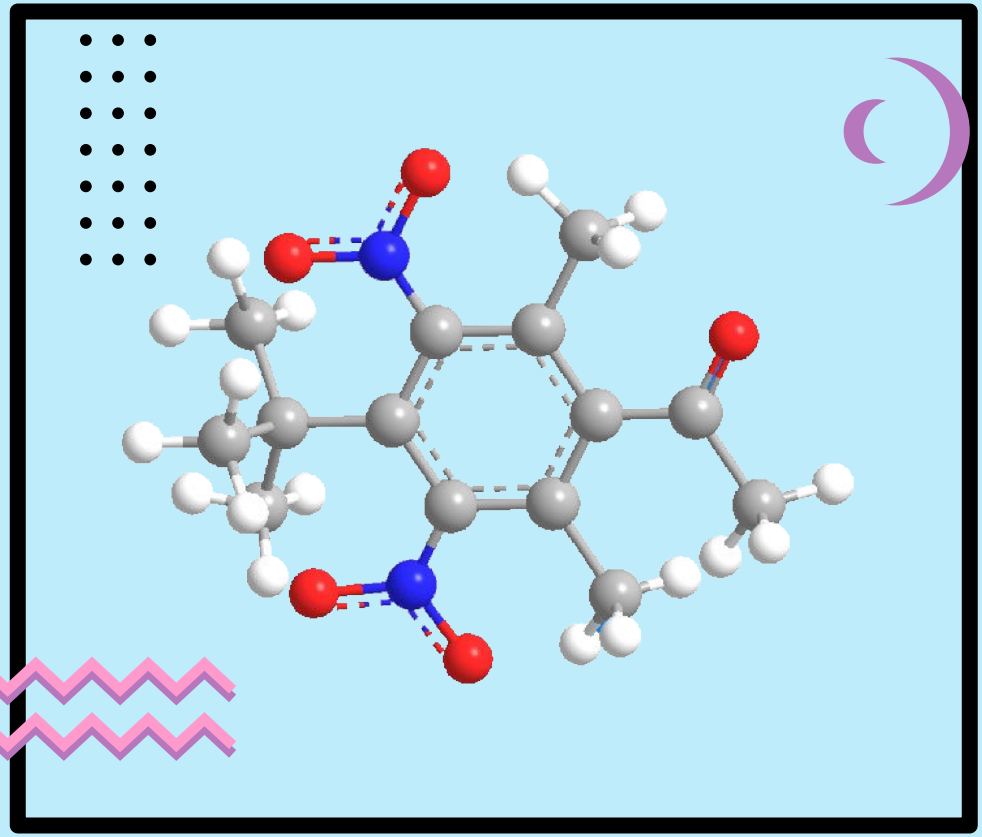

Chemical Classification

Musk OR Non-Musk

Table of Contents

Introduction

The given dataset contains details about organic chemical compounds including their chemical features, isomeric conformation, names and the classes in which they are classified. The compounds are classified as either ‘Musk’ or ‘Non-Musk’ compounds. Your task is to build a classification model on the given data using any Deep Learning approach that you deem appropriate viz. Multi-Layer Perceptron, CNN, RNN, etc. or you could also use transfer learning approaches through selection of appropriate pre-trained model. The data has to be split in a 80:20 ratio for training and validation datasets. You can perform whatever preprocessing and post-processing operations on the data that may help you improve the performance of your model. You are required to report the performance measures of the model viz. Accuracy(Training and Validation) and Loss(Training and Validation) graphs, F1 score, precision, recall, etc. along with a well detailed report of what models, pre-processing, post-processing approaches you have used and why you chose to use these approaches.

