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#### INTRODUCTION

#### **OVERVIEW**

Thyroid diseases, such as hypothyroidism and hyperthyroidism, are common endocrine disorders that affect the function of the thyroid gland. These diseases can have a significant impact on a patient's health and quality of life. Early and accurate diagnosis of thyroid diseases is important for effective treatment.

In recent years, machine learning techniques have been applied to the classification of thyroid diseases. The goal of these studies is to develop models that can accurately diagnose thyroid diseases based on clinical and laboratory data.

## **PURPOSE**

There are several machine learning algorithms that have been used for thyroid disease classification, including decision trees, random forests, k-nearest neighbors (KNN), support vector machines (SVM), artificial neural networks (ANN), and deep learning algorithms such as convolutional neural networks (CNNs) and recurrent neural networks (RNNs).

The input data for these models can include clinical features, such as age, gender, and symptoms, as well as laboratory test results, such as thyroid-stimulating hormone (TSH) levels and levels of thyroxine (T4) and triiodothyronine (T3).

The performance of these models is usually evaluated using metrics such as accuracy, precision, recall, and F1 score. In general, deep learning algorithms have shown better performance than other machine learning algorithms in thyroid disease classification tasks.

### LITERATURE SURVEY

## **EXISTING PROBLEM**

The current existing system includes:

Clinical examination: This involves a physical examination of the neck to check for any visible signs of thyroid enlargement or nodules.

Blood tests: Blood tests are used to measure the levels of hormones produced by the thyroid gland and to check for antibodies that may indicate autoimmune diseases such as Hashimoto's thyroiditis.

Ultrasound: An ultrasound scan can provide images of the thyroid gland and help to identify any nodules or other abnormalities.

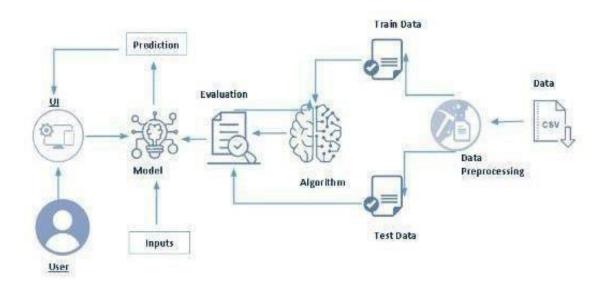
Fine needle aspiration biopsy (FNAB): This is a procedure in which a small sample of tissue is taken from a thyroid nodule using a fine needle, which is then examined under a microscope to check for cancer.

## PROPOSED SYSTEM

The proposed system is by using Artificial Intelligence (AI) and Machine Learning (ML) In recent years, there have been several studies exploring the use of AI and ML algorithms for predicting thyroid diseases. These systems are trained on large datasets of patient data and use various features such as demographic information, blood test results, and ultrasound images to make prediction.

## THEORETICAL ANALYSIS

#### **BLOCK DIAGRAM**



### HARDWARE AND SOFTWARE DESIGNING

## **Python**

Python is an interpreted, object-oriented, high-level programming language with dynamic semantics. It was created by Guido van Rossum, and first released on February 20, 1991. Its high-level built-in data structures, combined with dynamic typing and dynamic binding, make it very attractive for Rapid Application Development, as well as for use as a scripting or glue language to connect existing components together. Python's simple, easy to learn syntax emphasizes readability and therefore reduces the cost of program maintenance. Python supports modules and packages, which encourages program modularity and code reuse. The Python interpreter and the extensive standard library are available in source or binary form without charge for all major platforms, and can be freely distributed.

## **Anaconda Navigator**

Anaconda Navigator is a free and open-source distribution of the Python and R programming languages for data science and machine learning related applications. It can be installed on Windows, Linux, and macOS. Conda is an open-source, cross platform, package management system. Anaconda comes with

so very nice tools like JupyterLab, Jupyter Notebook, QtConsole, Spyder, Glueviz, Orange, Rstudio, Visual Studio Code. For this project, we will be using Jupyter notebook and Spyder.

