```
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
```

Data Collection

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

In [4]: ► df.head()

Out[4]:

	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

```
▶ df.tail()
In [5]:
    Out[5]:
                                   bmi children smoker
                     age
                             sex
                                                            region
                                                                      charges
               1333
                      50
                            male 30.97
                                              3
                                                          northwest 10600.5483
               1334
                      18 female 31.92
                                                                     2205.9808
                                              0
                                                          northeast
                                                                     1629.8335
               1335
                      18 female 36.85
                                                          southeast
                      21 female 25.80
                                                                     2007.9450
               1336
                                              0
                                                         southwest
               1337
                      61 female 29.07
                                              0
                                                          northwest 29141.3603
In [6]:

▶ df.shape
    Out[6]: (1338, 7)
In [7]:

    df.describe()

    Out[7]:
                              age
                                          bmi
                                                   children
                                                                 charges
               count 1338.000000
                                   1338.000000
                                               1338.000000
                                                             1338.000000
               mean
                        39.207025
                                     30.663397
                                                  1.094918
                                                            13270.422265
                        14.049960
                                      6.098187
                 std
                                                  1.205493
                                                            12110.011237
                        18.000000
                                     15.960000
                                                  0.000000
                                                             1121.873900
                 min
                25%
                        27.000000
                                     26.296250
                                                  0.000000
                                                             4740.287150
                50%
                                                  1.000000
                        39.000000
                                     30.400000
                                                             9382.033000
                75%
                        51.000000
                                     34.693750
                                                  2.000000
                                                            16639.912515
                max
                        64.000000
                                     53.130000
                                                  5.000000 63770.428010
```

To Find Null Values

To Find Duplicate Values

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	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

1337 rows × 7 columns

```
In [11]:

▶ sns.barplot(df)
   Out[11]: <Axes: >
              14000 -
             12000
              10000 -
               8000
               6000
               4000
               2000
```