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About Me

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Computer science Engineer

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Work Process

Analysis

Creating classifier

Training

Testing & Deploy

Code Demo

```
[ ]  # Importing Libraries

     import numpy as np
     import pandas as pd
     import matplotlib.pyplot as plt
     import seaborn as sns
```

```
[ ]  dataset = pd.read_csv("/content/drive/My Drive/Web Data/musk_csv.csv")
```

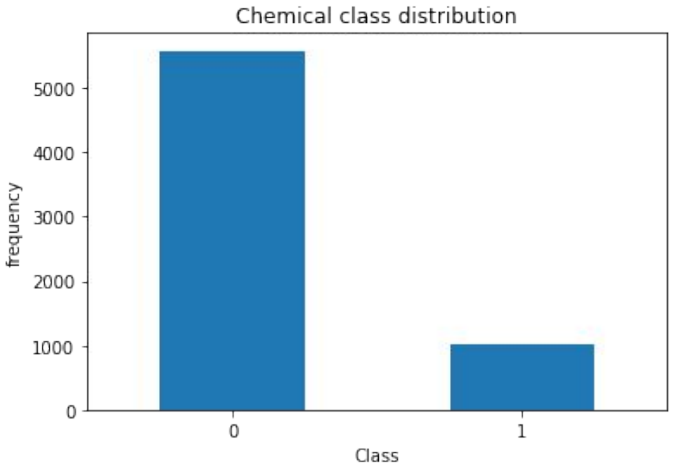
```
     dataset.head()
```

	ID	molecule_name	conformation_name	f1	f2	f3	f4	f5	f6	f7	f8	f9	f10	f11	f12	f13	f14	f15	f16	f17	f18	f19
0	1	MUSK-211	211_1+1	46	-108	-60	-69	-117	49	38	-161	-8	5	-323	-220	-113	-299	-283	-307	-31	-106	-227
1	2	MUSK-211	211_1+10	41	-188	-145	22	-117	-6	57	-171	-39	-100	-319	-111	-228	-281	-281	-300	54	-149	-98
2	3	MUSK-211	211_1+11	46	-194	-145	28	-117	73	57	-168	-39	-22	-319	-111	-104	-283	-282	-303	52	-152	-97
3	4	MUSK-211	211_1+12	41	-188	-145	22	-117	-7	57	-170	-39	-99	-319	-111	-228	-282	-281	-301	54	-150	-98
4	5	MUSK-211	211_1+13	41	-188	-145	22	-117	-7	57	-170	-39	-99	-319	-111	-228	-282	-281	-301	54	-150	-98

