Import libraries

```python
import pandas as pd # data manipulation
import numpy as np # linear algebra
import matplotlib.pyplot as plt # plots
import seaborn as sns # plots
import warnings # Ignore warning messages
```

```python
warnings.filterwarnings("ignore")
```

Load Data

```python
df=pd.read_csv("insurance.csv")
```

Using the histogram and the box plot. We confirm the presence of outliers. So we have to give it special processing.

Technically we can give outliers the same treatment as missing values.

- **Delete** those values.
- **Replace** them with a **statistical measure** or by any other value that is in a suitable range.
- **Add** new variables, as we mentioned in the EDA.

Feature engineering

```python
class Intervals():

    def __init__(self,feature):


        self.mean=feature.mean()
        self.sd=feature.std()
        self.interval_range=[1.5,2,2.5,3.0,3.5,4]

    def Upper_Interval(self):

        for interval in self.interval_range:

            upper_interval= self.mean+interval*self.sd
            upper_interval=round(upper_interval,2)

            print(f"Interval range {interval}: {upper_interval}")

    def Lower_Interval(self):

         for interval in self.interval_range:

            lower_interval=self.mean-interval*self.sd
            lower_interval=round(lower_interval,2)

            print(f"Interval range {interval}: {lower_interval}")
```

We create a class called intervals that have functions that will allow us to make it easier to find the best ideal interval.

```python
class Best_Interval(Intervals):

    def __init__(self,feature):
```

```
        Intervals.__init__(self,feature)

    def Upper_Interval_Ideal(self):

        return super().Upper_Interval()

    def Lower_Interval_Ideal(self):

        return super().Lower_Interval()
```

We apply a technique in OOP(Object-oriented programming) to make the code easier to understand. We apply inheritance so that the parent class (Intervals) passes attributes and functions to the child class (Best_Intervals).

Selected from the best interval

In [6]:
```
bmi_interval_ideal=Best_Interval(df["bmi"])
```

Upper limit BMI

In [7]:
```
bmi_interval_ideal.Upper_Interval_Ideal()
```

```
Interval range 1.5: 39.81
Interval range 2: 42.86
Interval range 2.5: 45.91
Interval range 3.0: 48.96
Interval range 3.5: 52.01
Interval range 4: 55.06
```

With an interval of 3 it gives a good value. To be able to replace outliers.

Replaces values greater than the upper range

In [8]:
```
df["bmi"]=np.where(df["bmi"]>48.95,48.95,df["bmi"])
```

Lower limit BMI

In [9]:
```
bmi_interval_ideal.Lower_Interval_Ideal()
```

```
Interval range 1.5: 21.52
Interval range 2: 18.47
Interval range 2.5: 15.42
Interval range 3.0: 12.37
Interval range 3.5: 9.32
Interval range 4: 6.27
```
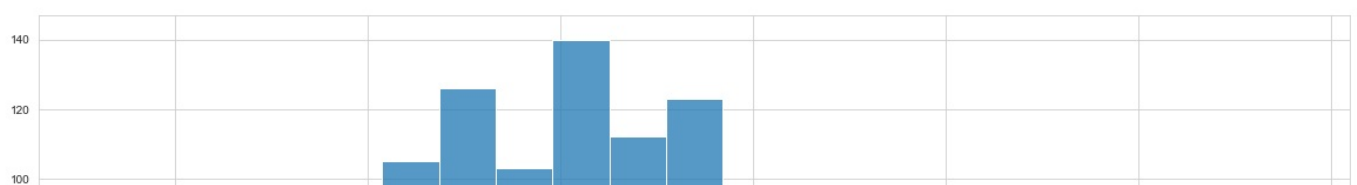
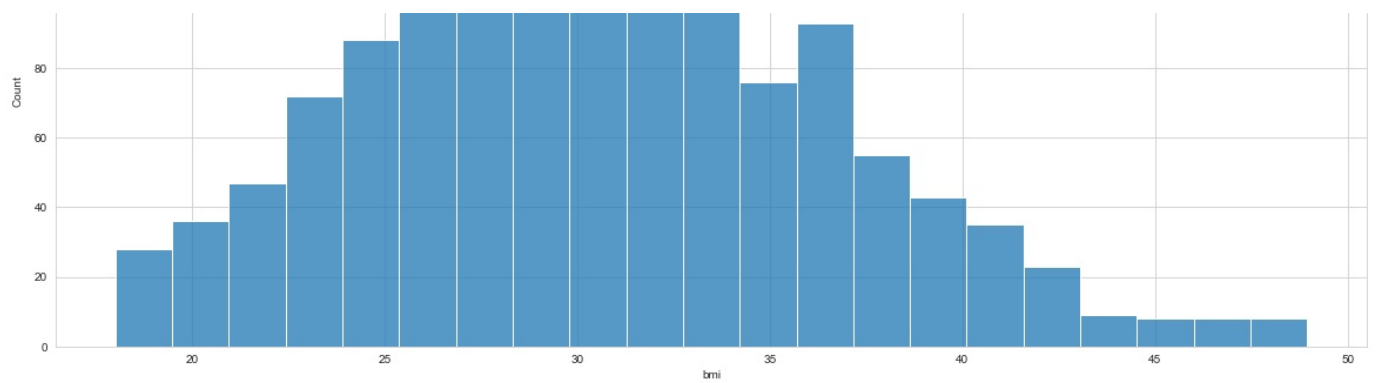The appropriate interval value is 2. Since considered with the minimum BMI values regardless of sex is 18.

