

### 3. DATA PREPROCESSING

January 2, 2026

#### Importing Libraries

```
[2]: import pandas as pd
import numpy as np
```

#### Reading the data

```
[3]: data = pd.read_csv('data/transformed/final_raw_data.csv')

data.head()
```

```
[3]:
```

	url	label
0	https://adoaeco.cn/Loggin	phishing
1	https://gageparkhighschool.com/QTeuUe	phishing
2	https://wnnox.miraltek.cfd/qzxn3	phishing
3	https://halfetitur.com/?token=r2IOIU0FEHfPf5Dn	phishing
4	https://yqcjl.miraltek.cfd/plis0	phishing

```
[4]: data.duplicated(subset=['url']).sum()
```

```
[4]: np.int64(0)
```

```
[5]: # URL Components features
url_components_df = pd.read_csv('data/transformed/1.url_components_data.csv')

url_components_df.head()
```

```
[5]:
```

	url	label	protocol	\
0	https://adoaeco.cn/Loggin	phishing	https	
1	https://gageparkhighschool.com/QTeuUe	phishing	https	
2	https://wnnox.miraltek.cfd/qzxn3	phishing	https	
3	https://halfetitur.com/?token=r2IOIU0FEHfPf5Dn	phishing	https	
4	https://yqcjl.miraltek.cfd/plis0	phishing	https	

  

	domain	subdomain	tld	sld	path	\
0	adoaeco.cn	NaN	cn	adoaeco	/Loggin	
1	gageparkhighschool.com	NaN	com	gageparkhighschool	/QTeuUe	
2	wnnox.miraltek.cfd	wnnox	cfd	miraltek	/qzxn3	
3	halfetitur.com	NaN	com	halfetitur	/	

```

4      yqcjl.miraltek.cfd      yqcjl  cfd      miraltek  /plis0

      query
0      NaN
1      NaN
2      NaN
3  token=r2I0IU0FEHfPf5Dn
4      NaN

```

```

[6]: # URL component length features data
len_features_df = pd.read_csv('data/transformed/2.component_len_features_data.
    ↪csv')

len_features_df.head()

```

```

[6]:
      url      label  url_len  \
0  https://adoaeco.cn/Loggin  phishing      25
1  https://gageparkhighschool.com/QTeuUe  phishing      37
2  https://wnnox.miraltek.cfd/qzxn3  phishing      32
3  https://halfetitur.com/?token=r2I0IU0FEHfPf5Dn  phishing      46
4  https://yqcjl.miraltek.cfd/plis0  phishing      32

   domain_len  path_len  query_len  url_depth  subdomain_count
0           10         6          0          1                1
1           22         6          0          1                1
2           18         5          0          1                1
3           14         0         22          1                1
4           18         5          0          1                1

```

```

[7]: # Domain features data
domain_features_df = pd.read_csv('data/transformed/3.domain_features_data.csv')

domain_features_df.head()

```

```

[7]:
      url      label  tld  tld_len  \
0  https://adoaeco.cn/Loggin  phishing  cn          2
1  https://gageparkhighschool.com/QTeuUe  phishing  com          3
2  https://wnnox.miraltek.cfd/qzxn3  phishing  cfd          3
3  https://halfetitur.com/?token=r2I0IU0FEHfPf5Dn  phishing  com          3
4  https://yqcjl.miraltek.cfd/plis0  phishing  cfd          3

   url_has_ipv4  url_has_port
0         False         False
1         False         False
2         False         False
3         False         False
4         False         False

```

```
[8]: # SLD features data
sld_features_df = pd.read_csv('data/transformed/4.sld_features_data.csv')

sld_features_df.head()
```

```
[8]:
```

	url	label	\
0	https://adoaeco.cn/Loggin	phishing	
1	https://gageparkhighschool.com/QTeuUe	phishing	
2	https://wnnox.miraltek.cfd/qzxn3	phishing	
3	https://halfetitur.com/?token=r2I0IU0FEHfPf5Dn	phishing	
4	https://yqcjl.miraltek.cfd/plis0	phishing	

  

	sld	sld_len	sld_has_digit	sld_has_hyphen	sld_token_count
0	adoaeco	7	False	False	1
1	gageparkhighschool	18	False	False	1
2	miraltek	8	False	False	1
3	halfetitur	10	False	False	1
4	miraltek	8	False	False	1

```
[9]: # Character features data
char_feature_df = pd.read_csv('data/transformed/5.char_features_data.csv')

char_feature_df.head()
```

```
[9]:
```

	url	label	dot_count_domain	\
0	https://adoaeco.cn/Loggin	phishing	1	
1	https://gageparkhighschool.com/QTeuUe	phishing	1	
2	https://wnnox.miraltek.cfd/qzxn3	phishing	2	
3	https://halfetitur.com/?token=r2I0IU0FEHfPf5Dn	phishing	1	
4	https://yqcjl.miraltek.cfd/plis0	phishing	2	