bmi

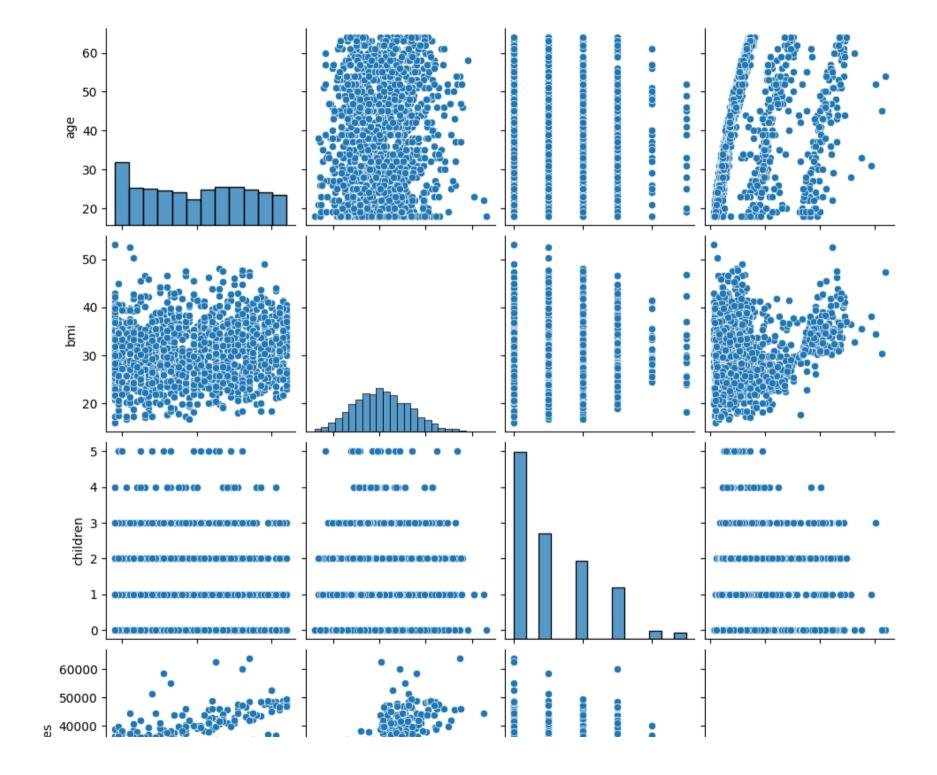
age

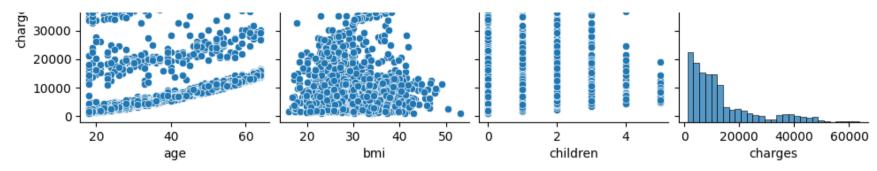
children

charges

In [12]: sns.pairplot(df)

Out[12]: <seaborn.axisgrid.PairGrid at 0x1f435bb0590>





```
M df.columns
In [13]:
   Out[13]: Index(['age', 'sex', 'bmi', 'children', 'smoker', 'region', 'charges'], dtype='object')
          T={"sex":{"female":1,"male":0}}
In [14]:
             df=df.replace(T)
             print(df)
                                bmi children smoker
                                                         region
                                                                     charges
                   age
                        sex
                          1 27.900
                                                      southwest
                                                                 16884.92400
                    19
                                            0
                                                 yes
                          0 33.770
                                                      southeast
                                                                  1725.55230
             1
                    18
                                            1
             2
                    28
                          0 33.000
                                                      southeast
                                                                  4449.46200
                                                  no
                    33
                            22.705
                                                      northwest
                                                                 21984.47061
             4
                    32
                             28.880
                                            0
                                                      northwest
                                                                  3866.85520
                                                  no
                                . . .
                                                            . . .
             1333
                             30.970
                    50
                                            3
                                                      northwest
                                                                 10600.54830
                                                  no
             1334
                          1 31.920
                                                      northeast
                                                                  2205.98080
                    18
                                                  no
             1335
                          1 36.850
                                                      southeast
                                                                  1629.83350
                    18
                                                  no
             1336
                    21
                          1 25.800
                                                      southwest
                                                                  2007.94500
             1337
                          1 29.070
                    61
                                                      northwest
                                                                 29141.36030
             [1337 rows x 7 columns]
```

Out[15]:

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

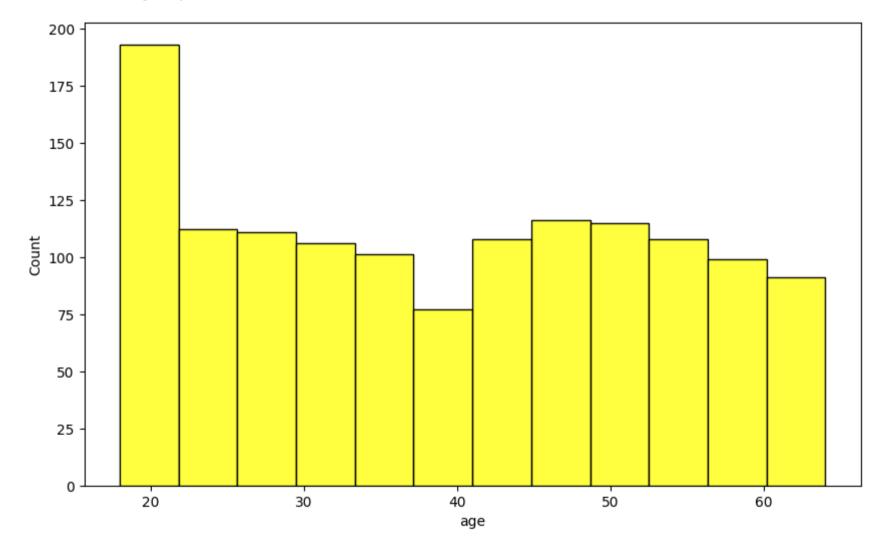
1337 rows × 7 columns

```
N | ho=df[['age', 'sex', 'bmi', 'children', 'smoker', 'charges']]
In [16]:
             plt.figure(figsize=(4,4))
             sns.heatmap(ho.corr(),annot=True)
   Out[16]: <Axes: >
                                                              - 1.0
                              0.02 0.11 0.042-0.026 0.3
                   age - 1
                                                              - 0.8
                   sex - 0.02
                                1 -0.046-0.018-0.077-0.058
                                                               0.6
                   bmi - 0.11 -0.046 1
                                        0.0130.0037 0.2
                                                              - 0.4
               children -0.042-0.0180.013
                                               0.00730.067
                smoker -- 0.026-0.0770.00370.0073
                                                     0.79
                                                              - 0.2
                         0.3 -0.058 0.2 0.067 0.79
               charges -
                                                       1
                                                               0.0
                                     bmi
                                           children
                                                 smoker
In [17]:
          x=df[['age', 'sex', 'bmi', 'children', 'smoker']]
             y=df['charges']
```