Replaces values greater than the lower range

In [10]:
```
df["bmi"]=np.where(df["bmi"]<18,18,df["bmi"])
```

In [11]:
```
sns.set_style(style="whitegrid")
```

In [12]:
```
plt.subplots(1,1,figsize=(20,8))
sns.histplot(data=df,x="bmi")
plt.show()
```

Separate into smokers and non-smokers

In [13]:
```python
smoker_no_split=df.query("smoker=='no'")
smoker_yes_split=df.query("smoker=='yes'")
```

In order to give a better treatment to the data.

Smoker no split

Upper limit Charges

In [14]:
```python
charges_interval_ideal=Best_Interval(smoker_no_split["charges"])
```

In [15]:
```python
charges_interval_ideal.Upper_Interval_Ideal()
```
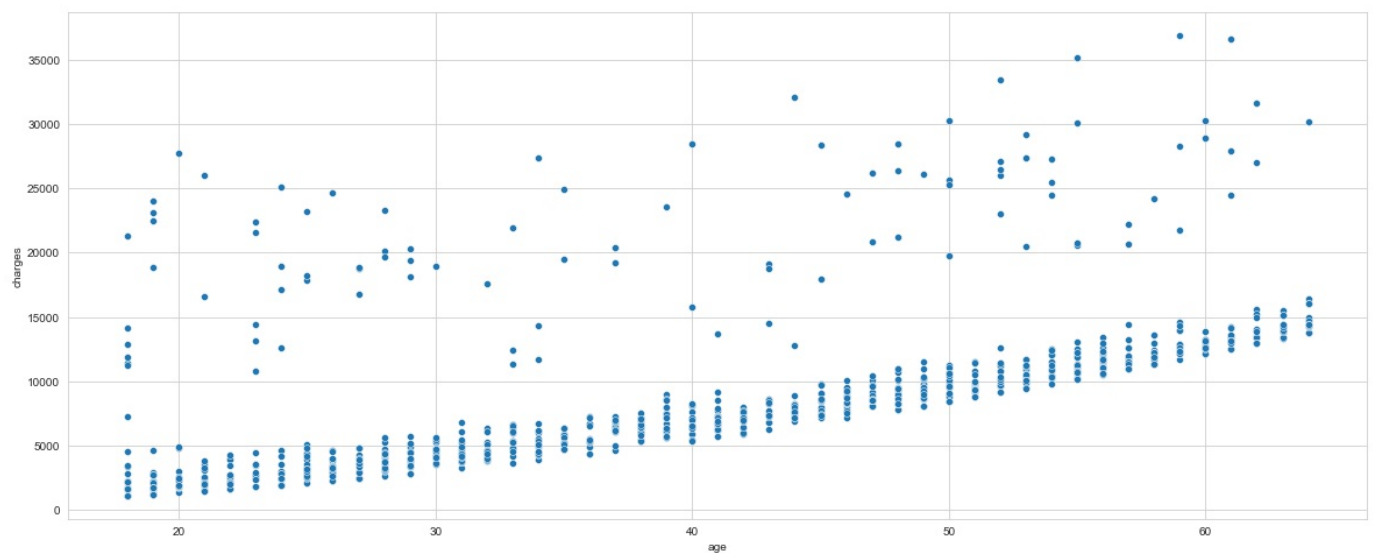
```
Interval range 1.5: 17424.94
Interval range 2: 20421.83
Interval range 2.5: 23418.72
Interval range 3.0: 26415.61
Interval range 3.5: 29412.5
Interval range 4: 32409.4
```

