

# Configuration Manual

MSc Research Project  
Programme Name

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MSc Project Submission Sheet  
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Master of Science in Cyber Security 2021-2022

**Programme:** ..... **Year:** .....

**Module:** .....

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**Submission Due Date:** .....

Towards improved phishing detection from URLs, using supervised machine learning

**Project Title:** .....

**Word Count:** 734 **Page Count:** 7

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# Configuration Manual

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Student ID: X20218061

## 1 Introduction to Configuration Manual

The manual outlines the Software and tools used to implement the Project. The manual contains the steps and instructions used to install the required software to implement the project and the systems used to get the required results.

## 2 Hardware Specification Details

Due to the high resource need and requirements for the Machine Learning algorithm and process. The Hardware specification is outlined below-:

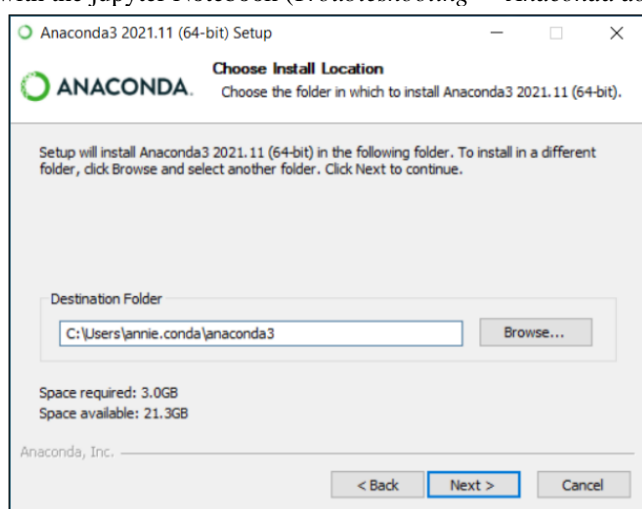
### Hardware

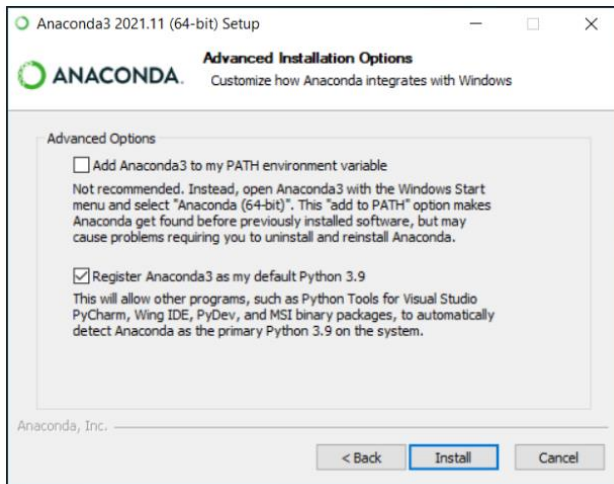
Technical Environment Hardware

Physical Device	
Device Model	Dell Inspiron Build Machine
Processor	Intel(R) Core (TM) i5-3570 CPU @ 3.40GHz 3.40 GHz
Installed RAM	32.0 GB RAM
System Type	Windows Operating System
HardDrive	500GB SSD Hard drive

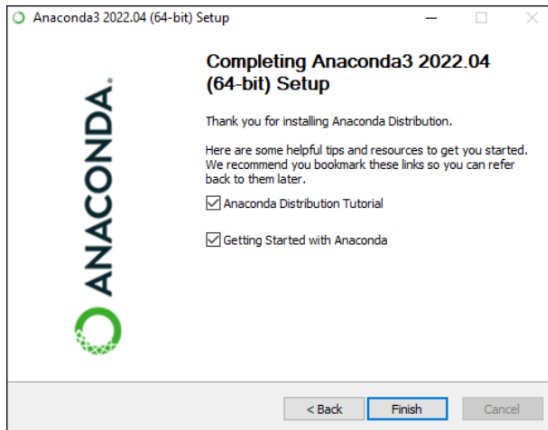
### Software Specification

Installed Anaconda for Windows using the below steps. Anaconda was the preferred method as it came added with the jupyter Notebook (*Troubleshooting — Anaconda documentation, no date*)

