

```
In [1]: import warnings
warnings.filterwarnings("ignore")
```

```
In [2]: import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LogisticRegression
```

```
In [3]: lung_data = pd.read_csv("survey_lung_cancer.csv")
```

```
In [4]: lung_data
```

Out[4]:

	GENDER	AGE	SMOKING	YELLOW_FINGERS	ANXIETY	PEER_PRESSURE	CHRONIC DISEASE	FATIGU
0	M	69	1	2	2	1	1	
1	M	74	2	1	1	1	2	
2	F	59	1	1	1	2	1	
3	M	63	2	2	2	1	1	
4	F	63	1	2	1	1	1	
...
304	F	56	1	1	1	2	2	
305	M	70	2	1	1	1	1	
306	M	58	2	1	1	1	1	
307	M	67	2	1	2	1	1	
308	M	62	1	1	1	2	1	

309 rows × 16 columns



```
In [5]: lung_data.head()
```

Out[5]:

	GENDER	AGE	SMOKING	YELLOW_FINGERS	ANXIETY	PEER_PRESSURE	CHRONIC DISEASE	FATIGUE
0	M	69	1	2	2	1	1	2
1	M	74	2	1	1	1	2	2
2	F	59	1	1	1	2	1	2
3	M	63	2	2	2	1	1	1
4	F	63	1	2	1	1	1	1



In [6]: lung_data.tail()

Out[6]:

	GENDER	AGE	SMOKING	YELLOW_FINGERS	ANXIETY	PEER_PRESSURE	CHRONIC DISEASE	FATIGU
304	F	56	1	1	1	2	2	
305	M	70	2	1	1	1	1	
306	M	58	2	1	1	1	1	
307	M	67	2	1	2	1	1	
308	M	62	1	1	1	2	1	

In [7]: *#dependent_variable*
 x = lung_data.iloc[:,0:-1]
 print(x)

	GENDER	AGE	SMOKING	YELLOW_FINGERS	ANXIETY	PEER_PRESSURE	\
0	M	69	1	2	2	1	
1	M	74	2	1	1	1	
2	F	59	1	1	1	2	
3	M	63	2	2	2	1	
4	F	63	1	2	1	1	
..	
304	F	56	1	1	1	2	
305	M	70	2	1	1	1	
306	M	58	2	1	1	1	
307	M	67	2	1	2	1	
308	M	62	1	1	1	2	

	CHRONIC DISEASE	FATIGUE	ALLERGY	WHEEZING	ALCOHOL CONSUMING	COUGHING	\
0	1	2	1	2	2	2	
1	2	2	2	1	1	1	
2	1	2	1	2	1	2	
3	1	1	1	1	2	1	
4	1	1	1	2	1	2	
..	
304	2	2	1	1	2	2	
305	1	2	2	2	2	2	
306	1	1	2	2	2	2	
307	1	2	2	1	2	2	
308	1	2	2	2	2	1	

	SHORTNESS OF BREATH	SWALLOWING DIFFICULTY	CHEST PAIN
0	2	2	2
1	2	2	2
2	2	1	2
3	1	2	2
4	2	1	1
..
304	2	2	1
305	2	1	2
306	1	1	2
307	2	1	2
308	1	2	1

[309 rows x 15 columns]

```
In [8]: #independent_variable
y = lung_data. iloc[:,-1:]
print(y)
```

```
      LUNG_CANCER
0             YES
1             YES
2             NO
3             NO
4             NO
..          ...
304          YES
305          YES
306          YES
307          YES
308          YES

[309 rows x 1 columns]
```

```
In [9]: lung_data.GENDER = lung_data.GENDER.map({"M":1,"F":2})
lung_data.LUNG_CANCER = lung_data.LUNG_CANCER.map({"YES":1,"NO":2})
```

```
In [10]: lung_data.shape
```

```
Out[10]: (309, 16)
```

```
In [11]: lung_data.isnull().sum()
```

```
Out[11]: GENDER          0
AGE          0
SMOKING      0
YELLOW_FINGERS  0
ANXIETY      0
PEER_PRESSURE  0
CHRONIC DISEASE  0
FATIGUE      0
ALLERGY      0
WHEEZING     0
ALCOHOL CONSUMING  0
COUGHING     0
SHORTNESS OF BREATH  0
SWALLOWING DIFFICULTY  0
CHEST PAIN   0
LUNG_CANCER  0
dtype: int64
```


In [12]: lung_data.dtypes

```
Out[12]: GENDER          int64
AGE          int64
SMOKING      int64
YELLOW_FINGERS int64
ANXIETY      int64
PEER_PRESSURE int64
CHRONIC DISEASE int64
FATIGUE      int64
ALLERGY      int64
WHEEZING     int64
ALCOHOL CONSUMING int64
COUGHING     int64
SHORTNESS OF BREATH int64
SWALLOWING DIFFICULTY int64
CHEST PAIN   int64
LUNG_CANCER  int64
dtype: object
```

In [13]: lung_data.head()

```
Out[13]:
```

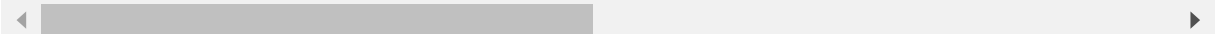
	GENDER	AGE	SMOKING	YELLOW_FINGERS	ANXIETY	PEER_PRESSURE	CHRONIC DISEASE	FATIGUE
0	1	69	1	2	2	1	1	2
1	1	74	2	1	1	1	2	2
2	2	59	1	1	1	2	1	2
3	1	63	2	2	2	1	1	1
4	2	63	1	2	1	1	1	1



In [14]: lung_data.tail()

```
Out[14]:
```

	GENDER	AGE	SMOKING	YELLOW_FINGERS	ANXIETY	PEER_PRESSURE	CHRONIC DISEASE	FATIGU
304	2	56	1	1	1	2	2	
305	1	70	2	1	1	1	1	
306	1	58	2	1	1	1	1	
307	1	67	2	1	2	1	1	
308	1	62	1	1	1	2	1	



In [15]: *#the describe() method returns description of data in DataFrame*
 lung_data.describe()

Out[15]:

