

main

June 11, 2025

```
[6]: import os
import pandas as pd

df = pd.read_csv("/home/ajay/Documents/sleeping_dog_don/Lung_cancer_detection/
↳dataset_med.csv")
df.head()
print(df.columns.tolist())
```

```
['id', 'age', 'gender', 'country', 'diagnosis_date', 'cancer_stage',
'family_history', 'smoking_status', 'bmi', 'cholesterol_level', 'hypertension',
'asthma', 'cirrhosis', 'other_cancer', 'treatment_type', 'end_treatment_date',
'survived']
```

```
[7]: if 'id' in df.columns:
    df.drop('id', axis=1, inplace=True)

if "diagnosis_date" in df.columns and 'end_treatment_date' in df.columns :
    df["diagnosis_date"] = pd.to_datetime(df['diagnosis_date'])
    df['end_treatment_date'] = pd.to_datetime(df['end_treatment_date'])

    #Creating New Column since we need exact no of days survived

    df["treatment_duration"] = (df['end_treatment_date']-df["diagnosis_date"]).
↳dt.days

    df.drop(['diagnosis_date', 'end_treatment_date'],axis=1,inplace=True)

    df.fillna(df.mode().iloc[0],inplace=True)

df.head()
```

```
[7]:   age  gender  country cancer_stage family_history smoking_status \
0  64.0   Male   Sweden      Stage I           Yes  Passive Smoker
1  50.0  Female Netherlands      Stage III          Yes  Passive Smoker
2  65.0  Female   Hungary      Stage III          Yes   Former Smoker
3  51.0  Female   Belgium      Stage I           No  Passive Smoker
4  37.0   Male  Luxembourg      Stage I           No  Passive Smoker
```

	bmi	cholesterol_level	hypertension	asthma	cirrhosis	other_cancer	\
0	29.4	199	0	0	1	0	
1	41.2	280	1	1	0	0	
2	44.0	268	1	1	0	0	
3	43.0	241	1	1	0	0	
4	19.7	178	0	0	0	0	

	treatment_type	survived	treatment_duration
0	Chemotherapy	0	523
1	Surgery	1	424
2	Combined	0	370
3	Chemotherapy	0	443
4	Combined	0	406

```
[8]: from sklearn.preprocessing import LabelEncoder , StandardScaler

categorical_columns = [
    'gender', 'country', 'cancer_stage', 'family_history', 'smoking_status', 'hypertension',
    'asthma', 'cirrhosis', 'other_cancer', 'treatment_type', 'survived']

lencoder = LabelEncoder()

for columns in categorical_columns :
    df[columns] = lencoder.fit_transform(df[columns])

df.head()
```

	age	gender	country	cancer_stage	family_history	smoking_status	bmi	\
0	64.0	1	26	0	1	3	29.4	
1	50.0	0	19	2	1	3	41.2	
2	65.0	0	12	2	1	1	44.0	
3	51.0	0	1	0	0	3	43.0	
4	37.0	1	17	0	0	3	19.7	

	cholesterol_level	hypertension	asthma	cirrhosis	other_cancer	\
0	199	0	0	1	0	
1	280	1	1	0	0	
2	268	1	1	0	0	
3	241	1	1	0	0	
4	178	0	0	0	0	

	treatment_type	survived	treatment_duration
0	0	0	523
1	3	1	424
2	1	0	370
3	0	0	443

```
[9]: sscaler = StandardScaler()

numerical_columns = ['age', 'bmi', 'cholesterol_level', 'treatment_duration']

df[numerical_columns] = sscaler.fit_transform(df[numerical_columns])

df.head()
```

```
[9]:
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	age	gender	country	cancer_stage	family_history	smoking_status	\
0	0.899796	1	26	0	1	3	
1	-0.500977	0	19	2	1	3	
2	0.999851	0	12	2	1	1	
3	-0.400922	0	1	0	0	3	
4	-1.801696	1	17	0	0	3	

	bmi	cholesterol_level	hypertension	asthma	cirrhosis	other_cancer	\
0	-0.130748	-0.797424	0	0	1	0	
1	1.279296	1.067550	1	1	0	0	
2	1.613882	0.791257	1	1	0	0	
3	1.494387	0.169599	1	1	0	0	
4	-1.289852	-1.280936	0	0	0	0	

	treatment_type	survived	treatment_duration
0	0	0	0.465906
1	3	1	-0.244658
2	1	0	-0.632238
3	0	0	-0.108287
4	1	0	-0.373851

```
[12]: from sklearn.model_selection import train_test_split
X = df.drop('survived',axis=1)
y = df['survived']
"""
X_train ,X_test , y_train , y_test = train_test_split(X , y , test_size= 0.2 ,_
↳random_state= 42 )
"""

from imblearn.over_sampling import SMOTE

smote = SMOTE(random_state=42)