## Jupyter Notebook

The Jupyter Notebook is an open-source web application that you can use to create and share documents that contain live code, equations, visualizations, and text. Jupyter Notebook is maintained by the people at Project Jupyter. Jupyter Notebooks are a spin-off project from the IPython project, which used to have an IPython Notebook project itself. The name, Jupyter, comes from the core supported programming languages that it supports: Julia, Python, and R. Jupyter ships with the IPython kernel, which allows you to write your programs in Python, but there are currently over 100 other kernels that you can also use

## **Spyder**

Spyder, the Scientific Python Development Environment, is a free integrated development environment (IDE) that is included with Anaconda. It includes editing, interactive testing, debugging, and introspection features. Initially created and developed by Pierre Raybaut in 2009, since 2012 Spyder has been maintained and continuously improved by a team of scientific Python developers and the community. Spyder is extensible with first-party and third party plugins includessupport for interactive tools for data inspection and embeds Python specific code. Spyder is also pre-installed in Anaconda Navigator, which is included in Anaconda.

## Flask

Web frame work used for building. It is a web application framework written in python which will be running in local browser with a user interface. In this application, whenever the user interacts with UI and selects emoji, it will suggest the best and top movies of that genre to the use.

## **Hardware Requirements:**

o Operating system: window 7 and above with 64bit o Processor Type -Intel Core i3-3220

o RAM: 4Gb and above

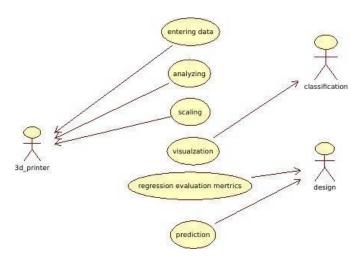
o Hard disk: min 100GB

#### 1. EXPERIMENTAL INVESTIGATION

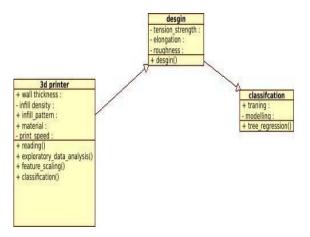
Here we are going to build a machine learning model that predicts whether the given message is a spam or not, based on these parameters a supervised machine learning model is built to predict the best

material to be used for building 3D models. A web application is build so that the user can type in the mentioned part a meters and the material which suits the best is showcased on UI.

