

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import tensorflow
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Layer,Dense,Dropout
```

```
[ ] data=pd.read_csv("/content/thyroidDF.csv")
data.head()
```

	age	sex	on_thyroxine	query_on_thyroxine	on_antithyroid_meds	sick	pregnant	thyroid_surgery	I131_treatment	query
0	29	F	f	f	f	f	f	f	f	f
1	29	F	f	f	f	f	f	f	f	f
2	41	F	f	f	f	f	f	f	f	f
3	36	F	f	f	f	f	f	f	f	f
4	32	F	f	f	f	f	f	f	f	f

5 rows x 31 columns

```

    'K': 'general health',
    'L': 'replacement therapy',
    'M': 'replacement therapy',
    'N': 'replacement therapy',
    'O': 'antithyroid treatment',
    'P': 'antithyroid treatment',
    'Q': 'antithyroid treatment',
    'R': 'miscellaneous',
    'S': 'miscellaneous',
    'T': 'miscellaneous'}
data['target']=data['target'].map(diagnoses)

```

```
[ ] data.dropna(subset=['target'],inplace=True)
```

```
[ ] data['target'].value_counts()
```

```

hypothyroid conditions    593
general health            436
binding protein           376
replacement therapy       336
miscellaneous             281
hyperthyroid conditions   147
hyperthyroid conditions    35
antithyroid treatment     19
antithyroid treatment     14

```

Code + Text

```
[ ] x.info()

<class 'pandas.core.frame.DataFrame'>
Int64Index: 2237 entries, 4 to 9169
Data columns (total 30 columns):
#   Column                Non-Null Count  Dtype
---  -
0   age                   2237 non-null   float64
1   sex                   2237 non-null   object
2   on_thyroxine          2237 non-null   object
3   query_on_thyroxine    2237 non-null   object
4   on_antithyroid_meds   2237 non-null   object
5   sick                  2237 non-null   object
6   pregnant              2237 non-null   object
7   thyroid_surgery       2237 non-null   object
8   I131_treatment        2237 non-null   object
9   query_hypothyroid     2237 non-null   object
10  query_hyperthyroid    2237 non-null   object
11  lithium               2237 non-null   object
12  goitre                2237 non-null   object
13  tumor                 2237 non-null   object
14  hypopituitary         2237 non-null   object
15  psych                 2237 non-null   object
16  TSH_measured          2237 non-null   object
17  TSH                   2087 non-null   float64
18  T3_measured           2237 non-null   object
19  T3                    1643 non-null   float64
20  TT4_measured          2237 non-null   object
```



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2237 rows x 1 columns

[] x

4	32	F	f	f	f	f	f	f	f
---	----	---	---	---	---	---	---	---	---

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```
[ ] from sklearn.preprocessing import StandardScaler
    sc = StandardScaler()
    x_bal = sc.fit_transform(x_bal)
    x_test_bal = sc.transform(x_test_bal)

[ ] x_bal

[ ] x_test_bal=pd.DataFrame(x_test_bal,columns=columns)

[ ] x_bal=pd.DataFrame(x_bal,columns=columns)

[ ] x_bal

[ ] from sklearn.inspection import permutation_importance
    results=permutation_importance(rfr,x_bal,y_bal,scoring='accuracy')

[ ] feature_importance=['age','sex','on_thyroxine','query_on_thyroxine','on_antithyroid_meds','sick','pregnant','thyroid_surg',
    importance=results.importances_mean
    importance=np.sort(importance)
    for i,v in enumerate(importance):
```

```
[ ] y_pred=xgb1.predict(x_test_os)

[ ] print(classification_report(y_test_os,y_pred))

[ ] accuracy_score(y_test_os,y_pred)

[ ] from sklearn.svm import svc
    from sklearn.metrics import accuracy_score,classification_report
    sv=SVC()

[ ] sv.fit(x_bal,y_bal)

[ ] y_pred=sv.predict(x_test_bal)

[ ] print(classification_report(y_test_bal,y_pred))

[ ] train_score=accuracy_score(y_bal,sv.predict(x_bal))
    train_score
```

