northeast	6406.410700
9	60	female	25.840	0	no	northwest	28923.136920
10	25	male	26.220	0	no	northeast	2721.320800
11	62	female	26.290	0	yes	southeast	27808.725100

In [91]:
```
df['region'].value_counts()
```

Out[91]:
```
region
southeast    364
southwest    325
northwest    325
northeast    324
Name: count, dtype: int64
```

In [92]:
```
df['bmi'].value_counts()
```

Out[92]:
```
bmi
32.300    13
28.310     9
30.495     8
30.875     8
31.350     8
30.800     8
34.100     8
28.880     8
33.330     7
35.200     7
25.800     7
32.775     7
27.645     7
32.110     7
38.060     7
25.460     7
30.590     7
27.360     7
```

In [93]:
```
convert={"sex":{"male":1,"female":0}}
df=df.replace(convert)
df
```

175	63	0	37.700	0	yes	southwest	48824.450000
176	38	1	27.835	2	no	northwest	6455.862650
177	54	1	29.200	1	no	southwest	10436.096000
178	46	0	28.900	2	no	southwest	8823.279000
179	41	0	33.155	3	no	northeast	8538.288450
180	58	1	28.595	0	no	northwest	11735.879050
181	18	0	38.280	0	no	southeast	1631.821200
182	22	1	19.950	3	no	northeast	4005.422500
183	44	0	26.410	0	no	northwest	7419.477900
184	44	1	30.690	2	no	southeast	7731.427100
185	36	1	41.895	3	yes	northeast	43753.337050
186	26	0	29.920	2	no	southeast	3981.976800
187	30	0	30.900	3	no	southwest	5325.651000

In [94]:
```
x=["bmi","children"]
y=["yes","no"]
all_inputs=df[x]
all_classes=df["sex"]
```

In [96]:
```
(x_train,x_test,y_train,y_test)=train_test_split(all_inputs,all_classes,test_size=0.03)
```

In [97]:
```
clf=DecisionTreeClassifier(random_state=0)
```

In [98]:
```
clf.fit(x_train,y_train)
```

Out[98]:
```
DecisionTreeClassifier(random_state=0)
```
In a Jupyter environment, please rerun this cell to show the HTML representation or trust the notebook.
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In [99]:
```
score=clf.score(x_test,y_test)
print(score)
```

```
0.4878048780487805
```

random forest

In [100]:
```
import pandas as pd
import numpy as ny
import matplotlib.pyplot as plt,seaborn as sns
```

In [101]:
```
df=pd.read_csv(r"C:\Users\sweet\Downloads\insuranceex.csv")
df
```

Out[101]:

	age	sex	bmi	children	smoker	region	charges
0	19	female	27.900	0	yes	southwest	16884.924000
1	18	male	33.770	1	no	southeast	1725.552300
2	28	male	33.000	3	no	southeast	4449.462000
3	33	male	22.705	0	no	northwest	21984.470610
4	32	male	28.880	0	no	northwest	3866.855200
5	31	female	25.740	0	no	southeast	3756.621600
6	46	female	33.440	1	no	southeast	8240.589600
7	37	female	27.740	3	no	northwest	7281.505600
8	37	male	29.830	2	no	northeast	6406.410700
9	60	female	25.840	0	no	northwest	28923.136920
10	25	male	26.220	0	no	northeast	2721.320800
11	62	female	26.290	0	ves	southeast	27808.725100

In [102]:
```
df['charges'].value_counts()
```

Out[102]:
```
charges
1639.563100     2
16884.924000    1
29330.983150    1
2221.564450     1
19798.054550    1
13063.883000    1
13555.004900    1
44202.653600    1
10422.916650    1
7243.813600     1
11945.132700    1
6311.952000     1
1682.597000     1
5272.175800     1
27218.437250    1
19719.694700    1
4877.981050     1
46255.112500    1
```

In [103]:
```
m={"region":{"southeast":1,"southwest":2,"northeast":3,"northwest":4}}
df=df.replace(m)
print(df)
```

```
122    20  female  28.975         0     no       4   2257.475250
123    44    male  31.350         1    yes       3  39556.494500
124    47  female  33.915         3     no       4  10115.008850
125    26  female  28.785         0     no       3   3385.399150
126    19  female  28.300         0    yes       2  17081.080000
127    52  female  37.400         0     no       2   9634.538000
128    32  female  17.765         2    yes       4  32734.186300
129    38    male  34.700         2     no       2   6082.405000
130    59  female  26.505         0     no       3  12815.444950
131    61  female  22.040         0     no       3  13616.358600
132    53  female  35.900         2     no       2  11163.568000
133    19    male  25.555         0     no       4   1632.564450
134    20  female  28.785         0     no       3   2457.211150
135    22  female  28.050         0     no       1   2155.681500
136    19    male  34.100         0     no       2   1261.442000
137    22    male  25.175         0     no       4   2045.685250
138    54  female  31.900         3     no       1  27322.733860
139    22  female  36.000         0     no       2   2166.732000
140    34    male  22.420         2     no       3  27375.904780
141    26    male  32.490         1     no       3   3490.549100
```

In [106]:
```
df.shape
```

Out[106]:
```
(1338, 7)
```

In [107]:
```
from sklearn.ensemble import RandomForestClassifier
rfc=RandomForestClassifier()
rfc.fit(x_train,y_train)
```

Out[107]:
```
RandomForestClassifier()
```

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In [108]:
```
rf=RandomForestClassifier()
```