X_resampled, y_resampled = smote.fit_resample(X, y)
```

```
X_train, X_test, y_train, y_test = train_test_split(X_resampled,
↳y_resampled, test_size=0.2 , random_state=42)
```

```
[13]: from sklearn.ensemble import RandomForestClassifier

model = RandomForestClassifier(class_weight=None, random_state=42)
model.fit(X_train, y_train)
```

```
[13]: RandomForestClassifier(random_state=42)
```

```
[14]: from sklearn.metrics import confusion_matrix, classification_report,
↳accuracy_score, roc_auc_score

y_pred = model.predict(X_test)
y_prob = model.predict_proba(X_test)[: , 1]

print(f"Accuracy : {accuracy_score(y_test, y_pred)}")
print(f"\n Classification Repoort \n : {classification_report(y_test, y_pred)}")
print(f"ROC-AUC SCORE : {roc_auc_score(y_test, y_prob)}")

import seaborn as sns

sns.heatmap(confusion_matrix(y_test , y_pred), annot = True , fmt = 'd', cmap =
↳'Blues')

import matplotlib.pyplot as plt

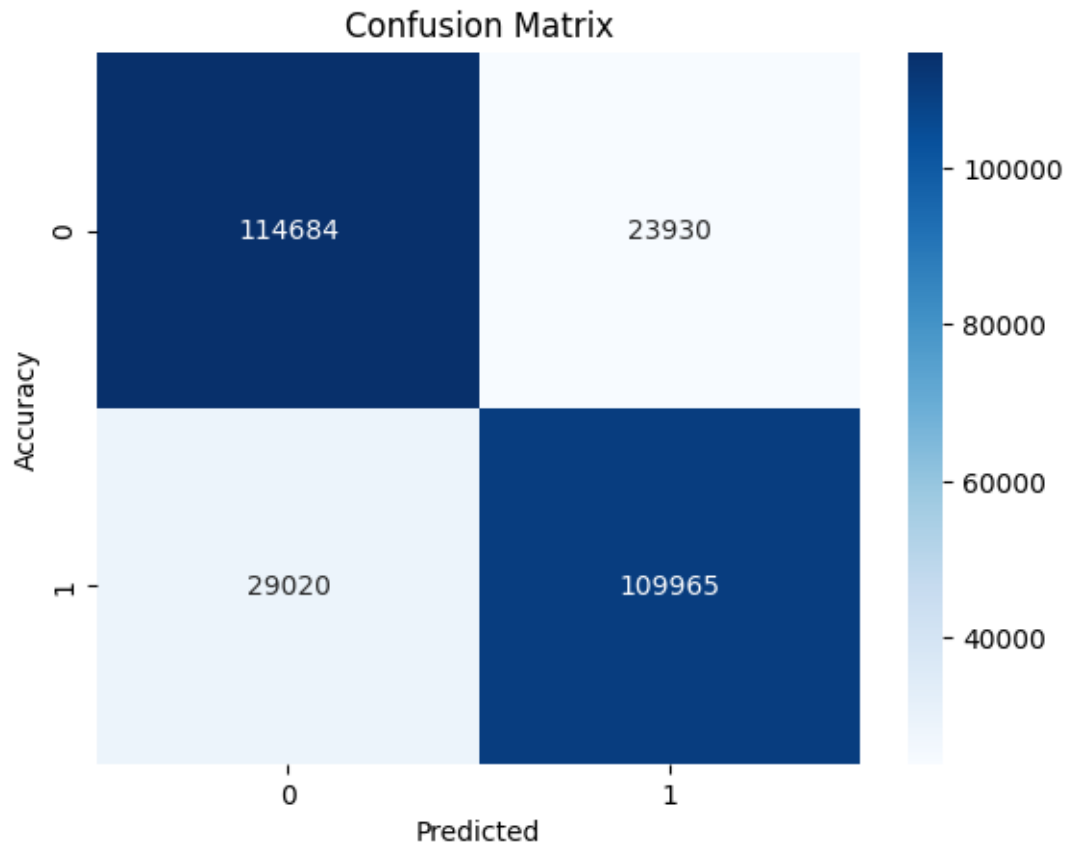
plt.title("Confusion Matrix")
plt.xlabel("Predicted")
plt.ylabel("Accuracy")
plt.show()

print(df['survived'].value_counts())
```

Accuracy : 0.80925723795835

Classification Repoort					
:		precision	recall	f1-score	support
	0	0.80	0.83	0.81	138614
	1	0.82	0.79	0.81	138985
	accuracy			0.81	277599
	macro avg	0.81	0.81	0.81	277599
	weighted avg	0.81	0.81	0.81	277599

ROC-AUC SCORE : 0.8811956814847721



```
survived
0    693996
1    196004
Name: count, dtype: int64
```

```
[15]: importances = pd.Series(model.feature_importances_, index=X.columns)
importances.sort_values(ascending=False).plot(kind='bar', figsize=(10,6))
plt.title("Feature Importances")
plt.show()
```