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```
[ ] col=['goitre','tumor','hypopituitary','psych','TSH','T3','TT4','T4U','FTI','TBG']
    da=[[0,0,0,0,0.000000,0.0,0.0,1.00,0.0,40.0]]
    dal=pd.DataFrame(data=da,columns=col)
    xgb1.predict(dal)

[ ] model.predict([[0,0,0,0,0.000000,0.0,0.0,1.00,0.0,40.0]])

[ ] print(classification_report(y_test_bal,y_pred))

[ ] train_score=accuracy_score(y_bal,rfr1.predict(x_bal))



[ ] train_score

[ ] y_pred=xgb.predict(x_test_bal)

[ ] print(classification_report(y_test_bal,y_pred))

[ ] train_score=accuracy_score(y_bal,xgb.predict(x_bal))
    train_score
```

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```
[ ] print(classification_report(y_test_bal,y_pred))

[ ] train_score=accuracy_score(y_bal,sv.predict(x_bal))
  train_score

[ ] y_pred=model.predict(x_test_bal)

[ ] print(classification_report(y_test_bal,y_pred))

[ ] accuracy_score(y_test_bal,y_pred)

[ ] params={
    'C':[0.1,1,10,100,1000],
    'gamma':[1,0.1,0.01,0.001,0.0001],
    'kernel':['rbf','sqrt']
  }

[ ] random_svc=RandomizedSearchCV(sv,params,scoring='accuracy',cv=5,n_jobs=-1)

[ ] random.svc.fit(x_bal,y_bal)
```



```
[ ] random_svc.best_params_  
  
[ ] sv1=SVC(kernel='rbf',gamma=0.1,c=100)  
  
[ ] sv1.fit(x_bal,y_bal)  
  
[ ] y_pred=sv1.predict(x_test_bal)  
  
[ ] print(classification_report(y_test_bal,y_pred))  
  
[ ] train_score=accuracy_score(y_bal,sv1.predict(x_bal))  
    trsin score  
  
[ ] import pickle  
    pickle.dump(sv1,open('thyroid_1_model.pk1','wb'))  
  
[ ] features=np.array([[0,0,0,0,0.000000,0.0,0.0,1.00,0.0,40.0]])  
    print(label_encoder.inverse_transform(xgb1.predict(features)))
```

```
[ ] @app.route("/pred", methods=['POST', 'GET'])
def predict():
    x=[[float(x) for x in request.from.value()]]
    print(x)
    col=['goitre', 'tumor', 'hypopituitary', 'psych', 'TSH', 'T3', 'TT4', 'T4U', 'FTI', 'TBG']
    x=pd.DataFrame(x, columns=col)
    print(x)
    pred=model.predict(x)
    pred=le.inverse_transform(pred)
    print(pred[0])
    return render_template('submit.html', prediction_text=str(pred))

[ ] if __name__ == "__main__":
    app.run(debug=False)

[ ] In [32]: runfile('C:/Users/SmartBridge-PC/Downloads/Thyroid/app.py', wdir='C:/users/SmartBridge-PC/Downloads/Thyroid')
```

```
[ ] pickle.dump(label_encoder,open('label_encoder.pkl','wb'))

[ ] data['target'].unique()

[ ] y['target'].unique()

[ ] import pickle
    pickle.dump(sv1,open('thyroid_1_model.pkl','wb'))

[ ] from flask import Flask,render_template,request
    import numpy as np
    import pickle
    import pandas as pd

[ ] model=pickle.load(open(r"C:\Users\SmartBridge-PC\Downloads\Thyroid\thyroid1_model.pkl",'rb'))
    le=pickle.load(open("label_encoder.pkl",'rb'))
    app=Flask(__name__)

[ ] @app.route("/")
    def about():
        return render_template('home.html')
```

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```
[ ] model.add(Dense(units=128,activation='relu',input_shape=(10,)))