In [16]:
```python
def scatter_plot_age_charges(dataset):

    plt.subplots(1,1,figsize=(20,8))
    sns.scatterplot(data=dataset,x="age",y="charges")
    plt.show()
```

In [17]:
```python
scatter_plot_age_charges(dataset=smoker_no_split)
```


Using an interval of 1.5 gives good results. We will use the upper interval to group the degree of medical problem if it is severe or not.

Group according to degree of medical problem

In [18]:
```python
smoker_no_split["medical_problem"]=smoker_no_split["charges"].apply(lambda x: "severe" if
x>17000 else "light")
```
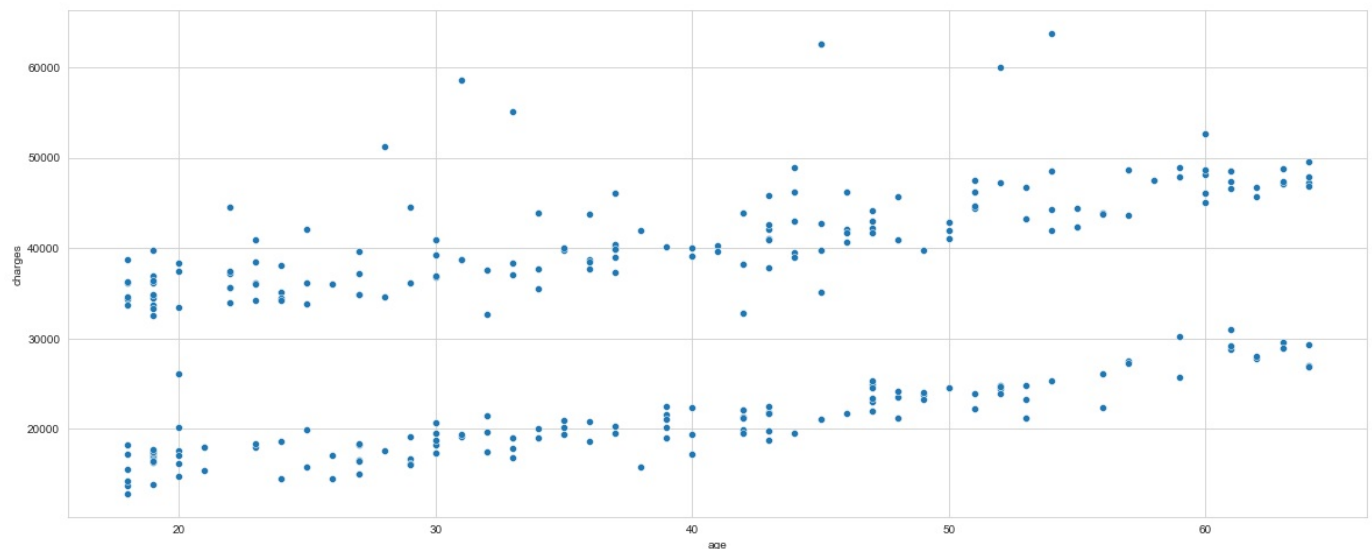
For those values greater than $17,000 US, it will classify them as **severe medical problems.**