	GENDER	AGE	SMOKING	YELLOW_FINGERS	ANXIETY	PEER_PRESSURE	CHR DIS
count	309.000000	309.000000	309.000000	309.000000	309.000000	309.000000	309.000000
mean	1.475728	62.673139	1.563107	1.569579	1.498382	1.501618	1.501618
std	0.500221	8.210301	0.496806	0.495938	0.500808	0.500808	0.500808
min	1.000000	21.000000	1.000000	1.000000	1.000000	1.000000	1.000000
25%	1.000000	57.000000	1.000000	1.000000	1.000000	1.000000	1.000000
50%	1.000000	62.000000	2.000000	2.000000	1.000000	2.000000	2.000000
75%	2.000000	69.000000	2.000000	2.000000	2.000000	2.000000	2.000000
max	2.000000	87.000000	2.000000	2.000000	2.000000	2.000000	2.000000

In [16]: *#the info() method prints information of the database*
 lung_data.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 309 entries, 0 to 308
Data columns (total 16 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   GENDER                                309 non-null    int64
1   AGE                                   309 non-null    int64
2   SMOKING                               309 non-null    int64
3   YELLOW_FINGERS                       309 non-null    int64
4   ANXIETY                               309 non-null    int64
5   PEER_PRESSURE                        309 non-null    int64
6   CHRONIC DISEASE                      309 non-null    int64
7   FATIGUE                              309 non-null    int64
8   ALLERGY                              309 non-null    int64
9   WHEEZING                             309 non-null    int64
10  ALCOHOL CONSUMING                    309 non-null    int64
11  COUGHING                             309 non-null    int64
12  SHORTNESS OF BREATH                  309 non-null    int64
13  SWALLOWING DIFFICULTY                309 non-null    int64
14  CHEST PAIN                           309 non-null    int64
15  LUNG_CANCER                          309 non-null    int64
dtypes: int64(16)
memory usage: 38.8 KB
```

In [17]: *#Splitting the Dataset: Training and Testing*
 from sklearn.model_selection import train_test_split
 x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=1/3,random_state=0)

In [18]: lung_data['LUNG_CANCER'].value_counts()

Out[18]:

```
1    270
2     39
Name: LUNG_CANCER, dtype: int64
```

```
In [19]: len(lung_data)
```

```
Out[19]: 309
```

```
In [20]: len(x_test)
```

```
Out[20]: 103
```

```
In [21]: len(x_train)
```

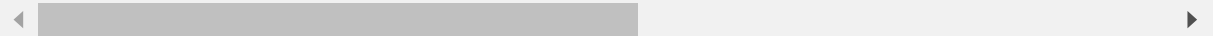
```
Out[21]: 206
```

```
In [22]: #dependent_variable  
x = lung_data.iloc[:,0:-1]  
x
```

```
Out[22]:
```

	GENDER	AGE	SMOKING	YELLOW_FINGERS	ANXIETY	PEER_PRESSURE	CHRONIC DISEASE	FATIGU
0	1	69	1	2	2	1	1	
1	1	74	2	1	1	1	2	
2	2	59	1	1	1	2	1	
3	1	63	2	2	2	1	1	
4	2	63	1	2	1	1	1	
...	
304	2	56	1	1	1	2	2	
305	1	70	2	1	1	1	1	
306	1	58	2	1	1	1	1	
307	1	67	2	1	2	1	1	
308	1	62	1	1	1	2	1	

309 rows × 15 columns



```
In [23]: #independent_variable
y = lung_data.iloc[:,-1:]
y
```

Out[23]:

LUNG_CANCER	
0	1
1	1
2	2
3	2
4	2
...	...
304	1
305	1
306	1
307	1
308	1

309 rows × 1 columns

```
In [24]: from sklearn.linear_model import LogisticRegression
from sklearn.metrics import confusion_matrix
from sklearn.metrics import accuracy_score
from sklearn.metrics import precision_score
from sklearn.metrics import recall_score
from sklearn.metrics import f1_score
```

```
In [26]: from sklearn.linear_model import LogisticRegression
x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=1/3,random_state=0)
```

Logistic Regression

```
In [27]: #Fitting simple linear regression to the training test
Model1 = LogisticRegression()
Model1.fit(x_train, y_train)
#Predicting the test set results
prediction1 = Model1.predict(x_test)
```

```
In [28]: prediction1
```

```
Out[28]: array([1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 2, 1, 2, 1, 1,
1, 1, 2, 2, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1,
1, 2, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1,
1, 1, 1, 1, 1, 1, 1, 1, 2, 1, 1, 1, 1, 1, 2, 1, 1, 1, 1, 1, 1, 1,
1, 1, 1, 1, 1, 2, 1, 1, 1, 1, 1, 1, 1, 1, 1], dtype=int64)
```

```
In [29]: from sklearn.metrics import confusion_matrix
from sklearn.metrics import accuracy_score
confusion_matrix(y_test, prediction1)
```

```
Out[29]: array([[85,  2],
               [10,  6]], dtype=int64)
```

```
In [30]: accuracy_score(y_test, prediction1)
```

```
Out[30]: 0.883495145631068
```

```
In [31]: from sklearn.metrics import precision_score
probs = Model1.predict_proba(x_test)
precision_score(y_test, prediction1, average = None)
```

```
Out[31]: array([0.89473684, 0.75      ])
```

```
In [32]: from sklearn.metrics import precision_score, recall_score, f1_score

# assuming your predicted and actual labels are stored in variables y_pred and y_
accuracy = accuracy_score(y_test, prediction1)
precision = precision_score(y_test, prediction1)
recall = recall_score(y_test, prediction1)
f1 = f1_score(y_test, prediction1)

print("Accuracy:", accuracy)
print("Precision:", precision)
print("Recall:", recall)
print("F1 score:", f1)
```

```
Accuracy: 0.883495145631068
Precision: 0.8947368421052632
Recall: 0.9770114942528736
F1 score: 0.9340659340659342
```

```
In [33]: from sklearn.metrics import recall_score
from sklearn.metrics import f1_score
```