[ ] model.add(Dense(units=128,activation='relu',kernel_initializer='random_uniform'))
model.add(Dropout(0.2))
model.add(Dense(units=256,activation='relu',kernel_initializer='random_uniform'))
model.add(Dropout(0.2))
model.add(Dense(units=128,activation='relu',kernel_initializer='random_uniform'))

[ ] model.add(Dense(units=1,activation='sigmoid'))

[ ] model.summary()

[ ] model.compile(loss='binary_crossentropy',optimizer='adam',metrics=['accuracy'])

[ ] model.fit(x_bal,y_bal,validation_data=[x_test_bal,y_test_bal],epochs=15)

[ ] rfr1.predict([[0,0,0,0,0.000000,0.0,0.0,1.00,0.0,40.0]])

[ ] sv.predict([[0,0,0,0,0.000000,0.0,0.0,1.00,0.0,40.0]])
```

+ Code + Text

```
[ ] i=feature_importance[i]
    print('feature:{:<20}score:{}'.format(i,v))
    plt.figure(figsize=(10,10))
    plt.bar(x=feature_importance,height=importance)
    plt.xticks(rotation=30,ha='right')
    plt.show()

[ ] x.head()

[ ] x_bal.drop(['age','sex','on_thyroxine','query_on_thyroxine','on_antithyroid_meds','sick','pregnant','thyroid_surgery'])

[ ] x_test_bal.drop(['age','sex','on_thyroxine','query_on_thyroxine','on_antithyroid_meds','sick','pregnant','thyroid_surgery'])

[ ] x_bal.head()

[ ] data.info()

[ ] import seaborn as sns
    corrmatrix=x.corr()
    f,ax=plt.subplots(figsize=(9,8))
    sns.heatmap(corrmatrix,ax=ax,cmap="YlGnBu",linewidths=0.1)
```

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```
[ ] from sklearn.ensemble import RandomForestClassifier
    rfr1=RandomForestClassifier().fit(x_os,y_os.values.ravel())
    y_pred=rfr1.predict(x_test_os)
    rfr1=RandomForestClassifier()
```

```
[ ] rfr1.fit(x_os,y_os.values.ravel())
```

```
[ ] y_pred=rfr1.predict(x_test_os)
```

```
[ ] y_pred=rfr1.predict(x_test_os)
```

```
[ ] from sklearn.metrics import classification_report
    print(classification_report(y_test_os,y_pred))
```

```
[ ] train_score=accuracy_score(y_os,rfr1.predict(x_os))
    train_score
```

```
[ ] from xgboost import XGBClassifier
    xgb1 = XGBClassifier()
    xgb1.fit(x_os,y_os)
```

```
[ ] from sklearn.model_selection import train_test_split
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.20,random_state=0)
```

```
▶ from imblearn.over_sampling import SMOTE
y_train.value_counts()
```

```
↳ 841108102    1
   850503041    1
   841120013    1
   850102019    1
   850530045    1
   ..
   850509071    1
   860916057    1
   850527065    1
   861118071    1
   860707017    1
Name: patient_id, Length: 1789, dtype: int64
```

```
[ ] os = SMOTE(random_state=0,k_neighbors=1)
x_bal,y_bal=os.fit_resample(x_train,y_train)
x_test_bal,y_test_bal=os.fit_resample(x_test,y_test)
```

```
[ ]
```

```
9162  36  F      f      f      f      f      f      f      f      f
9169  69  M      f      f      f      f      f      f      f      f
```

2237 rows × 30 columns

```
[ ] label_encoder=LabelEncoder()
y_dt=label_encoder.fit_transform(y)
```

```
▶ y=pd.DataFrame(y_dt,columns=['target'])
```

```
[ ] y
```

	target
0	0
1	1
2	2

Code + Text

RAM

Disk

	32	41	M	f	f	f	f	f		
	33	71	F	t	f	f	f	f	f	f
	39	55	F	t	f	f	f	f	f	f

	9153	64	M	f	f	f	f	f	f	f
	9157	60	M	f	f	t	f	f	f	f
	9158	64	M	f	f	f	f	f	f	f
	9162	36	F	f	f	f	f	f	f	f
	9169	69	M	f	f	f	f	f	f	f