In [19]:
```python
plt.subplots(1,1,figsize=(20,8))
sns.scatterplot(data=smoker_no_split,x="age",y="charges",hue="medical_problem",palette="Set2")
plt.show()
```
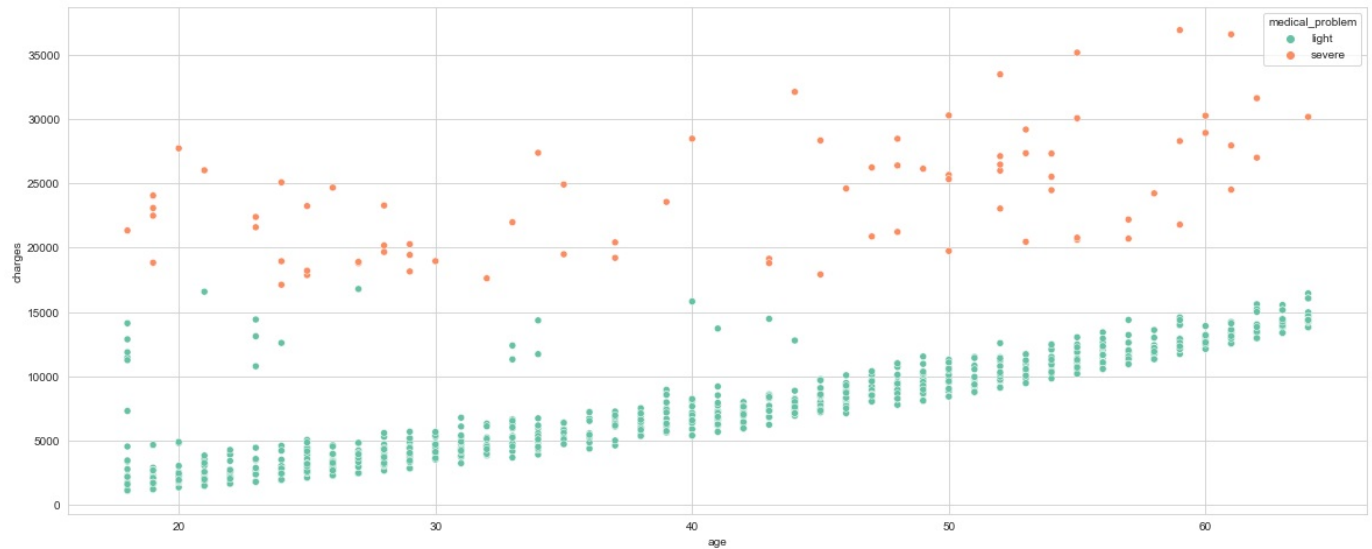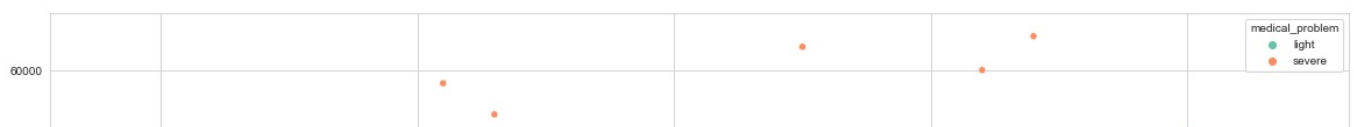

Smoker yes split

In [20]:
```python
scatter_plot_age_charges(dataset=smoker_yes_split)
```


From a range higher than $32,000 USD we could create a new group, in a similar way to non-smokers but with a different range

In [21]:
```python
smoker_yes_split["medical_problem"]=smoker_yes_split["charges"].apply(lambda x: "severe" if x >
32000 else "light")
```

In [22]:
```python
plt.subplots(1,1,figsize=(20,8))
sns.scatterplot(data=smoker_yes_split,x="age",y="charges",hue="medical_problem",palette="Set2")
plt.show()
```
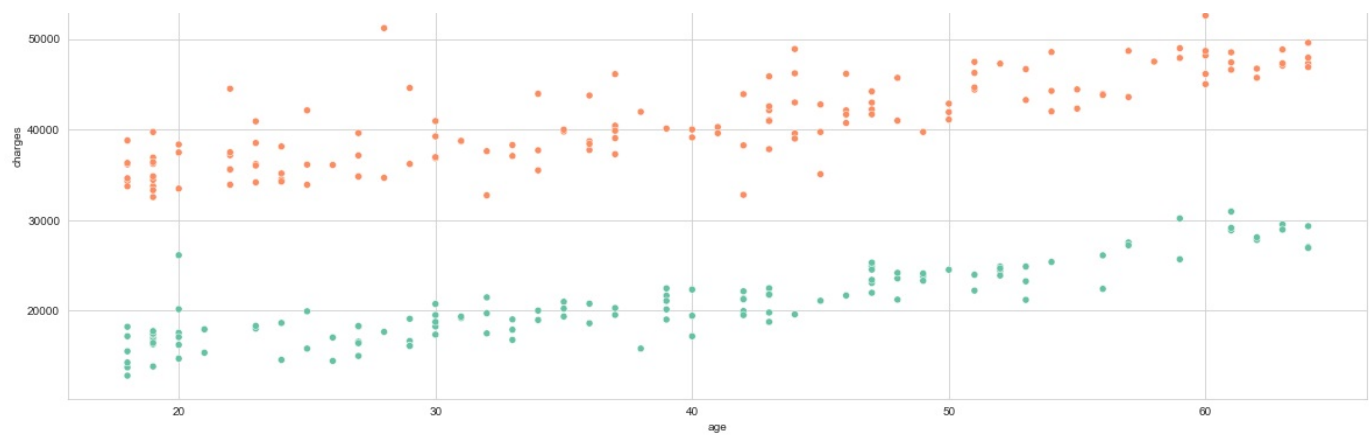
Replace Outliers Values

In [23]:
```python
smoker_yes_split["charges"]=smoker_yes_split["charges"].apply(lambda x: 48000 if x > 48000 else x)
```

We replace values greater than $48,000 and substitute them with that amount, since it is the closest value to the outliers.

We create a new dataframe with the clean data

In [24]:
```python
df_clear=pd.concat([smoker_no_split,smoker_yes_split])
```

In [25]:
```python
def boxplot():


    fig,(ax_box_1,ax_box_2)=plt.subplots(1,2,figsize=(20,8))

    ax_box_1.set_title("Adding new feature")

    sns.boxplot(data=df_clear,x="smoker",
            y="charges",
            hue="medical_problem",
            palette="Set2",
            ax=ax_box_1)

    ax_box_2.set_title("Without the new feature")
    sns.boxplot(data=df,
            x="smoker",
            y="charges",
            palette="Set2",
            ax=ax_box_2)
    plt.show()
```
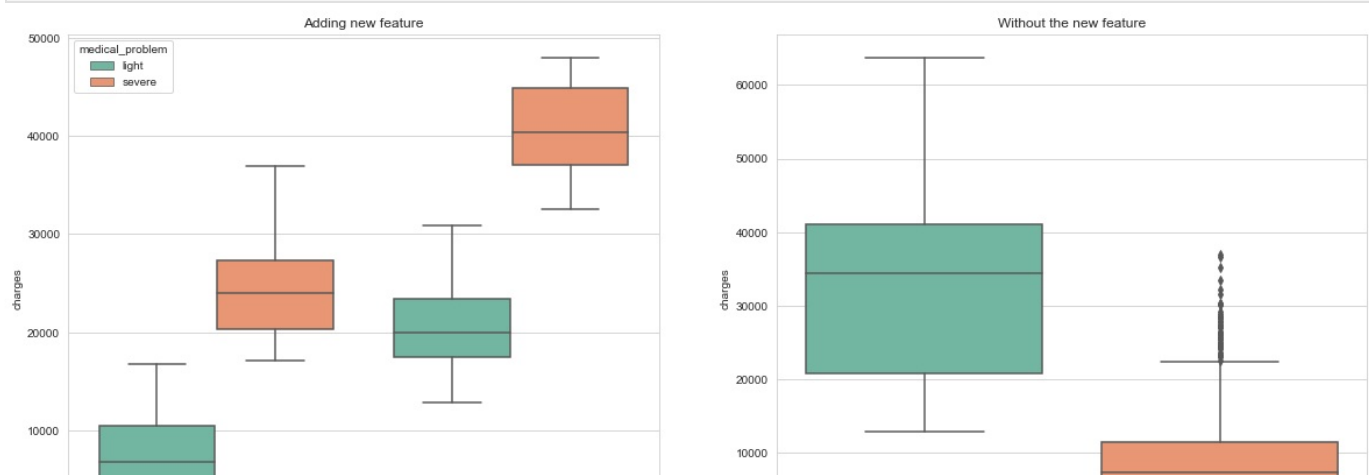
In [26]:
```python
boxplot()
```

We save the dataset with the clean data

In [27]:
```python
df_clear.to_csv("insurence_clearv2.csv",index=False)
```

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