Data Visualize: Visualization The Unique Counts

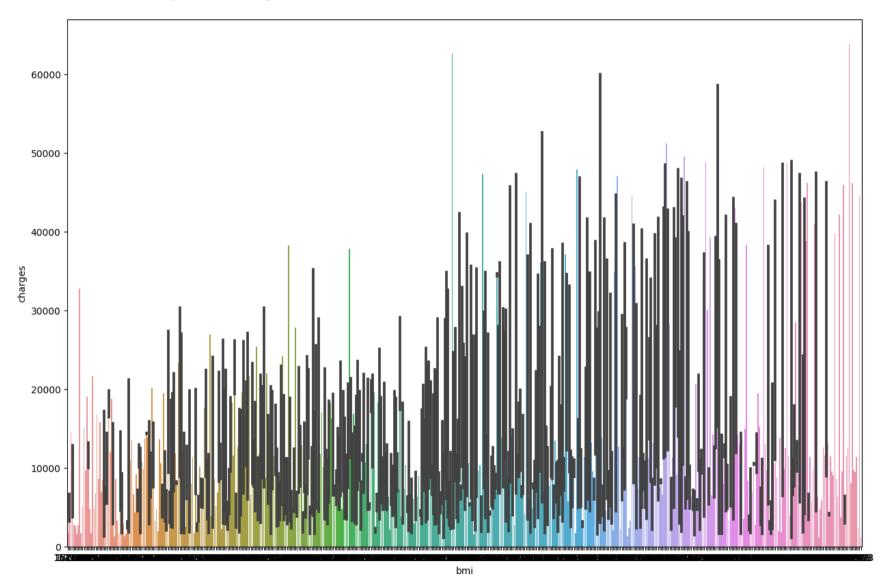
```
In [25]: In plt.figure(figsize=(10,6))
sns.histplot(data=df,x='age',color='yellow')
```

Out[25]: <Axes: xlabel='age', ylabel='Count'>



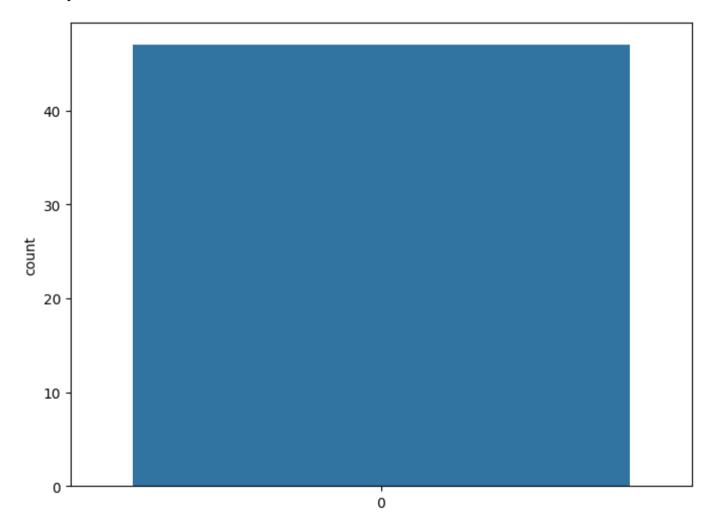
In [26]: plt.figure(figsize=(15,10))
sns.barplot(x=df.bmi,y=df.charges)