```
In [34]: recall_score(y_test, prediction1, average = None)
```

```
Out[34]: array([0.97701149, 0.375      ])
```

```
In [35]: f1_score(y_test, prediction1, average = None)
```

```
Out[35]: array([0.93406593, 0.5       ])
```



```
In [36]: cm = confusion_matrix(y_true = y_test, y_pred = prediction1)
#plot_confusion_matrix(cm, level, title = "confusion_matrix")
sns.heatmap(cm, annot=True, cmap="Blues", fmt="d")
plt.xlabel("Predicted Labels")
plt.ylabel("True Labels")
plt.title("Confusion Matrix")
plt.show()
```

```
In [44]: from sklearn.metrics import confusion_matrix
from sklearn.metrics import accuracy_score
confusion_matrix(y_test, prediction2)
```

```
Out[44]: array([[86,  1],
               [12,  4]], dtype=int64)
```

```
In [45]: from sklearn.metrics import precision_score, recall_score, f1_score

# assuming your predicted and actual labels are stored in variables y_pred and y_
accuracy = accuracy_score(y_test, prediction2)
precision = precision_score(y_test, prediction2)
recall = recall_score(y_test, prediction2)
f1 = f1_score(y_test, prediction2)

print("Accuracy:", accuracy)
print("Precision:", precision)
print("Recall:", recall)
print("F1 score:", f1)
```

```
Accuracy: 0.8737864077669902
Precision: 0.8775510204081632
Recall: 0.9885057471264368
F1 score: 0.9297297297297297
```

```
In [46]: accuracy_score(y_test, prediction2)
```

```
Out[46]: 0.8737864077669902
```

```
In [47]: probs = Model1.predict_proba(x_test)
precision_score(y_test, prediction2, average = None)
```

```
Out[47]: array([0.87755102, 0.8       ])
```

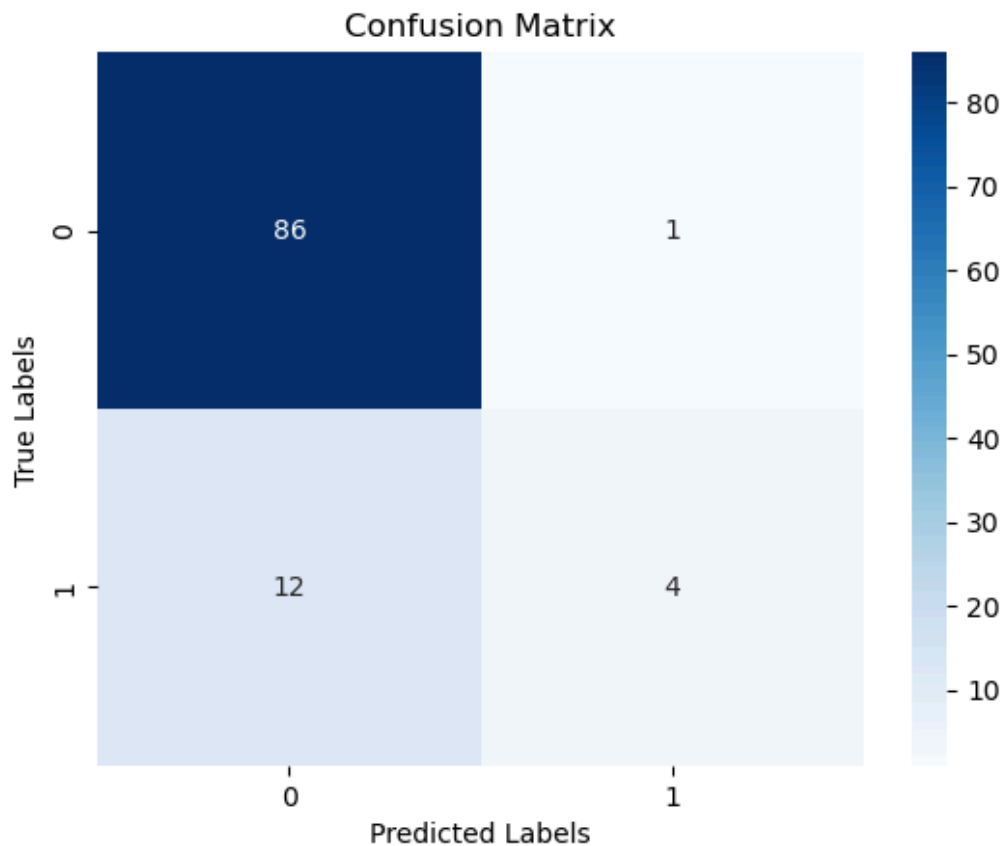
```
In [48]: recall_score(y_test, prediction2, average = None)
```

```
Out[48]: array([0.98850575, 0.25       ])
```

```
In [49]: f1_score(y_test, prediction2, average = None)
```

```
Out[49]: array([0.92972973, 0.38095238])
```

```
In [50]: cm = confusion_matrix(y_true = y_test, y_pred = prediction2)
#plot_confusion_matrix(cm, level, title = "confusion_matrix")
sns.heatmap(cm, annot=True, cmap="Blues", fmt="d")
plt.xlabel("Predicted Labels")
plt.ylabel("True Labels")
plt.title("Confusion Matrix")
plt.show()
```



Decision Tree

```
In [51]: #Decision Tree
from sklearn.tree import DecisionTreeClassifier
tree = DecisionTreeClassifier(random_state = 0, criterion = "entropy")
tree.fit(x_train, y_train)
prediction3 = classifier.predict(x_test)
```

```
In [52]: prediction3
```

```
Out[52]: array([1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 2, 1, 1, 1,
1, 1, 2, 2, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1,
1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1,
1, 1, 1, 1, 1, 1, 1, 1, 2, 1, 1, 1, 1, 1, 1, 1, 1, 2, 1, 1, 1, 1,
1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1], dtype=int64)
```

```
In [53]: confusion_matrix(y_test, prediction3)
```

```
Out[53]: array([[86,  1],
[12,  4]], dtype=int64)
```

```
In [54]: from sklearn.metrics import precision_score, recall_score, f1_score

# assuming your predicted and actual labels are stored in variables y_pred and y_
accuracy = accuracy_score(y_test, prediction3)
precision = precision_score(y_test, prediction3)
recall = recall_score(y_test, prediction3)
f1 = f1_score(y_test, prediction3)

print("Accuracy:", accuracy)
print("Precision:", precision)
print("Recall:", recall)
print("F1 score:", f1)