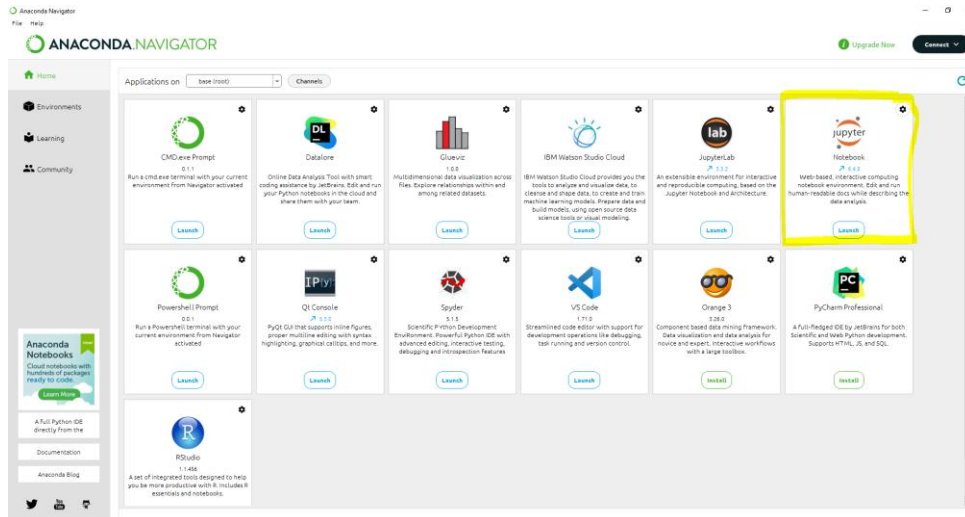




12. After a successful installation you will see the "Thanks for installing Anaconda" dialog box:



13. If you wish to read more about Anaconda.org and how to get started with Anaconda, check the boxes "Anaconda Distribution Tutorial" and "Learn more about Anaconda". Click the **Finish** button.



### 3 Setup of the Machine learning jupyter Notebook version 6.4.8

The project is aimed at using python programming language due to the availability of mature and well tested data management and machine learning libraries such as scikit, pandas, NumPy to help analyze and train the dataset. Python has been the preferred programming language due to the availability of these libraries and the ease of using the language. The main object would be to tag either the websites belonging to phishing or legitimate using the machine learning framework. After this, the results would be analyzed in relation to precision, recall, and accuracy. The result would also show the details of the program and how effectively it identifies the websites

#### 3.1 Feature Extraction Procedures

Feature extraction was needed because the original raw data cannot be used for the machine learning model. Most of the useful information is captured as a result of the extracted features. This creates a smaller set of features captured from the raw data. We would extract useful information from the .csv data for the purpose of the machine learning classifiers to work from.

##### Features Extracted as Outlined Below

###### Address Bar-based Features

- Domain of URL
- IP Address in URL
- "@" Symbol in URL
- getDomainEntropy
- Length of URL
- Depth of URL
- Redirection "/" in URL
- "http/https" in Domain name
- Using URL Shortening Services "TinyURL"
- Prefix or Suffix "-" in Domain Each of these features are explained and the coded below:

###### Domain based Features

- DNS Record
- Website Traffic
- Age of Domain
- End Period of Domain
- Each of these features are explained and the coded below:

###### HTML & Javascript based Features

- IFrame Redirection
- Status Bar Customization
- Disabling Right Click
- Website Forwarding