Out[26]: <Axes: xlabel='bmi', ylabel='charges'>

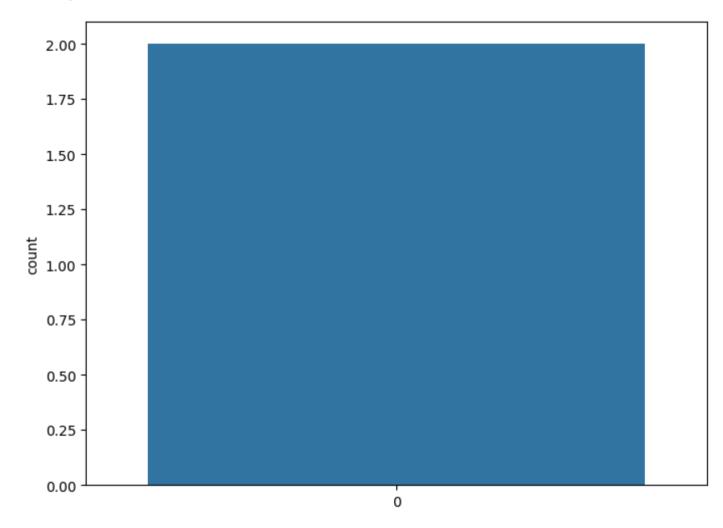


```
In [28]:  plt.figure(figsize=(8,6))
sns.countplot(df['age'].unique())
```

Out[28]: <Axes: ylabel='count'>



Out[58]: <Axes: ylabel='count'>



```
In [31]:
          Insuranced=df[['age','sex','bmi','children','smoker','charges']]
             plt.figure(figsize=(6,6))
             sns.heatmap(Insuranced.corr(),annot=True)
   Out[31]: <Axes: >
                                                                            - 1.0
               age
                             0.02
                                     0.11
                                             0.042
                     1
                                                     -0.026
                                                                0.3
                                                                            - 0.8
               Sex
                    0.02
                              1
                                     -0.046
                                             -0.018 -0.077 -0.058
                                                                            - 0.6
               bmi
                    0.11
                            -0.046
                                             0.013
                                                     0.0037
                                       1
                                                               0.2
               children
                                                                            - 0.4
                    0.042
                            -0.018
                                     0.013
                                                     0.0073
                                                              0.067
                                               1
               smoker
                                                                            - 0.2
                    -0.026
                            -0.077 0.0037 0.0073
                                                        1
                                                               0.79
               charges
```

To Find Mean And Median

-0.058

sex

0.3

age

0.2

bmi

0.067

0.79

children smoker charges

1

- 0.0

```
print(df["age"].mean(skipna=True))
In [35]:
             print(df["age"].median(skipna=True))
             39.222139117427076
             39.0
          print(df["children"].mean(skipna=True))
In [36]:
             print(df["children"].median(skipna=True))
             1.0957367240089753
             1.0
In [37]: print(df["bmi"].mean(skipna=True))
             print(df["bmi"].median(skipna=True))
             30.66345175766642
             30.4
          print(df["charges"].mean(skipna=True))
In [38]:
             print(df["charges"].median(skipna=True))
             13279.121486655948
             9386.1613
```