Accuracy: 0.8737864077669902
Precision: 0.8775510204081632
Recall: 0.9885057471264368
F1 score: 0.9297297297297297
```

```
In [55]: accuracy_score(y_test, prediction3)
```

```
Out[55]: 0.8737864077669902
```

```
In [56]: probs = Model1.predict_proba(x_test)
precision_score(y_test, prediction3, average = None)
```

```
Out[56]: array([0.87755102, 0.8       ])
```

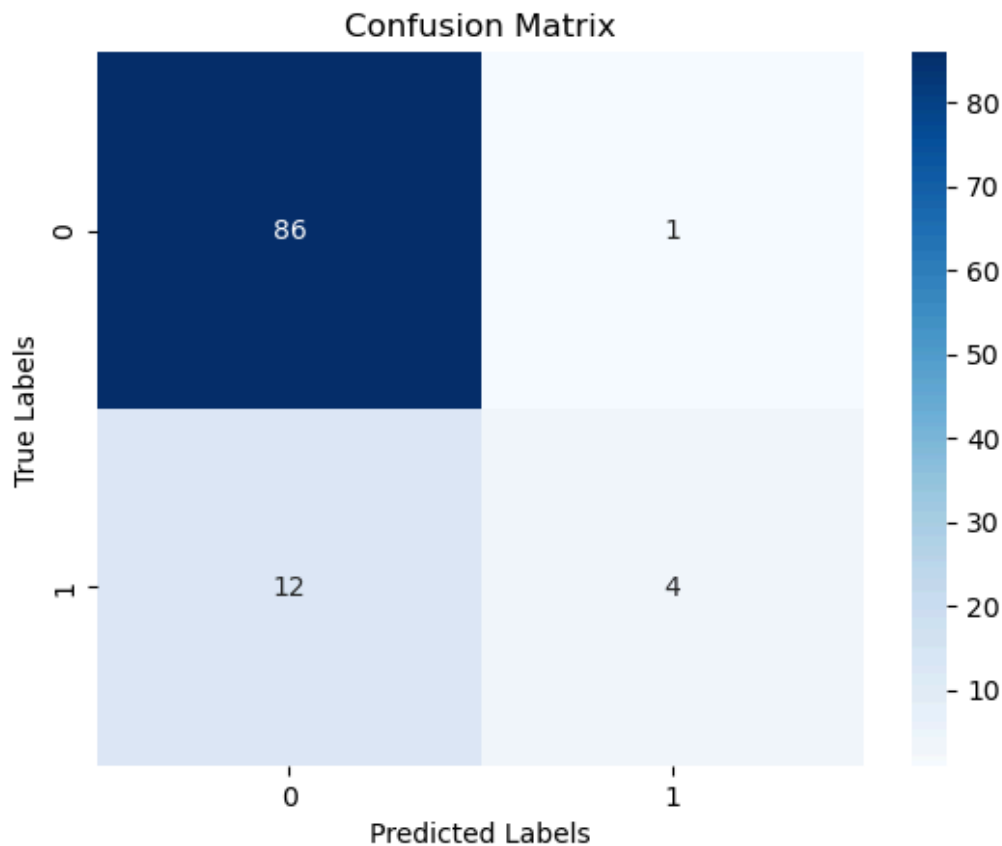
```
In [57]: recall_score(y_test, prediction3, average = None)
```

```
Out[57]: array([0.98850575, 0.25       ])
```

```
In [58]: f1_score(y_test, prediction3, average = None)
```

```
Out[58]: array([0.92972973, 0.38095238])
```

```
In [59]: cm = confusion_matrix(y_true = y_test, y_pred = prediction3)
#plot_confusion_matrix(cm, level, title = "confusion_matrix")
sns.heatmap(cm, annot=True, cmap="Blues", fmt="d")
plt.xlabel("Predicted Labels")
plt.ylabel("True Labels")
plt.title("Confusion Matrix")
plt.show()
```



Support Vector Machine

```
In [60]: #Support Vector Machine
from sklearn.ensemble import BaggingClassifier
from sklearn.multiclass import OneVsRestClassifier
from sklearn.svm import SVC
svm = OneVsRestClassifier(BaggingClassifier(SVC(C=10, kernel='rbf', random_state=9),
svm.fit(x_train, y_train)
prediction4 = svm.predict(x_test)
```

```
In [61]: confusion_matrix(y_test, prediction4)
```

```
Out[61]: array([[87,  0],
               [16,  0]], dtype=int64)
```

```
In [62]: from sklearn.metrics import precision_score, recall_score, f1_score

# assuming your predicted and actual labels are stored in variables y_pred and y_
accuracy = accuracy_score(y_test, prediction4)
precision = precision_score(y_test, prediction4)
recall = recall_score(y_test, prediction4)
f1 = f1_score(y_test, prediction4)

print("Accuracy:", accuracy)
print("Precision:", precision)
print("Recall:", recall)
print("F1 score:", f1)
```

```
Accuracy: 0.8446601941747572
Precision: 0.8446601941747572
Recall: 1.0
F1 score: 0.9157894736842105
```

```
In [63]: accuracy_score(y_test, prediction4)
```

```
Out[63]: 0.8446601941747572
```

```
In [64]: probs = Model1.predict_proba(x_test)
precision_score(y_test, prediction4, average = None)
```