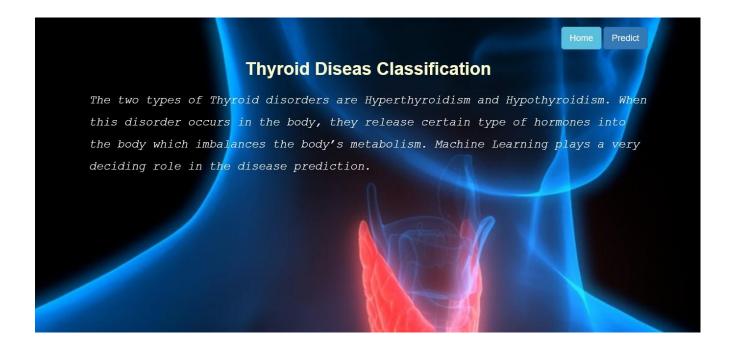
# 2. <u>FLOWCHART</u>

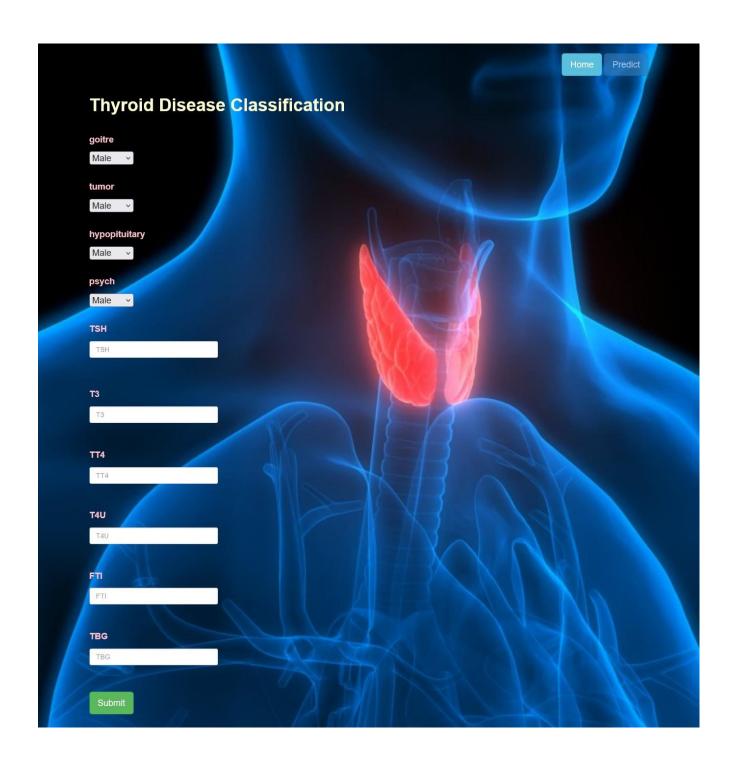


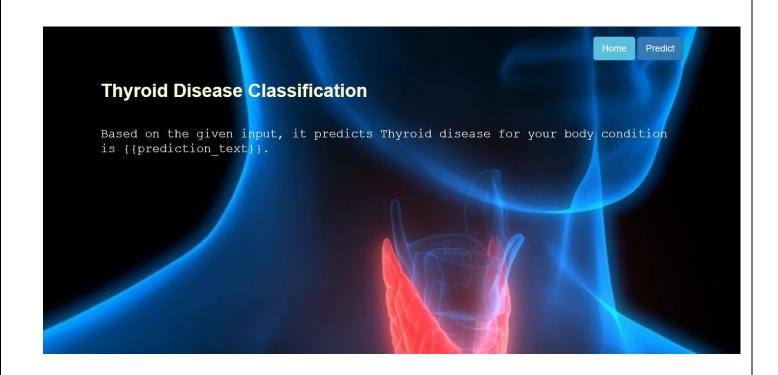
## **USE CASE DIAGRAM**



# 3. RESULT







## 4.ADVANTAGES & DISADVANTAGES

## **ADVANTAGES**

- Easy to use
- Cost efficient
- Time efficient

## **DISADVANTAGE**

- 1. Initial costs of printer
- 2. Post processing
- 3. Printing time
- 4. Special skill required for 3D models
- 5. Manufacturing Job Losses

## 8. Applications

3D printing has gone through a number of changes over the years. In the early days, 3D printing was time-consuming and costly, and not very practical for applications outside of industry. However, with the advent of today's more flexible and cost-effective 3D printing methods, there are areas where 3D printing has become a practical tool.

# It is applicable in different sectors such as

- Engineering And Design
- Consumer products
- Manufacturing
- Education
- Aerospace
- Medical
- Movies / Theatres
- Architectures

#### 9. CONCLUSION

3D printing technology could revolutionize and re-shape the world. Advance in 3D technology can significantly change and improve the way we manufacture products goods worldwide.

If the last industrial revolution brought us mass production and the advent of economics of scale – the digital 3D printing revolution could bring mass manufacturing back a full of circle – to an era of mass personalization, and return to individual craftsmanship.

#### 10. FUTURE SCOPE

Future applications for 3D printing might include creating open-source scientific equipment to create opensource labs

Science-based applications like reconstructing fossils in palaeontology . Replicating ancient and priceless artifacts in archaeology

Reconstructing bones and body parts in forensic pathology. The technology currently being researched for building construction.

#### 11. BIBILOGRAPHY

- http://mashable.com/2014/03/06/3d-printed-blood-vessels/
- http://www.3dprinter.net/