5 rows × 170 columns

Analysis

There are 6597 total chemicals



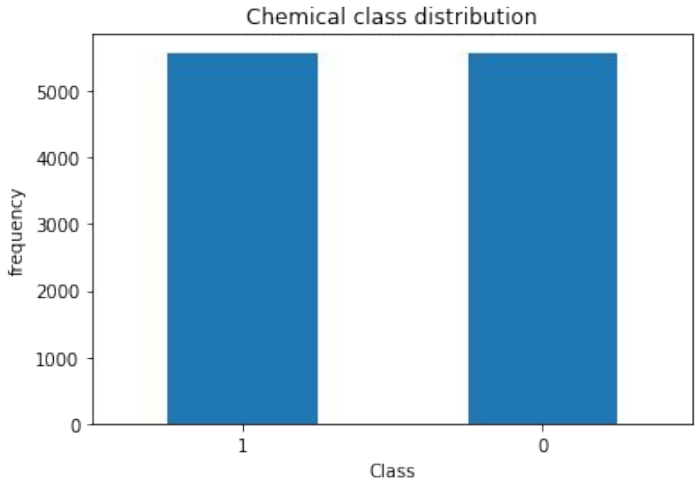
6597 Total

1017 Musk

5581 Non Musk

Analysis

Data Balancing



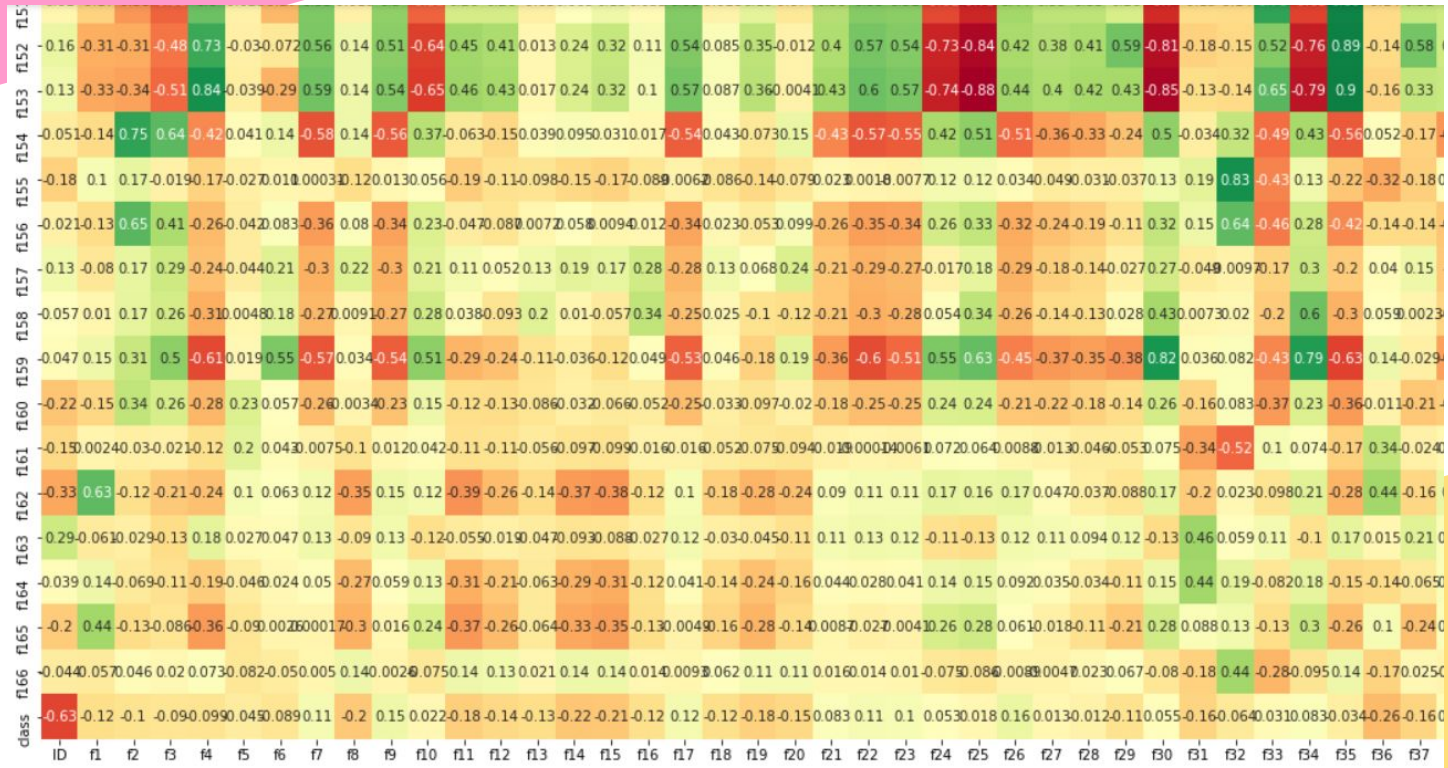