```
Out[64]: array([0.84466019, 0.          ])
```

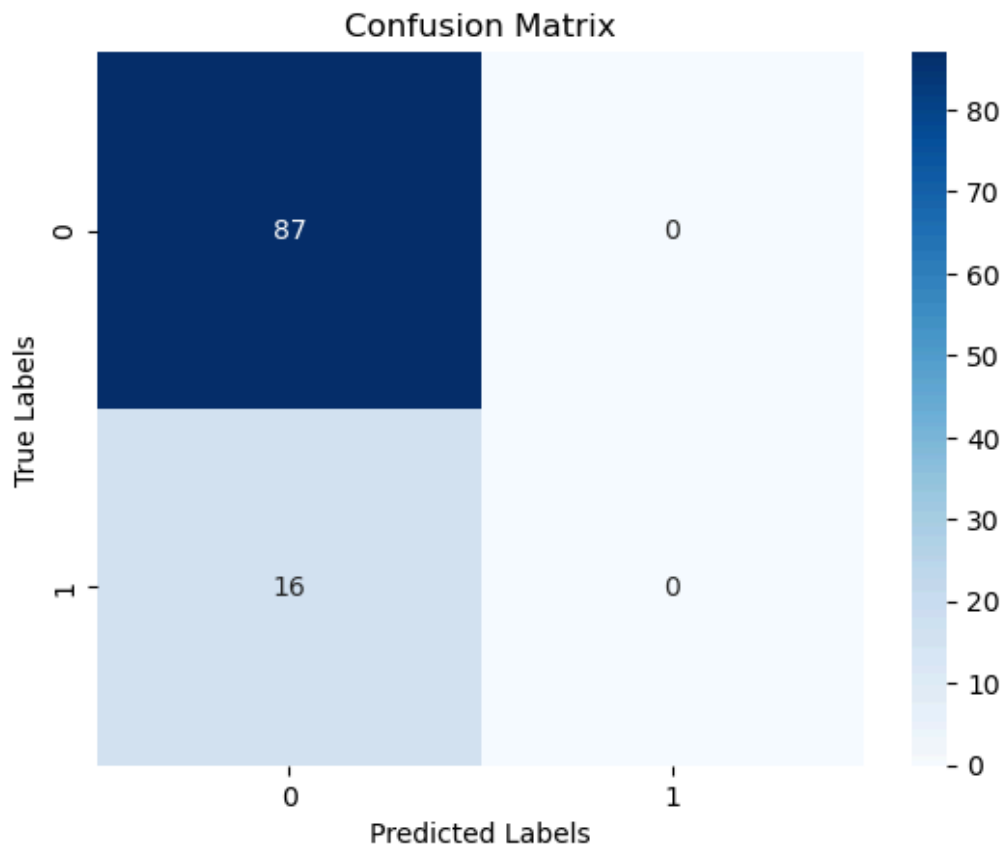
```
In [65]: recall_score(y_test, prediction4, average = None)
```

```
Out[65]: array([1., 0.])
```

```
In [66]: f1_score(y_test, prediction4, average = None)
```

```
Out[66]: array([0.91578947, 0.          ])
```

```
In [67]: cm = confusion_matrix(y_true = y_test, y_pred = prediction4)
#plot_confusion_matrix(cm, level, title = "confusion_matrix")
sns.heatmap(cm, annot=True, cmap="Blues", fmt="d")
plt.xlabel("Predicted Labels")
plt.ylabel("True Labels")
plt.title("Confusion Matrix")
plt.show()
```



Navie Bayes

```
In [68]: from sklearn.naive_bayes import GaussianNB
nbcla = GaussianNB()
nbcla.fit(x_train, y_train)
prediction5 = nbcla.predict(x_test)
```

```
In [69]: from sklearn.metrics import confusion_matrix
from sklearn.metrics import accuracy_score
confusion_matrix(y_test, prediction5)
```

```
Out[69]: array([[81,  6],
               [ 8,  8]], dtype=int64)
```

```
In [70]: from sklearn.metrics import precision_score, recall_score, f1_score

# assuming your predicted and actual labels are stored in variables y_pred and y_
accuracy = accuracy_score(y_test, prediction5)
precision = precision_score(y_test, prediction5)
recall = recall_score(y_test, prediction5)
f1 = f1_score(y_test, prediction5)

print("Accuracy:", accuracy)
print("Precision:", precision)
print("Recall:", recall)
print("F1 score:", f1)

Accuracy: 0.8640776699029126
Precision: 0.9101123595505618
Recall: 0.9310344827586207
F1 score: 0.9204545454545454
```

```
In [71]: accuracy_score(y_test, prediction5)
```

```
Out[71]: 0.8640776699029126
```

```
In [72]: probs = Model1.predict_proba(x_test)
precision_score(y_test, prediction5, average = None)
```

```
Out[72]: array([0.91011236, 0.57142857])
```

```
In [73]: recall_score(y_test, prediction5, average = None)
```

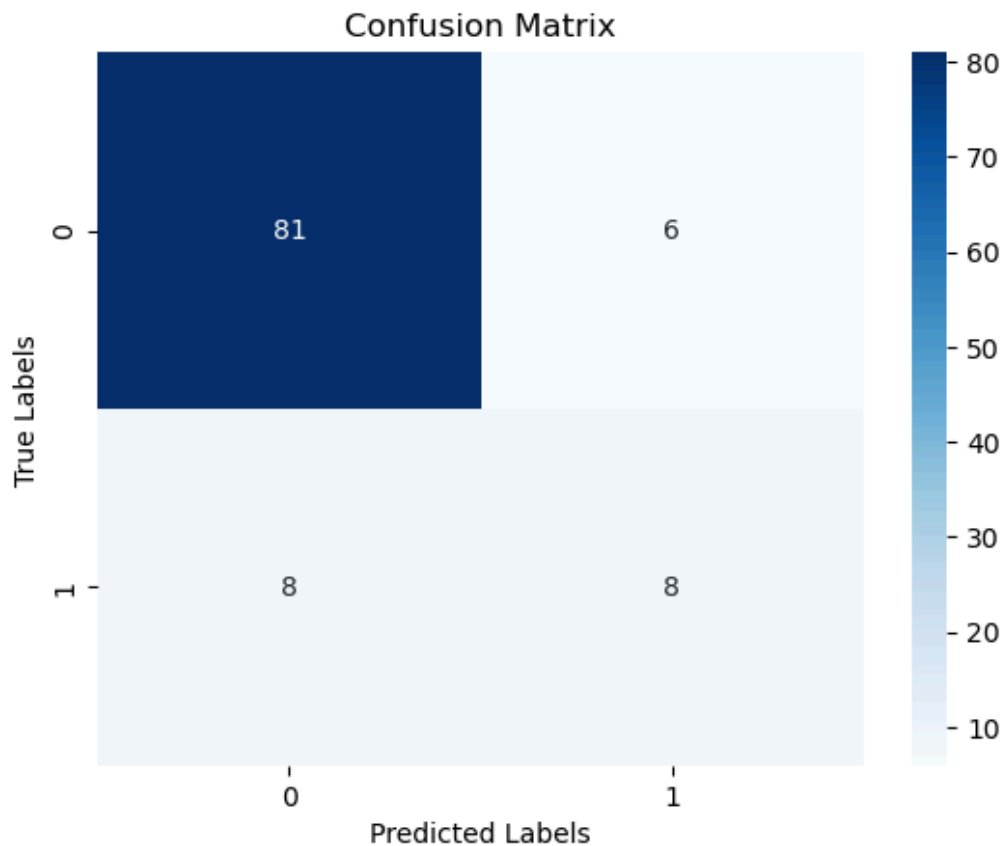
```
Out[73]: array([0.93103448, 0.5       ])
```