# 12. APPENDIX

```
| Second Content of Second Content S
```

```
In [3]: data.isnull().sum()
2604
           TT4
T4U_measured
T4U
                                     809
           T4U
FTI_measured
FTI
TBG_measured
TBG
referral_source
target
patient_id
dtype: int64
                                      8823
          Assembling Redundant attributes from dataset

#The columns Listed below were removed because of redundacy.

#The columns Listed below were removed because of redundacy.

#They are boolean and state whether or not a value has been recorded for their respective blood tests.

#TSL measured

data.drop(['TSL measured', 'TSL measured', 'TSL measured', 'TFL measured', 'TBG measured', 'referral_source', 'patient_id'], axis=1, inplace = True)
In [5]: data.head()
Out[5]: age sex on_thyroxine query_on_thyroxine on_antithyroid_meds sick pregnant thyroid_surgery I131_treatment query_hypothyroid ... tumor hypopituitary psych TSH T3 TT4 T4U FTI T8G target
                                                                       1 29 F f f
                                                                                                                                                                                     f NaN NaN NaN NaN 11.0
           2 41 F
                                                                                                                                                  f ... f f NaN NaN NaN NaN NaN NaN 26.0
           3 36 F
          4 32 F
                                                                                                                                                                                    f NaN NaN NaN NaN NaN 36.0
5 rows × 23 columns
```

```
In [6]: data['target']
Out[6]: 0
       3 4
      9167 -
9168 -
9169 I
9170 -
9171 -
Name: target, Length: 9172, dtype: object
'T': 'miscellaneous'}
data['target'] = data['target'].map(diagnoses) #remapping
In [8]: data
Out [8]: age sex on_thyroxine query_on_thyroxine on_antithyroid_meds sick pregnant thyroid_surgery I131_treatment query_hypothyroid ... tumor hypopituitary psych TSH T3 T74 T4U FTI T8G
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[11]:	hy ge bi re mi hy an Na da	poth nera ndin plac scel pert tith me: ta.h	yroid condition l health g protein ement therapy laneous hyroid conditio yroid treatment target, dtype: ead() e sex on_thyroxin 2 F 3 F	436 376 336 281 182 33 int64	f f	f f	f t	oregnant thyroid	f f	f f	t	f f	f	f f 68.0	NaN Nal	N NaN N 48.0	NaN 1.02	NaN 36.0 47.0 NaN	miscellaneous I hypothyroid conditions	
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mean	52.792579	14.930791	1.961875	116.390495	1.013439	120.363369	47.717347												
std	19.677450	46.204092	1.452238	60.351600	0.280222	70.996728	32.398750												
min	1.000000	0.005000	0.050000	2.000000	0.170000	1.400000	9.299999												
25%	36.000000	0.255000	1.000000	76.000000	0.850000	83.000000	32.000000												
50%	56.000000	2.000000	1.700000	109.000000	0.960000	109.000000	36.000000												
75%	69.000000	8.799999	2.500000	156.000000	1.120000	157.000000	46.750000												
max	95.000000	530.000000	18.000000	600.000000	2.330000	881.000000	200.000000												
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	× 23 columns																		
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import #chang data[" data 4 18 32 33 39	x 23 columns t numpy as r ging age of age 'j=np.wl age sex on 32.0 F 63.0 F 41.0 M 71.0 F	observation lere((data.a thyroxine qu f t f t	ge>100), np	ine on_antith  f f f f	yroid_meds s f f f f	f f t f f f f f	thyroid_surg	f f f f f f f f f f f f f f f f f f f	f f f t	f f f f	f f f	f f f	NaN 68.000000 0.050000 0.050000	NaN NaN 1.6 NaN 2.4	NaN N 48.0 1. 39.0 1. 126.0 1.	N Nah 02 47.0 00 39.0 18 91.0	1 36.0 NaN NaN NaN NaN	miscellaneous hypothyroid conditions miscellaneous binding protein replacement therapy	
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```
converting categorical to numerical values
  In [27. #applying ordinal_encoding to x values
#Encoding the categorical data
#Encoding the independent(output) variable
from sklearn.preprocessing import OrdinalEncoder, LabelEncoder
#Cotegorical data
                   ordinal_encoder = OrdinalEncoder(dtype = 'int64')
x.iloc[:, 1:16] = ordinal_encoder.fit_transform(x.iloc[:, 1:16])
#ordinal_encoder.fit_transform(x[['sex']])
 In [28_ x.head()
Out [28]: age sex on, thyroxine query_on, throwine query_on, throwing and the properties of the proper
                 39 55.0 0
                                                                                                                                                                                                                                                                                                                   0 9.599999 2.4 136.0 1.48 92.0 NaN