Logistic Regresssion

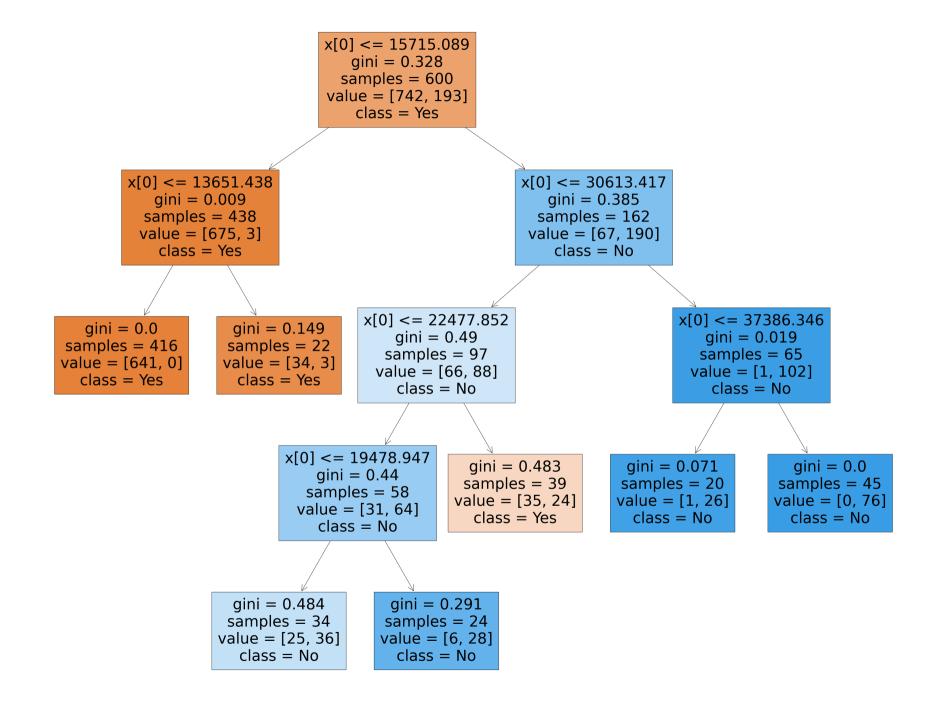
0.9253731343283582

```
In [43]:
  Out[43]: <Axes: >
         1.0 -
         0.8 -
         0.6 -
         0.4
         0.2
         0.0
                      20000
                                        50000
                 10000
                                  40000
                            30000
                                              60000
```

Decision Tree

Random Forest

```
In [49]:
          ▶ from sklearn.model selection import GridSearchCV
            grid search=GridSearchCV(estimator=rfc.param grid=params.cv=2.scoring="accuracy")
          In [50]:
            C:\Users\munigreeshma\AppData\Local\Programs\Python\Python311\Lib\site-packages\sklearn\model selection\ valid
            ation.py:686: DataConversionWarning: A column-vector v was passed when a 1d array was expected. Please change
            the shape of v to (n samples,), for example using ravel().
              estimator.fit(X train, y train, **fit params)
            C:\Users\munigreeshma\AppData\Local\Programs\Python\Python311\Lib\site-packages\sklearn\model selection\ valid
            ation.py:686: DataConversionWarning: A column-vector y was passed when a 1d array was expected. Please change
            the shape of v to (n samples,), for example using ravel().
              estimator.fit(X train, y train, **fit params)
            C:\Users\munigreeshma\AppData\Local\Programs\Python\Python311\Lib\site-packages\sklearn\model selection\ valid
            ation.py:686: DataConversionWarning: A column-vector v was passed when a 1d array was expected. Please change
            the shape of y to (n samples,), for example using ravel().
              estimator.fit(X train, y train, **fit params)
            C:\Users\munigreeshma\AppData\Local\Programs\Python\Python311\Lib\site-packages\sklearn\model selection\ valid
            ation.py:686: DataConversionWarning: 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().
              estimator.fit(X train, y train, **fit params)
            C:\Users\munigreeshma\AppData\Local\Programs\Python\Python311\Lib\site-packages\sklearn\model selection\ valid
            ation.py:686: DataConversionWarning: 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().
In [51]:
          ▶ grid search.best score
   Out[51]: 0.9219193250242501
          In [52]:
            rf best
   Out[52]:
                                      RandomForestClassifier
             RandomForestClassifier(max depth=5, min samples leaf=20, n estimators=50)
```



In [55]: N score=rfc.score(x_test,y_test)
print(score)

0.900497512437811

CONCLUSION:

Based on accuracy scores of all models that were implemented finally we can conclude that "Logistic Regression" is the best model for the given dataset