```
In [74]: f1_score(y_test, prediction5, average = None)
```

```
Out[74]: array([0.92045455, 0.53333333])
```



```
In [75]: cm = confusion_matrix(y_true = y_test, y_pred = prediction5)
#plot_confusion_matrix(cm, level, title = "confusion_matrix")
sns.heatmap(cm, annot=True, cmap="Blues", fmt="d")
plt.xlabel("Predicted Labels")
plt.ylabel("True Labels")
plt.title("Confusion Matrix")
plt.show()
```



Random Forest

```
In [76]: from sklearn.ensemble import RandomForestClassifier

# Initialize the classifier
rf_classifier = RandomForestClassifier(n_estimators=100, random_state=42)

# Train the model using training dataset
rf_classifier.fit(x_train, y_train)

# Make predictions on test dataset
prediction6 = rf_classifier.predict(x_test)

# Evaluate the accuracy of the model
#accuracy = rf_classifier.score(x_test, y_test)
#print("Accuracy:", accuracy)
```

```
In [77]: confusion_matrix(y_test, prediction6)
```

```
Out[77]: array([[85,  2],
               [ 9,  7]], dtype=int64)
```

```
In [78]: from sklearn.metrics import precision_score, recall_score, f1_score

# assuming your predicted and actual labels are stored in variables y_pred and y_
accuracy = accuracy_score(y_test, prediction6)
precision = precision_score(y_test, prediction6)
recall = recall_score(y_test, prediction6)
f1 = f1_score(y_test, prediction6)

print("Accuracy:", accuracy)
print("Precision:", precision)
print("Recall:", recall)
print("F1 score:", f1)

Accuracy: 0.8932038834951457
Precision: 0.9042553191489362
Recall: 0.9770114942528736
F1 score: 0.9392265193370166
```

```
In [79]: accuracy_score(y_test, prediction6)
```

```
Out[79]: 0.8932038834951457
```

```
In [80]: probs = Model1.predict_proba(x_test)
precision_score(y_test, prediction6, average = None)
```

```
Out[80]: array([0.90425532, 0.77777778])
```

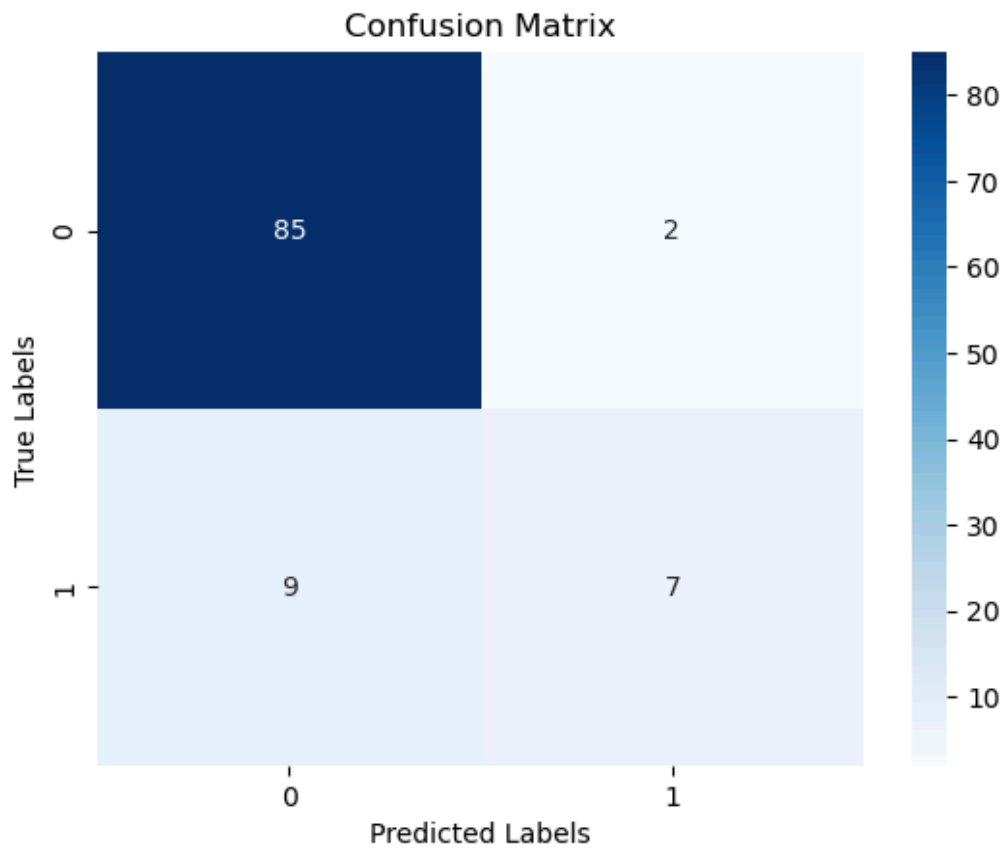
```
In [81]: recall_score(y_test, prediction6, average = None)
```