               5 rows × 22 columns
 In [29_ x.replace(np.nan, '0', inplace=True)
    x.head()
 39 55.0 0
                                                                                                                                                                                                                                                                                                                   0 9.599999 2.4 136.0 1.48 92.0
               5 rows × 22 columns
In |38_ #applying label_encoding to y values
label_encoder = LabelEncoder()
y_dt= label_encoder.fit_transform(y)
```

```
In [31]

vpd.Out[31]:

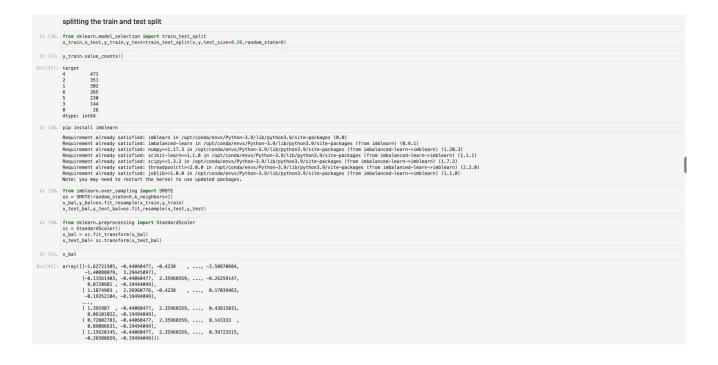
trupt

0ut[31]:

vpd.Out[31]:

vpd.
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```
In [42_ x_test_bal
Out[42]: array([[-1.5229667 , -0.44868477 , -0.4238 , ..., 1.06342846, ...]
[-0.874763, -0.44866477 , -0.4238 , ..., 1.76703086, ...]
[-0.30213342 , -0.19494091], ...]
[-0.9496008 , 2.26960776 , -0.4238 , ..., -0.39789962, ...]
                  [1.39813447, -e.4466477, 2.35968359, ..., 0.81835453, 0.7694189, -e.19494891], [1.33846247, -e.44669477, 2.35968359, ..., 0.81987378, 0.67327619, -e.19494891], -e.19443252, -e.44669477, -e.4238 , ..., 0.24838842, 0.37610348, -e.19494949])
 In (43_ y_bal.value_counts()
Out[43]: target
          dtype: int64
 In [44. columns=['age', 'sex', 'on_thyroxine', 'query_on_thyroxine', 'on_antithyroid_meds', 'sick', 'pregnant', 'thyroid_surgery', 'I33_treatment', 'query_hypothyroid', 'query_hyperthyroid', 'tithium', 'goitre', 'tumor', 'hypopituitary', 'psych', '
 In [45_ x_test_bal= pd.DataFrame(x_test_bal,columns=columns)
 In [46_ x_bal= pd.DataFrame(x_bal,columns=columns)
 In [47_ x_bal
                                 sex on_thyroxine query_on_thyroxine on_antithyroid_meds sick pregnant thyroid_surgery 1131_treatment query_hypothyroid ... goitre tumor hypopituitary psych
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           3 -1.366594 -0.440605 -0.423800
                                                           -0.105069
                                                                       -0.158703 -0.141815 -0.137297
                                                                                                                     -0.239601
              4 -0.167738 -0.440605
                                        -0.423800
                                                             -0.105069
                                                                                 -0.158703 -0.141815 -0.137297
                                                                                                                     -0.239601
                                                                                                                                    -0.162675
                                                                                                                                                      -0.230986 ... -0.052319 -0.137297
                                                                                                                                                                                            -0.024637 -0.107982 -0.306321 4.541622 1.459767 -0.127283 1.496783 -0.19494
                                                                                  -0.158703 -0.141815 -0.137297
           3293 0.383062 -0.440605
                                         2.359604
                                                            -0.105069
                                                                            -0.158703 -0.141815 -0.137297
                                                                                                                     -0.239601
                                                                                                                                    -0.162675 -0.230986 ... -0.052319 -0.137297 -0.024637 -0.107982 -0.309176 -0.856540 0.565143 -0.513902 1.085434 -0.19494
          3294 1.395987 -0.440605
                                         2.359604
                                                             -0.105069
                                                                                 -0.158703 -0.141815 -0.137297
                                                                                                                      -0.239601
                                                                                                                                     -0.162675
                                                                                                                                                       -0.230986 ... -0.052319 -0.137297