11162 Total

5581 Musk

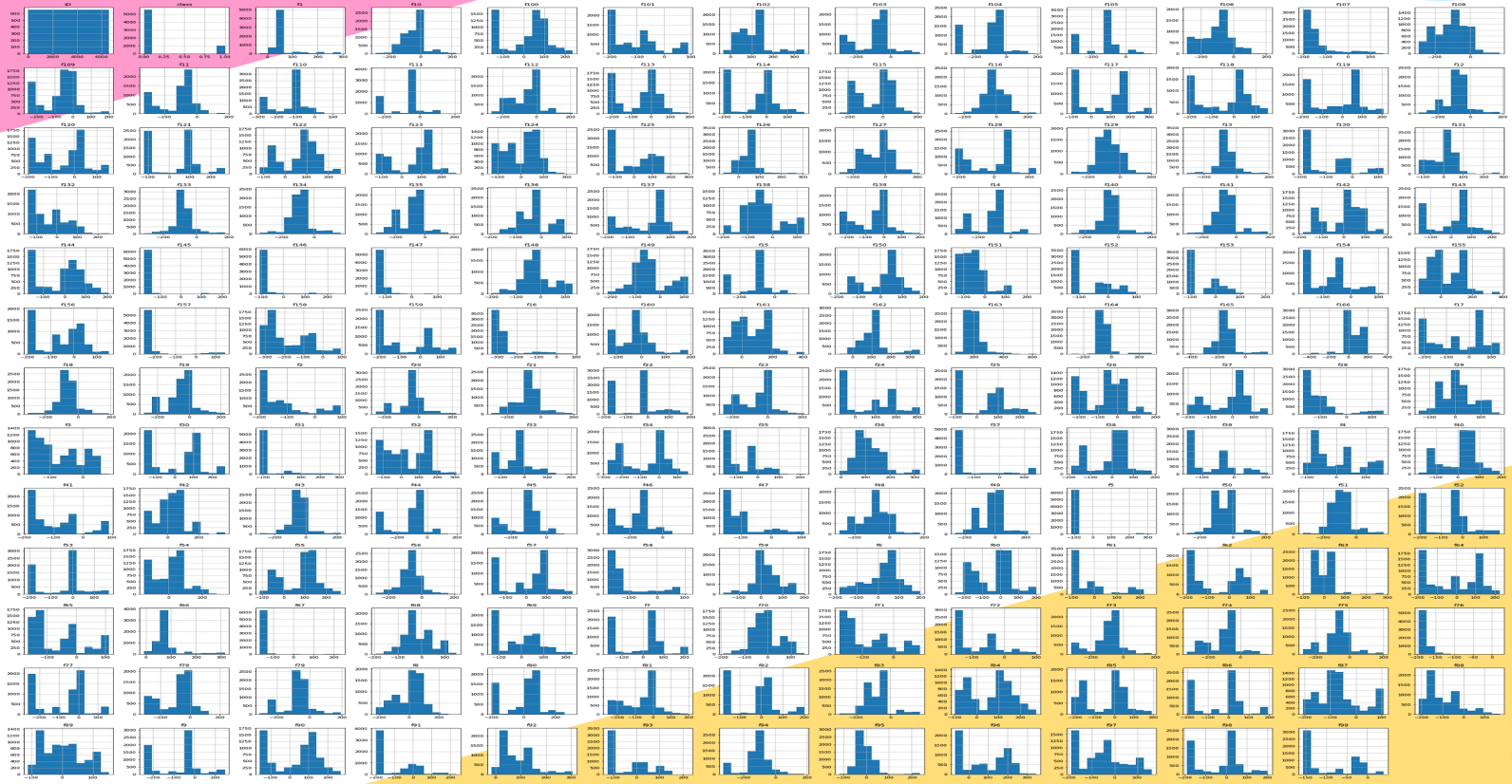
5581 Non Musk

Combine Sampling Method

Samples of Heatmap



Samples of Histogram



Work Flow

Splitting Train / Test