```
Out[81]: array([0.97701149, 0.4375    ])
```

```
In [82]: f1_score(y_test, prediction6, average = None)
```

```
Out[82]: array([0.93922652, 0.56      ])
```

```
In [83]: cm = confusion_matrix(y_true = y_test, y_pred = prediction6)
#plot_confusion_matrix(cm, level, title = "confusion_matrix")
sns.heatmap(cm, annot=True, cmap="Blues", fmt="d")
plt.xlabel("Predicted Labels")
plt.ylabel("True Labels")
plt.title("Confusion Matrix")
plt.show()
```

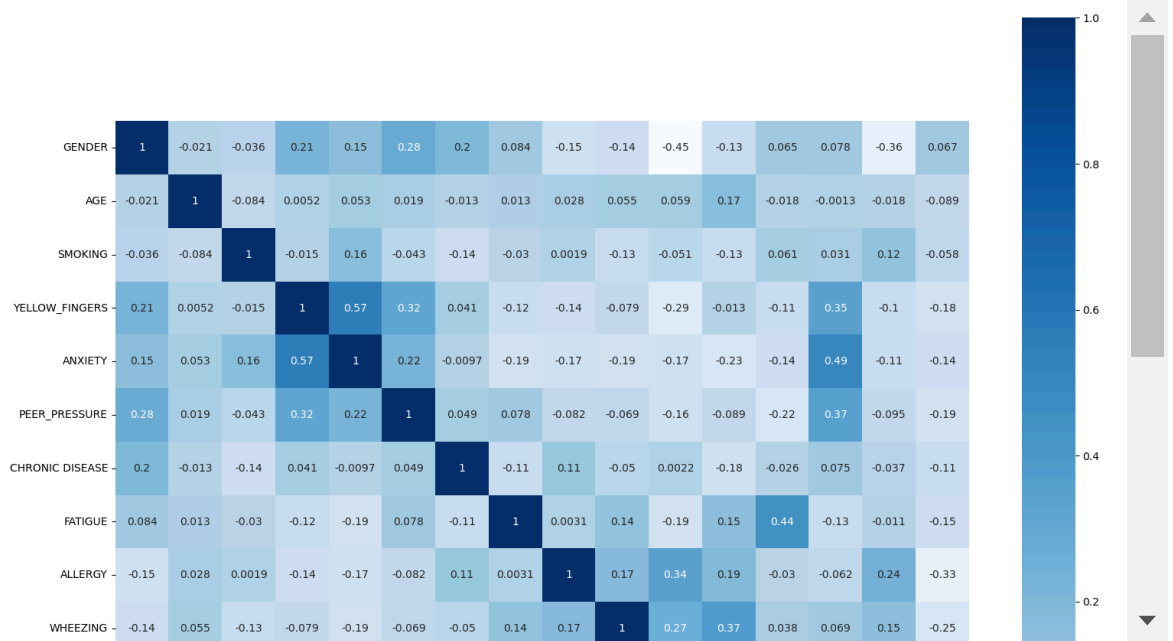


```
In [84]: #Finding Correlation
cn=lung_data.corr()
cn
```

Out[84]:

	GENDER	AGE	SMOKING	YELLOW_FINGERS	ANXIETY	PEER_PRESSURE
GENDER	1.000000	-0.021306	-0.036277	0.212959	0.152127	0.275564
AGE	-0.021306	1.000000	-0.084475	0.005205	0.053170	0.018685
SMOKING	-0.036277	-0.084475	1.000000	-0.014585	0.160267	-0.042822
YELLOW_FINGERS	0.212959	0.005205	-0.014585	1.000000	0.565829	0.323083
ANXIETY	0.152127	0.053170	0.160267	0.565829	1.000000	0.216841
PEER_PRESSURE	0.275564	0.018685	-0.042822	0.323083	0.216841	1.000000
CHRONIC DISEASE	0.204606	-0.012642	-0.141522	0.041122	-0.009678	0.048515
FATIGUE	0.083560	0.012614	-0.029575	-0.118058	-0.188538	0.078148
ALLERGY	-0.154251	0.027990	0.001913	-0.144300	-0.165750	-0.081800
WHEEZING	-0.141207	0.055011	-0.129426	-0.078515	-0.191807	-0.068771
ALCOHOL CONSUMING	-0.454268	0.058985	-0.050623	-0.289025	-0.165750	-0.159973
COUGHING	-0.133303	0.169950	-0.129471	-0.012640	-0.225644	-0.089019
SHORTNESS OF BREATH	0.064911	-0.017513	0.061264	-0.105944	-0.144077	-0.220175
SWALLOWING DIFFICULTY	0.078161	-0.001270	0.030718	0.345904	0.489403	0.366590
CHEST PAIN	-0.362958	-0.018104	0.120117	-0.104829	-0.113634	-0.094828
LUNG_CANCER	0.067254	-0.089465	-0.058179	-0.181339	-0.144947	-0.186388