  

	hyphen_count_domain_path	underscore_count_path_query	slash_count	\
0	0	0	3	
1	0	0	3	
2	0	0	3	
3	0	0	3	
4	0	0	3	

  

	digit_count	alphabet_count	spl_char_count
0	0	20	5
1	0	32	5
2	1	25	6
3	4	35	7
4	1	25	6

```
[10]: # Entropy features data
entropy_feature_df = pd.read_csv('data/transformed/6.entropy_feature_data.csv')
```

```
entropy_feature_df.head()
```

```
[10]:
```

	url	label	url_entropy \
0	https://adoaeco.cn/Loggin	phishing	3.863465
1	https://gageparkhighschool.com/QTeuUe	phishing	4.208925
2	https://wnnox.miraltek.cfd/qzxn3	phishing	4.452820
3	https://halfetitur.com/?token=r2I0IU0FEHfPf5Dn	phishing	4.760096
4	https://yqcjl.miraltek.cfd/plis0	phishing	4.241729

  

	domain_entropy	sld_entropy	path_entropy
0	2.721928	2.235926	2.521641
1	3.629220	3.419382	2.521641
2	3.947703	3.000000	2.584963
3	3.664498	3.121928	-0.000000
4	3.836592	3.000000	2.584963

```
[11]: # Token features data
token_feature_df = pd.read_csv('data/transformed/7.token_features_data.csv')

token_feature_df.head()
```

```
[11]:
```

	url	label \
0	https://adoaeco.cn/Loggin	phishing
1	https://gageparkhighschool.com/QTeuUe	phishing
2	https://wnnox.miraltek.cfd/qzxn3	phishing
3	https://halfetitur.com/?token=r2I0IU0FEHfPf5Dn	phishing
4	https://yqcjl.miraltek.cfd/plis0	phishing

  

	domain_token_count	path_token_count	total_tokens	avg_token_length
0	2	1	3	5.00
1	2	1	3	9.00
2	3	1	4	5.25
3	2	1	3	8.50
4	3	1	4	5.25

```
[12]: # Hexadecimal feature data
hex_feature_df = pd.read_csv('data/transformed/8.hex_features_data.csv')

hex_feature_df.head()
```

```
[12]:
```

	url	label	has_hex \
0	https://adoaeco.cn/Loggin	phishing	False
1	https://gageparkhighschool.com/QTeuUe	phishing	False
2	https://wnnox.miraltek.cfd/qzxn3	phishing	False
3	https://halfetitur.com/?token=r2I0IU0FEHfPf5Dn	phishing	False
4	https://yqcjl.miraltek.cfd/plis0	phishing	False

	hex_char_count	hex_ratio
0	0	0.0
1	0	0.0
2	0	0.0
3	0	0.0
4	0	0.0

```
[13]: df_dict = {
    'URL components' : url_components_df,
    'Length features' : len_features_df,
    'Domain features' : domain_features_df,
    'SLD features' : sld_features_df,
    'Character features' : char_feature_df,
    'Entropy features' : entropy_feature_df,
    'Token features' : token_feature_df,
    'Hexadecimal features' : hex_feature_df
}
```

### Handling null values

```
[14]: def null_cols(df):
    null_counts = df.isnull().sum()
    null_cols = null_counts[null_counts > 0]

    if not null_counts.empty:
        print(null_cols)
    else:
        print('No null values found')
```

```
[15]: for df_name, df in df_dict.items():
    print(df_name)
    null_cols(df)
    print()
```

URL components

domain	2283
subdomain	64887
tld	2435
sld	2286
path	48380
query	214330

dtype: int64

Length features

Series([], dtype: int64)

Domain features

```

tld      2435
dtype: int64

SLD features
sld      2286
dtype: int64

Character features
Series([], dtype: int64)

Entropy features
Series([], dtype: int64)

Token features
Series([], dtype: int64)

Hexadecimal features
Series([], dtype: int64)