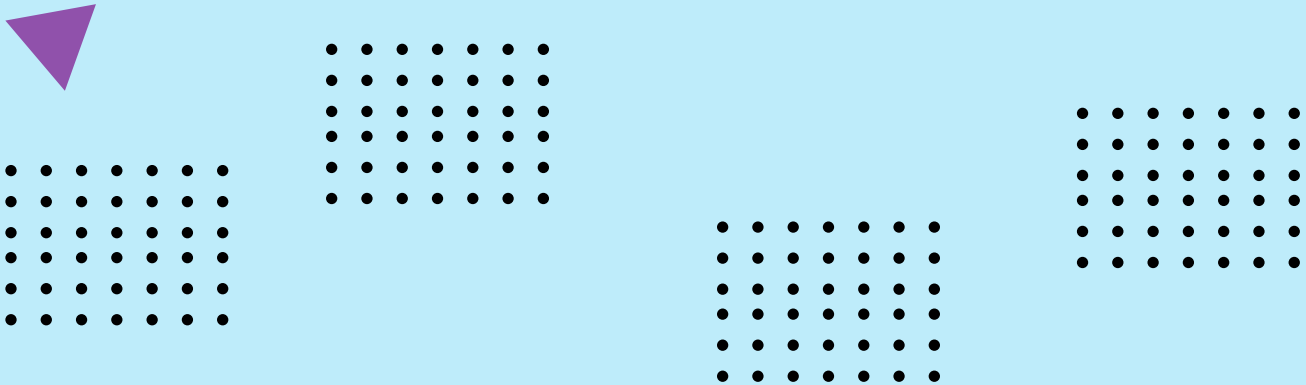
Used sklearn.model_selection import train_test_split

Scaling Variables

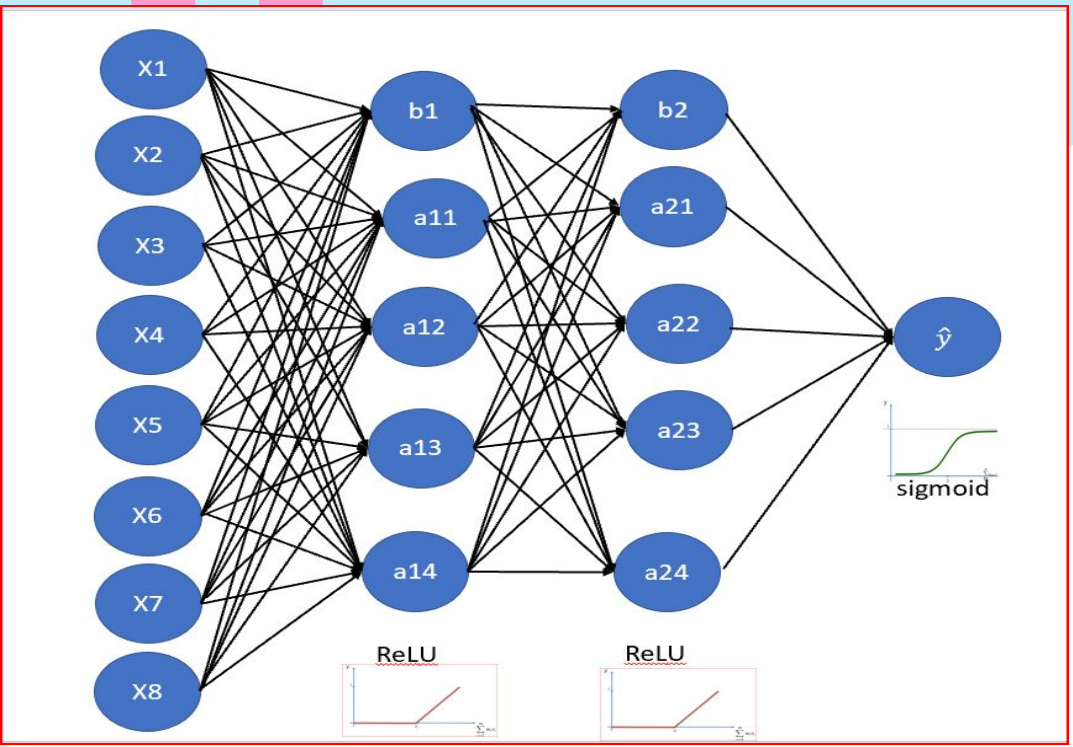
Used sklearn.preprocessing import standardscaler

Model

Used keras,fit into it and test it



Accuracy



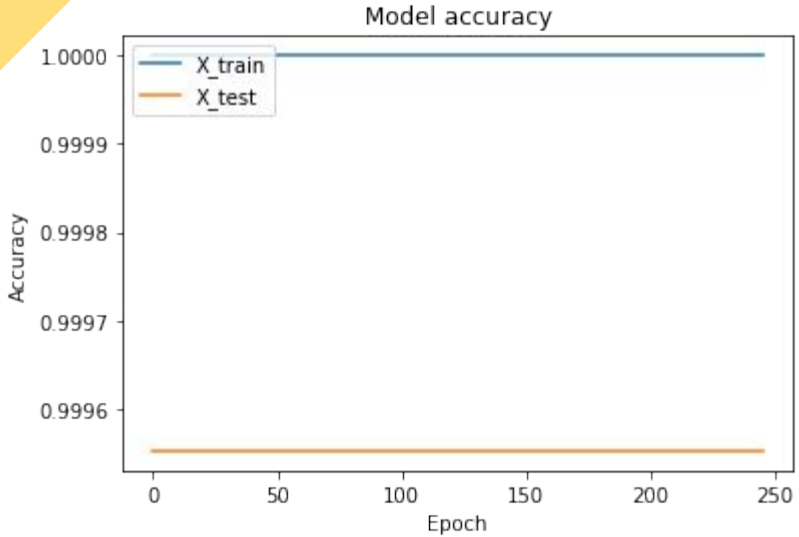
batch_size=40,epochs=246

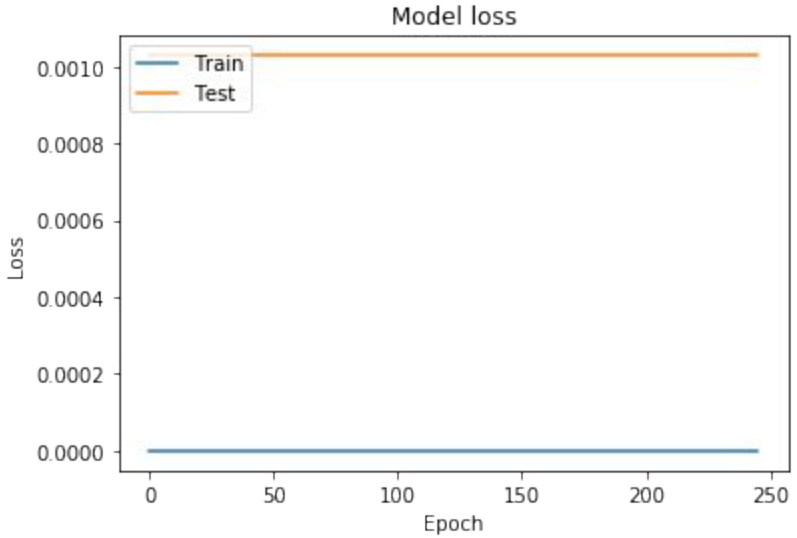
95.32%

batch_size=10,epochs=893

99.14%

Training & Validation





Recall & F-1 Score

	precision	recall	f1-score	support
0	0.50	1.00	0.67	1122
1	0.00	0.00	0.00	1111
accuracy			0.50	2233
macro avg	0.25	0.50	0.33	2233
weighted avg	0.25	0.50	0.34	2233

H5 model trained

```
classifier.save('chemical_classifier.h5')
```

```
from keras.models import load_model
```

```
new_model = load_model('chemical_classifier.h5')
```

```
new_model.summary()
```

```
Model: "sequential_1"
_________________________________________________________________
Layer (type)                 Output Shape              Param #
=================================================================
dense_1 (Dense)              (None, 84)                14028
_________________________________________________________________
dense_2 (Dense)              (None, 84)                7140
_________________________________________________________________
dense_3 (Dense)              (None, 1)                 85
=================================================================
Total params: 21,253
Trainable params: 21,253
Non-trainable params: 0
_________________________________________________________________
```

```
new_model.get_weights()
```

```
        [-0.14923573],
        [-0.14868845],
        [-0.10130396],
        [ 0.16415805],
        [-0.24770512],
```

Used For Development & upload

Thank You