2237 rows x 30 columns



x



	age	sex	on_thyroxine	query_on_thyroxine	on_antithyroid_meds	sick	pregnant	thyroid_surgery	I131_treatment
4	32	F	f		f	f	f	f	f
18	63	F	t		f	t	f	f	f
32	41	M	f		f	f	f	f	f
33	71	F	t		f	f	f	f	f
39	55	F	t		f	f	f	f	f
...
9153	64	M	f		f	f	f	f	f
9157	60	M	f		f	t	f	f	f
9158	64	M	f		f	f	f	f	f
9162	36	F	f		f	f	f	f	f

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```
data['target'].value_counts()
```

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```
hypothyroid conditions    593
general health            436
binding protein           376
replacement therapy       336
miscellaneous             281
hyperthyroid conditions   147
hyperthyroid conditions    35
antithyroid treatment     19
antithyroid treatment     14
Name: target, dtype: int64
```

```
[ ] data[data.age>100]
```

```
   age  sex  on_thyroxine  query_on_thyroxine  on_antithyroid_meds  sick  pregnant  thyroid_surgery  I131_treatment  query_
0 rows x 31 columns
```

◀

```
[ ] x=data.iloc[:,0:-1]
    y=data.iloc[:, -1]
```

+ Code + Text

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id	age	sex	height	weight	fat	muscle	bone	fat2	muscle2	bone2	fat3	muscle3	bone3
9153	64	M	f	f	f	f	f	f	f	f	f	f	f
9157	60	M	f	f	t	f	f	f	f	f	f	f	f
9158	64	M	f	f	f	f	f	f	f	f	f	f	f
9162	36	F	f	f	f	f	f	f	f	f	f	f	f
9169	69	M	f	f	f	f	f	f	f	f	f	f	f

2237 rows x 30 columns

2237 rows x 30 columns

```
[ ] x['sex'].unique()
array(['F', 'M', nan], dtype=object)

[ ] x['sex'].replace(np.nan, 'F', inplace=True)

[ ] x['sex'].value_counts()
```



x

	age	sex	on_thyroxine	query_on_thyroxine	on_antithyroid_meds	sick	pregnant	thyroid_surgery	I131_treatment	qu
4	32	F	f	f	f	f	f	f	f	f
18	63	F	t	f	f	t	f	f	f	f
32	41	M	f	f	f	f	f	f	f	f
33	71	F	t	f	f	f	f	f	f	f
39	55	F	t	f	f	f	f	f	f	f
...
9153	64	M	f	f	f	f	f	f	f	f
9157	60	M	f	f	t	f	f	f	f	f
9158	64	M	f	f	f	f	f	f	f	f

+ Code + Text

```
[ ] T3_measured      0
    T3              2604
    TT4_measured    0
    TT4             442
    T4U_measured    0
    T4U             809
    FTI_measured    0
    FTI             802
    TBG_measured    0
    TBG            8823
    referral_source  0
    target          0
    patient_id      0
    dtype: int64
```

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```
▶ diagnoses={ 'A': 'hyperthyroid conditions',
               'B': 'hyperthyroid conditions',
               'C': 'hyperthyroid conditions',
               'D': 'hyperthyroid conditions',
               'E': 'hypothyroid conditions',
               'F': 'hypothyroid conditions',
               'G': 'hypothyroid conditions',
               'H': 'hypothyroid conditions',
               'I': 'binding protein',
               'J': 'binding protein',
               'K': 'general health',
```

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+ Code + Text

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▶ data.shape

(9172, 31)

▶ data.isnull().sum()

▶

age	0
sex	307
on_thyroxine	0
query_on_thyroxine	0
on_antithyroid_meds	0
sick	0
pregnant	0
thyroid_surgery	0
I131_treatment	0
query_hypothyroid	0
query_hyperthyroid	0
lithium	0
goitre	0
tumor	0
hypopituitary	0
psych	0
TSH_measured	0
TSH	842
T3_measured	0
T3	2604