### 3.1.1 Good URLs Procedures

```
jupyter all_features_extraction Last Checkpoint: 3 minutes ago (autosaved)
File Edit View Insert Cell Kernel Widgets Help
+ - - - - - Run - - - - - Markdown - - - - - True

In [1]: #importing required packages for this module
import pandas as pd

Good URL

Now we get a list of good URLs from https://www.unb.ca/cic/datasets/url-2016.html
This list is not necessarily up to date but for my machine learning process, I believe the result would meet my classification needs.

In [2]: data0 = pd.read_csv("CSVs/Benign_list_big_final.csv")
data0.shape

Out[2]: (35378, 1)
```

```
In [36]: #Extracting the features & storing them in a List
import sys
legi_5_features = []
dead_5_urls = []
label = 0

for i in range(0, 5):
    #try:
    url = legiurl_5['URLs'][i]
    #url = getDomain(url)
    print(i, url, len(legiurl['URLs'][i]))
    legi_5_features.append(featureExtraction(url,i,label))

print(legi_5_features)
```

```
In [37]: #Extracting the features & storing them in a List
from datetime import datetime

legi_features = []
dead_urls = []
label = 0
start_time = datetime.now()

for i in range(0, 30):
    url = legiurl['URLs'][i]
    domain = getDomain(url)
    print(i, domain, len(legiurl['URLs'][i]))
    legi_features.append(featureExtraction(url,i,label))

end_time = datetime.now()
print('Duration: {}'.format(end_time - start_time))

print(legi_features)
```

### 3.1.2 Bad URLs Procedures

#### Bad URLs

```
In [3]: #####Loading the phishing URLs data to dataframe#####
from urllib.parse import urlparse,urlencode
import re
data1 = pd.read_csv("CSVs/online-valid-status.csv") #, errors='ignore')
domain_list = []
def getDomain(url):
    domain = urlparse(url).netloc
    if re.match(r"^\www.",domain):
        domain = domain.replace("www.", "")
    return domain
```

```
In [40]: #Extracting the features & storing them in a list
from datetime import datetime

phishing_features = []
dead_urls = []
label = 1
start_time = datetime.now()

for i in range(0, 30):
    url = phishurl['url'][i]
    domain = getDomain(url)
    print(i, domain, len(phishurl['url'][i]))
    phishing_features.append(featureExtraction(url,i,label))

end_time = datetime.now()
print('Duration: {}'.format(end_time - start_time))

print(phishing_features)
```

### 3.1.3 Extracted 5000 records Randomly from both the Good URLs and Bad URLs

#### Extraction Process

This is a large dataset. I would extract 5,000 records randomly for the purpose of this exercise

```
In [4]: #Collecting 5,000 Legitimate URLs randomly
legiurl = data0.sample(n = 5000, random_state = 20).copy()
legiurl = legiurl.reset_index(drop=True)
legiurl.head()
```

```
Out[4]:
      URLs
0  http://cheezburger.com/70977793/video-game-new-...
1  http://motthegioi.vn/hoi-ky-mcnamara/ky-40-duo-...
2  http://thenextweb.com/socialmedia/2014/11/05/w-...
3  http://espn.go.com/nfl/insider/story/_id/1296-...
4  http://bestblackhatforum.com/Thread-GET-A-Minu...
```

```
In [5]: #Collecting 5,000 Phishing URLs randomly
phishurl = data1.sample(n = 5000, random_state = 12).copy()
phishurl = phishurl.reset_index(drop=True)
phishurl.head()
```

```
Out[5]:
   phish_id      url      phish_detail_url  submission_time  verified  verification_time  online  targ
0  7593537  http://www.myjascoseb.myceojascob...  http://www.phishtank.com/phish_detail.php?  2022-07-  yes  2022-07-  yes  NICC
   14T04:38:32+00:00
1  7582352  http://www.sacaivseoseincasseid.cccaseasoc...  http://www.phishtank.com/phish_detail.php?  2022-07-  yes  2022-07-  yes  NICC
   08T13:42:36+00:00
2  7609718  http://www.acocceon.aseocoon.selfie.it/  http://www.phishtank.com/phish_detail.php?  2022-07-  yes  2022-07-  yes  AEC
   21T09:51:45+00:00
```

## References

*Troubleshooting — Anaconda documentation* (no date). Available at: <https://docs.anaconda.com/anaconda/user-guide/troubleshooting/> (Accessed: 30 October 2022).