In [110]:
```
params={'max_depth':[2,3,5,10,20],'min_samples_leaf':[5,10,20,50,100,200],'n_estimators':[10,25,30,50,100,200]}
```

In [113]:
```
from sklearn.model_selection import GridSearchCV
grid_search=GridSearchCV(estimator=rf,param_grid=params,cv=2,scoring="accuracy")
grid_search.fit(x_train,y_train)
```

Out[113]:
```
GridSearchCV(cv=2, estimator=RandomForestClassifier(),
             param_grid={'max_depth': [2, 3, 5, 10, 20],
                         'min_samples_leaf': [5, 10, 20, 50, 100, 200],
                         'n_estimators': [10, 25, 30, 50, 100, 200]},
             scoring='accuracy')
```

In a Jupyter environment, please rerun this cell to show the HTML representation or trust the notebook.
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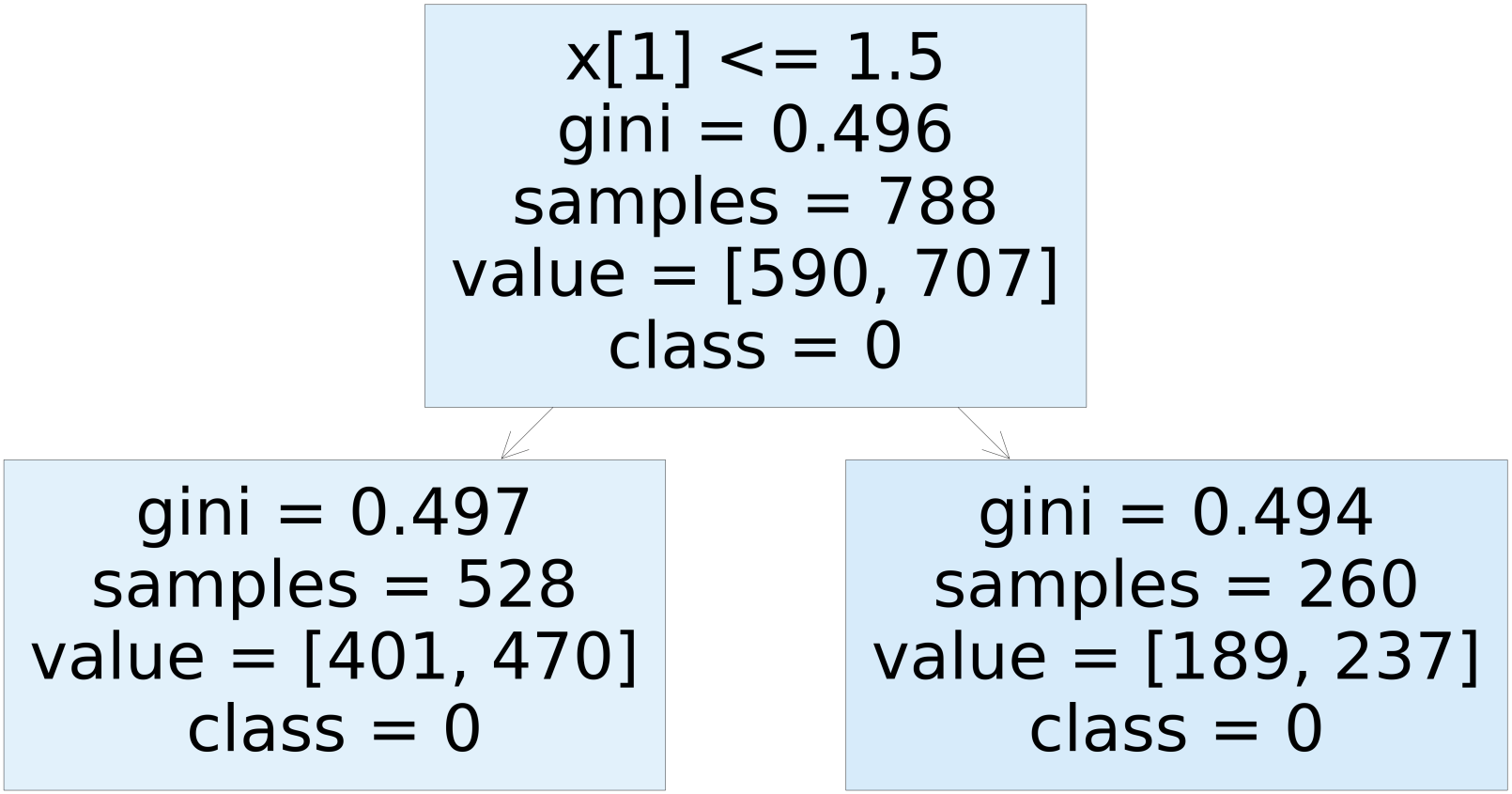
In [115]:
```
grid_search.best_score_
```

Out[115]:
```
0.5219628012707109
```

In [116]:
```
rf_best=grid_search.best_estimator_
print(rf_best)
```

```
RandomForestClassifier(max_depth=20, min_samples_leaf=200, n_estimators=10)
```

In [117]:
```
from sklearn.tree import plot_tree
plt.figure(figsize=(80,40))
plot_tree(rf_best.estimators_[4],class_names=['1','0'],filled=True);
```



In [119]:
```
rf_best.feature_importances_
```

Out[119]:
```
array([0.47848867, 0.52151133])
```

In [122]:
```
score=rfc.score(x_test,y_test)
print(score)
```

```
0.4146341463414634
```