```
In [85]: #Correlation
cmap=sns.diverging_palette(260,-10,s=50, l=75, n=6,
as_cmap=True)
plt.subplots(figsize=(18,18))
sns.heatmap(cn,cmap="Blues",annot=True, square=True)
plt.show()
```

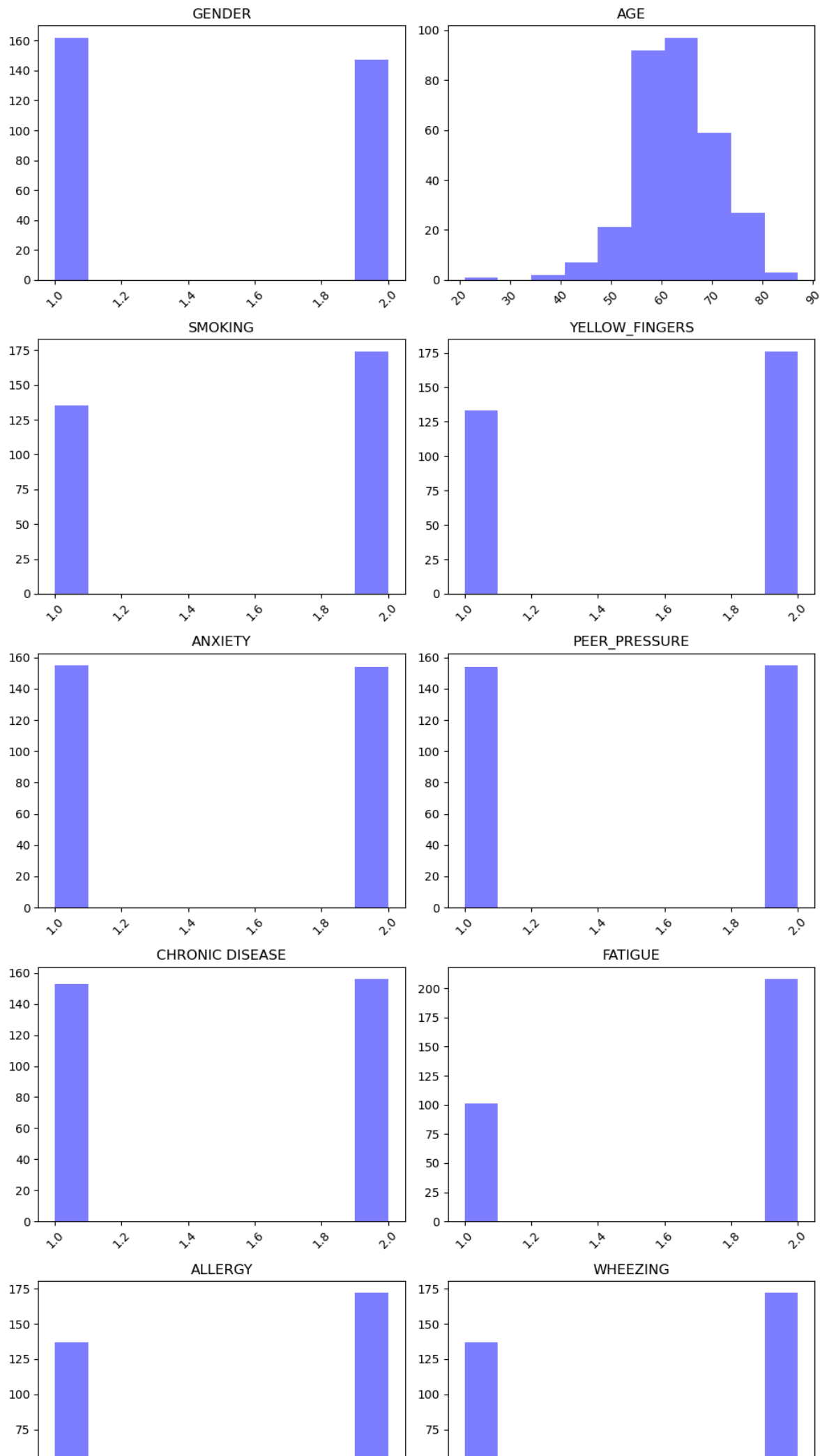


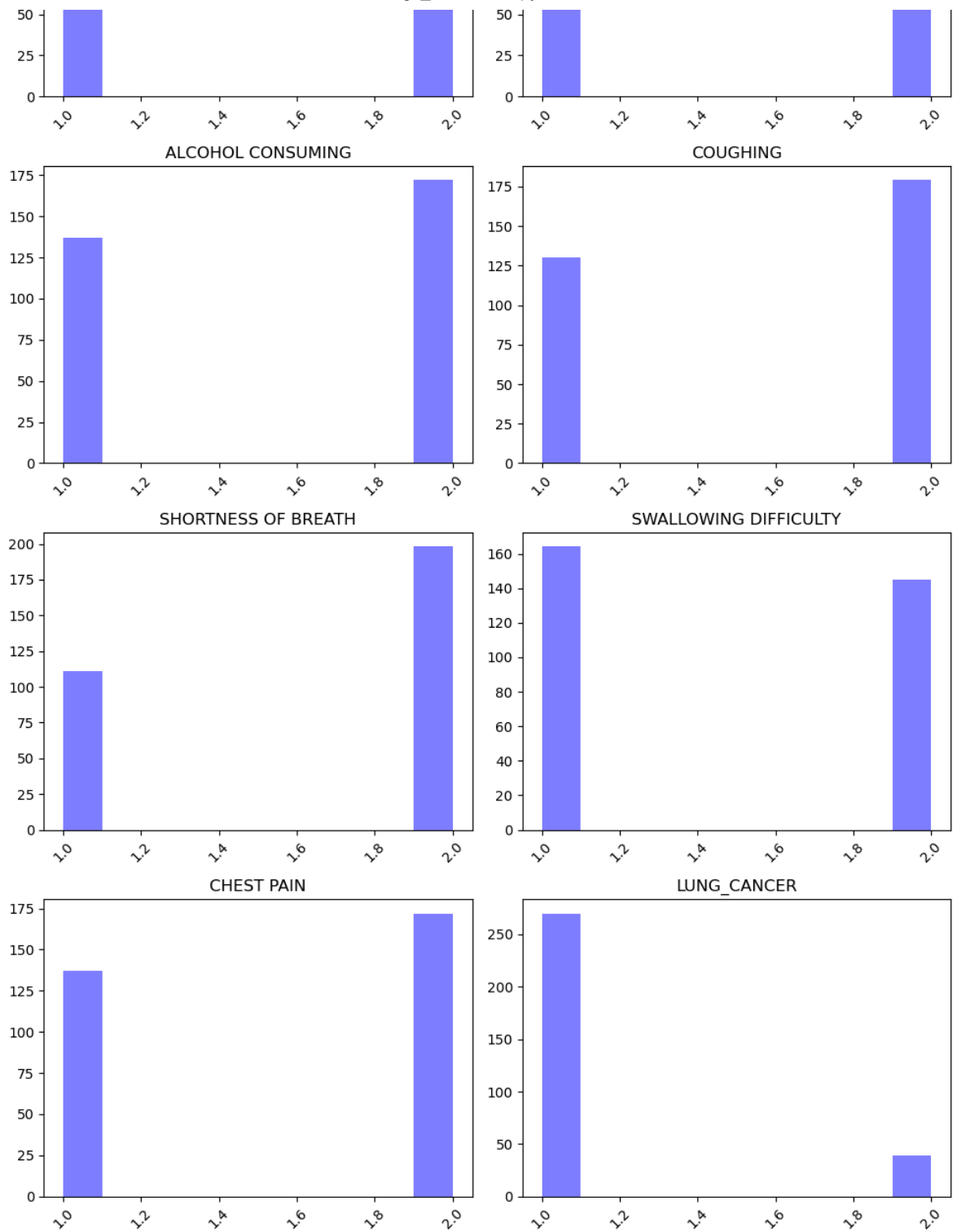
```
In [86]: num_list = list(lung_data.columns)

fig = plt.figure(figsize=(10,30))

for i in range(len(num_list)):
    plt.subplot(8,2,i+1)
    plt.title(num_list[i])
    plt.xticks(rotation=45)
    plt.hist(lung_data[num_list[i]],color='blue',alpha=0.5)

plt.tight_layout()
```



In []: