```
In [1]:
import pandas as pd
In [2]:
df = pd.read_csv('Oxygen Dataset Final.csv')
In [3]:
df.head()
Out[3]:
   age gender spo2 pr c/nc oxy_flow
0
           0 74.0 72.0 1.0
                                  6.0
           1 NaN 110.0 NaN
                                 28.0
           0 99.0
                                 NaN
    56
                   98.0
                         1.0
    26
           1 NaN 110.0
                         1.0
                                 4.0
    52
           0 69.0 84.0 1.0
                                  0.0
In [4]:
df.tail()
Out[4]:
       age gender spo2
                         pr c/nc oxy_flow
199995
               0 99.0 NaN
199996
       48
                1 99.0 NaN
                             1.0
                                      5.0
        24
               0 99.0 110.0
                             1.0
                                     23.0
199997
199998 100
               1 99.0 95.0
                             1.0
                                     25.0
199999 22
               1 99.0 82.0 0.0
                                     32.0
In [5]:
df.shape
Out[5]:
(200000, 6)
In [6]:
df.columns
Out[6]:
Index(['age', 'gender', 'spo2', 'pr', 'c/nc', 'oxy_flow'], dtype='object')
In [7]:
df.duplicated().sum()
Out[7]:
23486
In [8]:
df = df.drop_duplicates()
In [9]:
df.isnull().sum()
Out[9]:
               0
age
gender
               0
            23115
spo2
            24769
pr
c/nc
            25191
oxy_flow
           29689
dtype: int64
```

```
df_new = df.dropna()
In [11]:
df_new.isnull().sum()
Out[11]:
             a
age
             0
gender
             0
spo2
pr
             0
c/nc
             0
oxy_flow
             0
dtype: int64
In [12]:
df_new.info()
<class 'pandas.core.frame.DataFrame'>
Int64Index: 90848 entries, 0 to 199999
Data columns (total 6 columns):
#
     Column
                Non-Null Count Dtype
0
     age
                90848 non-null
                                 int64
     gender
                90848 non-null
                                 int64
2
     spo2
                90848 non-null
                                 float64
 3
                90848 non-null
                                 float64
     pr
 4
     c/nc
                90848 non-null
                                  float64
 5
     oxy_flow 90848 non-null float64
dtypes: float64(4), int64(2)
memory usage: 4.9 MB
In [13]:
df_new.describe()
Out[13]:
                         gender
                                       spo2
                                                                 c/nc
                                                                          oxy_flow
               age
                                                      pr
                                             90848.000000
                                                         90848.000000
                                                                      90848.000000
count 90848.000000 90848.000000
                                90848.000000
                       0.332533
                                                             0.776231
mean
          46.551526
                                   87.808967
                                                91.681809
                                                                         19.125914
          21.759339
                       0.471124
                                   15.848112
                                                16.060918
                                                             0.416772
                                                                         17.933869
  std
          17.000000
                       0.000000
                                   35.000000
                                                40.000000
                                                             0.000000
                                                                          0.000000
  min
  25%
          29.000000
                       0.000000
                                   82.000000
                                                81.000000
                                                             1.000000
                                                                          7.000000
  50%
          44.000000
                       0.000000
                                   95.000000
                                                95.000000
                                                              1.000000
                                                                         16.000000
  75%
          62.000000
                       1.000000
                                   99.000000
                                               106.000000
                                                              1.000000
                                                                         25.000000
         100.000000
                                                                         76.000000
                       1.000000
                                   99.000000
                                               110.000000
                                                              1.000000
  max
In [14]:
df_new.nunique()
Out[14]:
             84
age
gender
             2
spo2
             71
pr
c/nc
              2
oxy_flow
             77
dtype: int64
In [15]:
```

In [10]:

import matplotlib.pyplot as plt

import seaborn as sns

df\_new['gender'].unique()

array([0, 1], dtype=int64)

In [16]:

Out[16]:

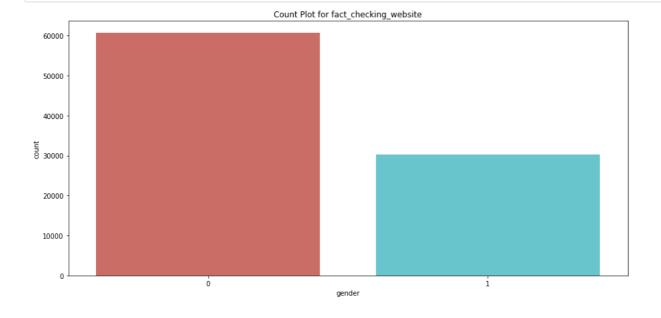
```
In [17]:
```

```
df_new['gender'].value_counts()

Out[17]:
0  60638
1  30210
Name: gender, dtype: int64

In [18]:

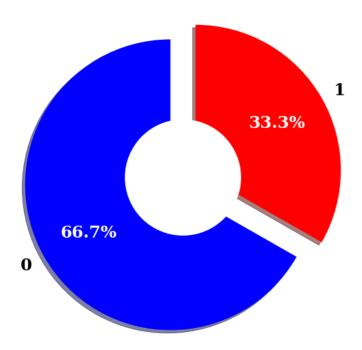
plt.figure(figsize=[15,7],)
plt.title('Count Plot for fact_checking_website')
sns.countplot(x = 'gender', data = df_new, palette = 'hls')
plt.show()
```



## In [19]:

```
gender_data = df_new['gender'].value_counts()
explode = (0.1, 0.1)
plt.figure(figsize=(14, 10))
patches, texts, pcts = plt.pie(gender_data,
                               labels = gender_data.index,
colors = ['blue', 'red'],
                                pctdistance = 0.65,
                                shadow = True,
                                startangle = 90,
                                explode = explode,
autopct = '%1.1f%%',
                               plt.setp(pcts, color='white')
hfont = {'fontname':'serif', 'weight': 'bold'}
plt.title('gender_data', size=45, **hfont)
centre_circle = plt.Circle((0,0),0.40,fc='white')
fig = plt.gcf()
fig.gca().add_artist(centre_circle)
plt.show()
```

# gender\_data



```
In [20]:

df_new['c/nc'].unique()

Out[20]:
array([1., 0.])
In [21]:
```

```
Out[21]:

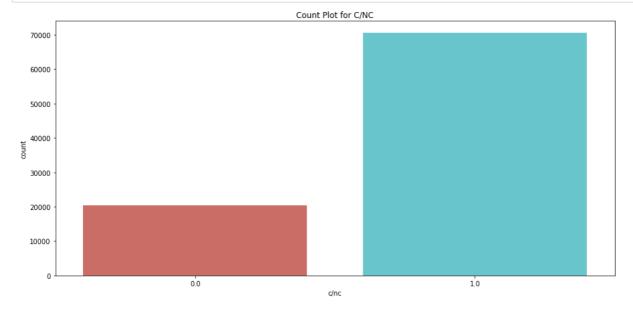
1.0 70519
0.0 20329
```

df\_new['c/nc'].value\_counts()

Name: c/nc, dtype: int64

## In [22]:

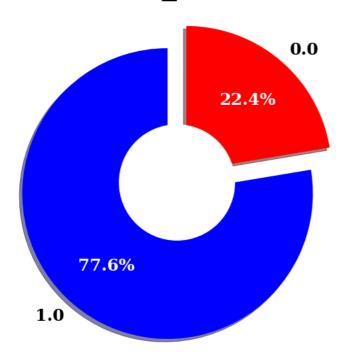
```
plt.figure(figsize=[15,7],)
plt.title('Count Plot for C/NC')
sns.countplot(x = 'c/nc', data = df_new, palette = 'hls')
plt.show()
```



## In [23]:

```
cnc_data = df_new['c/nc'].value_counts()
explode = (0.1, 0.1)
plt.figure(figsize=(14, 10))
patches, texts, pcts = plt.pie(cnc_data,
                                  labels = cnc_data.index,
colors = ['blue', 'red'],
                                  pctdistance = 0.65,
                                  shadow = True,
                                  startangle = 90,
                                  explode = explode,
autopct = '%1.1f%%',
                                  plt.setp(pcts, color='white')
hfont = {'fontname':'serif', 'weight': 'bold'}
plt.title('cnc_data', size=45, **hfont)
centre_circle = plt.Circle((0,0),0.40,fc='white')
fig = plt.gcf()
fig.gca().add_artist(centre_circle)
plt.show()
```

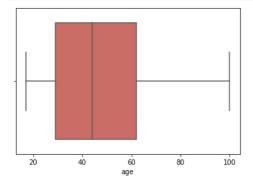
## cnc\_data

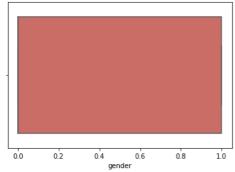


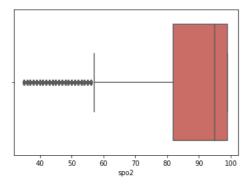
#### In [24]:

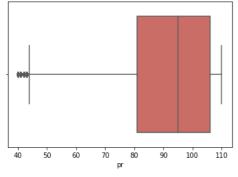
```
import warnings
warnings.filterwarnings('ignore')
```

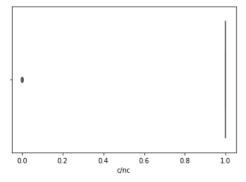
```
for i in df_new.columns:
    sns.boxplot(df_new[i], palette = 'hls')
    plt.show()
```

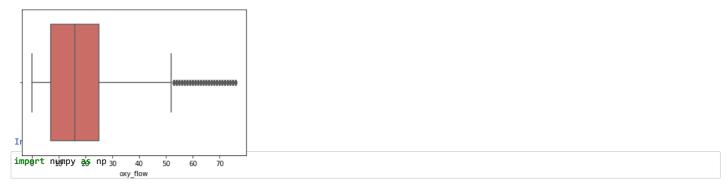












## In [27]:

```
from scipy import stats
df_new = df_new[(np.abs(stats.zscore(df_new)) < 3).all(axis=1)]
df_new.shape</pre>
```

#### Out[27]:

(83925, 6)

## In [28]:

df\_new.columns

## Out[28]:

```
Index(['age', 'gender', 'spo2', 'pr', 'c/nc', 'oxy_flow'], dtype='object')
```

## In [29]:

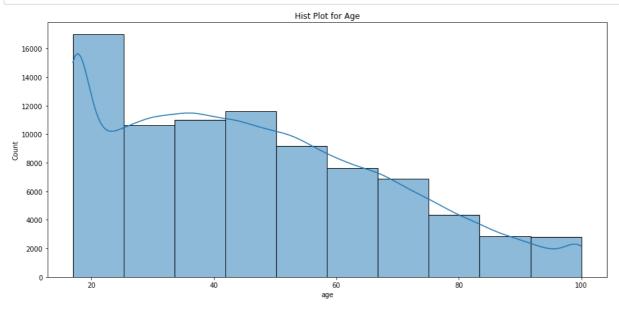
df\_new.head()

## Out[29]:

	age	gender	spo2	pr	c/nc	oxy_flow
0	27	0	74.0	72.0	1.0	6.0
4	52	0	69.0	84.0	1.0	0.0
5	82	0	93.0	93.0	1.0	28.0
9	68	0	90.0	92.0	1.0	33.0
13	40	0	99.0	109.0	1.0	27.0

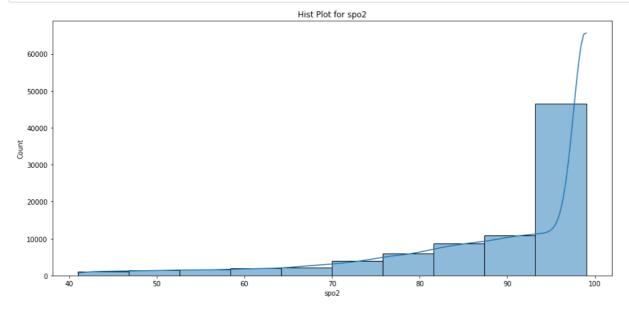
## In [30]:

```
plt.figure(figsize=[15,7],)
plt.title('Hist Plot for Age')
sns.histplot(x = 'age', data = df_new, bins = 10, kde = True, palette = 'hls')
plt.show()
```



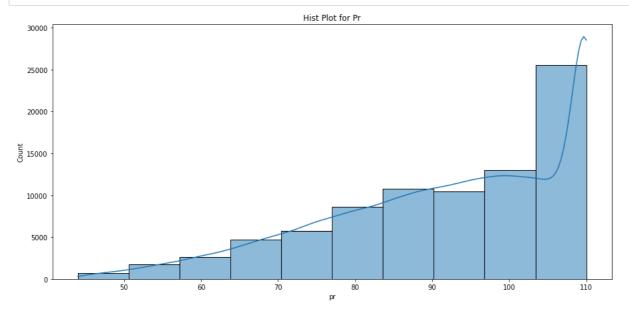
## In [31]:

```
plt.figure(figsize=[15,7],)
plt.title('Hist Plot for spo2')
sns.histplot(x = 'spo2', data = df_new, bins = 10, kde = True, palette = 'hls')
plt.show()
```



## In [32]:

```
plt.figure(figsize=[15,7],)
plt.title('Hist Plot for Pr')
sns.histplot(x = 'pr', data = df_new, bins = 10, kde = True, palette = 'hls')
plt.show()
```



#### In [33]:

```
df_corr = df_new.corr()
```

```
In [34]:
```

```
plt.figure(figsize=[15,7],)
sns.heatmap(df_corr, annot = True)
plt.show()
```



#### Tn [35]

```
from sklearn.model_selection import train_test_split

X = df_new.iloc[:, :-1]
y = df_new.iloc[:, -1]
```

#### In [36]:

from sklearn import preprocessing

## In [37]:

```
scaler = preprocessing.StandardScaler()
X = scaler.fit_transform(X)
```

## In [38]:

```
X_train, X_test, y_train, y_test = train_test_split(X, y, random_state=42, test_size=0.2)
X_train.shape, X_test.shape, y_train.shape, y_test.shape
```

## Out[38]:

```
((67140, 5), (16785, 5), (67140,), (16785,))
```

#### In [39]:

```
from sklearn.linear_model import LinearRegression
model_lr = LinearRegression()
model_lr.fit(X_train, y_train)
```

## Out[39]:

```
▼ LinearRegression
LinearRegression()
```

## In [40]:

```
y_pred = model_lr.predict(X_test)
```

## In [41]:

```
from sklearn.metrics import mean_absolute_error, mean_squared_error, r2_score
```

```
In [42]:
print("MAE",mean_absolute_error(y_test,y_pred))
print("MSE",mean_squared_error(y_test,y_pred))
print("RMSE",np.sqrt(mean_squared_error(y_test,y_pred)))
r2 = r2_score(y_test,y_pred)
print('R2 ',r2)
MAE 9.866725834475675
MSE 171.62334942211533
RMSE 13.100509510019652
R2 0.00935015491224922
In [43]:
import xgboost as xgb
xgb_r = xgb.XGBRegressor(objective ='reg:linear')
                          n_{estimators} = 10, seed = 123)
xgb_r.fit(X_train,y_train)
[14:47:07] WARNING: C:/Users/administrator/workspace/xgboost-win64_release_1.6.0/src/objective/regression_obj.cu:203: reg:l
inear is now deprecated in favor of reg:squarederror.
Out[43]:
                                     XGBRegressor
XGBRegressor(base_score=0.5, booster='gbtree', callbacks=None,
              colsample_bylevel=1, colsample_bynode=1, colsample_bytree=1,
              early_stopping_rounds=None, |enable_categorical=False,
              eval_metric=None, gamma=0, gpu_id=-1, grow_policy='depthwise',
              importance_type=None, interaction_constraints='',
              learning_rate=0.300000012, max_bin=256, max_cat_to_onehot=4,
              max_delta_step=0, max_depth=6, max_leaves=0, min_child_weight=1,
              missing=nan, monotone_constraints='()', n_estimators=10, n_jobs=0,
              num_parallel_tree=1, objective='reg:linear', predictor='auto',
              random_state=123, reg_alpha=0, ...)
In [44]:
y_pred = xgb_r.predict(X_test)
In [45]:
print("MAE",mean_absolute_error(y_test,y_pred))
print("MSE",mean_squared_error(y_test,y_pred))
print("RMSE",np.sqrt(mean_squared_error(y_test,y_pred)))
r2 = r2_score(y_test,y_pred)
print('R2 ',r2)
MAE 9.862197978644986
MSE 172.36876321877892
RMSE 13.128928487076884
R2 0.005047453300390914
```

In [46]:

In [47]:

Out[47]:

In [48]:

from sklearn.tree import DecisionTreeRegressor

regressor = DecisionTreeRegressor()

regressor.fit(X\_train, y\_train)

y\_pred = regressor.predict(X\_test)

v DecisionTreeRegressor
DecisionTreeRegressor()

## In [49]:

```
print("MAE",mean_absolute_error(y_test,y_pred))
print("MSE",mean_squared_error(y_test,y_pred))
print("RMSE",np.sqrt(mean_squared_error(y_test,y_pred)))
r2 = r2_score(y_test,y_pred)
print('R2 ',r2)
```

MAE 13.397369255364445 MSE 317.6290358851169 RMSE 17.822150147642592 R2 -0.8334285879775272

## In [50]:

#Conclusion: The data or the models used are not sufficient to predict the # oxygen flow for a patient.