                                                                                                                                                                                             -0.024637 -0.107982 -0.095452 -0.172405 0.248906 0.436150 0.061010 -0.19494
          3295 0.728028 -0.440605 2.359604
                                                                             -0.158703 -0.141815 -0.137297
                                                                                                                  -0.239601
                                                                                                                                   -0.162675 -0.230986 ... -0.052319 -0.137297 -0.024637 -0.107982 -0.311566 0.087864 1.071643 0.143333 0.890866 -0.19494
                                                           -0.105069
                                                                                                                                                       -0.230986 ... -0.052319 -0.137297
          3296 1.156281 -0.440605
                                                             -0.105069
                                                                                  -0.158703 -0.141815 -0.137297
                                                                                                                                     -0.162675
                                                                                                                                                                                            -0.024637 -0.107982 -0.072439 0.079407 -0.200359 0.397235 -0.265887 -0.1949
          3297 rows × 22 columns
```

```
In [48. from sklearn.ensemble import RandomForestClassifier
from sklearn.ensemble import accuracy_score, classification_report
rfr = RandomForestClassifier(.fit(_bal,_bal)
y_pred = rfr.predict(x_test_bal)
accuracy_score(y_test_bal,_pred)
x_bal.shape,x_test_bal.shape,x_test_bal.shape

//tmp/wauser/ipykernet_lo28/2606972409,py:31 StatcOnversionWarning: A column-vector y was passed when a 1d array was expected. Please change the shape of y to (n_samples,), for example using ravel().
rfr = RandomForestClassifier().fit(x_bal,_bal)

Out(48): ((3297, 22), (3297, 1), (854, 22), (854, 1))

In [49. test_scoremaccuracy_score(y_test_bal,y_pred)
test_score

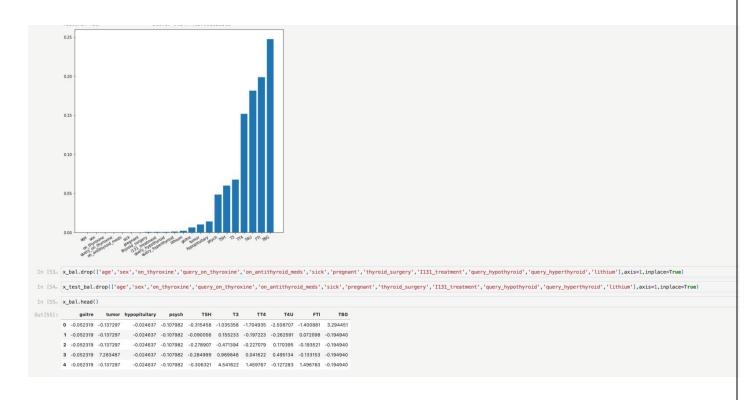
Out(49): 0.9098380655737705

In [50. train_score = accuracy_score(y_bal,rfr.predict(x_bal))
train_score

Out(50): 1.0
```

```
performing feature importance

In [SL. gaperian feature ignortance feature; ignortance feature; ignortance permutation_importance feature; ignortance feature; ignortance; ignorta
```





```
In [70_ random_svc.best_params_
 Out[70]: {'kernel': 'rbf', 'gamma': 1, 'C': 1000}
  In [78- sv1=SVC(kernel= 'rbf', gamma= 0.1,C= 100)
SVC(C=100, gamma=0.1)
  In [80_ y_pred= sv1.predict(x_test_bal)
  In [81_ print(classification_report(y_test_bal,y_pred))
                                              precision recall f1-score support
                                                                                                                       122
122
122
122
122
122
122
                  accuracy
macro avg
weighted avg
                                                                                                 0.73
0.73
0.73
                                                                                                                       854
854
854
                                                    0.73
0.73
                                                                        0.73
0.73
  In [82= train_score= accuracy_score(y_bal,sv1.predict(x_bal))
train_score
 Out[82]: 0.8125568698817106
  In [83_ # saving the model
                   import pickle
pickle.dump(sv1,open('thyroid_1_model.pkl','wb'))
  In [85_ features = np.array([[0,0,0,0,0.000000,0.0,0.0,1.00,0.0,40.0]])
print(label_encoder.inverse_transform(sv1.predict(features)))
                    ['binding protein']
                    /opt/conda/envs/Python-3.9/lib/python3.9/site-packages/sklearn/base.py:450: UserWarning: X does not have valid feature names, but SVC was fitted with feature names warnings.warn(
 In [86. type(features)
 Out[86]: numpy.ndarray
  In [87_ pickle.dump(label_encoder,open('label_encoder.pkl','wb'))
  In [88_ data['target'].unique()
In [89_ y['target'].unique()
 Out[89]: array([5, 4, 1, 6, 2, 3, 0])
  thyroid_1_model.pkl
label_encoder.pkl
                    label_encoder.pkl
thyroid_1_model.pkl
thyroid_disease_new.tgz
                 [pip install watson-machine-learning-client —upgrade

Collecting watson-machine-learning-client —upgrade

Collecting watson-machine-learning-client —upgrade —upgrade
   In [92_ !pip install watson-machine-learning-client --upgrade
```