```

The URL Components data, Domain features and SLD features consists of null values

```
[16]: domain_features_df[domain_features_df['tld'].isnull()]
```

```

[16]:
      url      label  tld  \
2994      https://91.92.241.186  phishing  NaN
3667      https://140.99.164.68/x0  phishing  NaN
6300      https://31.172.87.101/x0  phishing  NaN
14536      https://43.153.99.18  phishing  NaN
16733      https://185.187.56.126  phishing  NaN
...
252716      http://191.101.7.221/fire/aasdqwe  phishing  NaN
252811      http://91.239.25.38:6892  phishing  NaN
252961  http://178.217.186.224/panel/etc/info/toke/cp...  phishing  NaN
252990      http://185.75.46.73/information.cgi  phishing  NaN
252997  http://38.118.40.209/CFIDE/debug/serveur.html?...  phishing  NaN

      tld_len  url_has_ipv4  url_has_port
2994         0          True          False
3667         0          True          False
6300         0          True          False
14536        0          True          False
16733        0          True          False
...
252716        0          True          False
252811        0          True           True
252961        0          True          False
252990        0          True          False

```

```
252997      0      True      False
```

```
[2435 rows x 6 columns]
```

```
[17]: sld_features_df[sld_features_df['sld'].isnull()]
```

```
[17]:
```

	url	label	sld \
2994	https://91.92.241.186	phishing	NaN
3667	https://140.99.164.68/x0	phishing	NaN
6300	https://31.172.87.101/x0	phishing	NaN
14536	https://43.153.99.18	phishing	NaN
16733	https://185.187.56.126	phishing	NaN
...	...	...	...
252716	http://191.101.7.221/fire/aasdqwe	phishing	NaN
252811	http://91.239.25.38:6892	phishing	NaN
252961	http://178.217.186.224/panel/etc/info/toke/cp...	phishing	NaN
252990	http://185.75.46.73/information.cgi	phishing	NaN
252997	http://38.118.40.209/CFIDE/debug/serveur.html?...	phishing	NaN

  

	sld_len	sld_has_digit	sld_has_hyphen	sld_token_count
2994	0	False	False	1
3667	0	False	False	1
6300	0	False	False	1
14536	0	False	False	1
16733	0	False	False	1
...	...	...	...	...
252716	0	False	False	1
252811	0	False	False	1
252961	0	False	False	1
252990	0	False	False	1
252997	0	False	False	1

```
[2286 rows x 7 columns]
```

The URLs where TLDs & SLDs having null values are mostly IP address based URLs. So, the numerical features dependent on TLD & SLD will be 0. In URL components data, there are many null values in Domain, SLD and TLD. These are IP address based URLs. Other features also have many null values since we are considering numerical features, we will ignore those values.

### Combining all the features into a single dataset

```
[18]: df = pd.DataFrame()      # dataframe to store all the processed features
```

```
[19]: # URL components data
```

```
df['has_https'] = url_components_df['protocol'].map(lambda x: 1 if x == 'https' else 0)
```

```
[20]: # URL Length features data

df[['url_len', 'domain_len', 'path_len', 'query_len', 'url_depth', 'subdomain_count']] =
    len_features_df.select_dtypes('number')
```

```
[21]: # Domain features data

df['tld_len'] = domain_features_df['tld_len']
df[['url_has_ipv4', 'url_has_port']] =
    domain_features_df[['url_has_ipv4', 'url_has_port']].astype('int64')
```

```
[22]: # SLD features data

df['sld_len'] = sld_features_df['sld_len']
df[['sld_has_digit', 'sld_has_hyphen']] =
    sld_features_df[['sld_has_digit', 'sld_has_hyphen']].astype('int64')
df['sld_token_count'] = sld_features_df['sld_token_count']
```

```
[23]: # Character features data

df[['dot_count_domain', 'hyphen_count_domain_path', 'underscore_count_path_query', 'slash_count',
    char_feature_df.select_dtypes('number')
```

```
[24]: # Entropy features data

df[['url_entropy', 'domain_entropy', 'sld_entropy', 'path_entropy']] =
    entropy_feature_df.select_dtypes('number')
```

```
[25]: # Token features data

df[['domain_token_count', 'path_token_count', 'total_tokens', 'avg_token_length']] =
    token_feature_df.select_dtypes('number')
```

We will ignore Hexadecimal-based features since the hexadecimal features in the data are very low and its contribution is very less in predictions.

```
[26]: # Adding label

df['class'] = url_components_df['label'].apply(lambda x: 1 if x == 'phishing'
    else 0)
```

```
[27]: df.head()
```

```
[27]:   has_https  url_len  domain_len  path_len  query_len  url_depth  \
0          1      25          10         6          0          1
1          1      37          22         6          0          1
2          1      32          18         5          0          1
3          1      46          14         0         22          1
```



4	1	32	18	5	0	1	
---	---	----	----	---	---	---	--

  

	subdomain_count	tld_len	url_has_ipv4	url_has_port	...	spl_char_count	\
0	1	2	0	0	...	5	
1	1	3	0	0	...	5	
2	1	3	0	0	...	6	
3	1	3	0	0	...	7	
4	1	3	0	0	...	6	

  

	url_entropy	domain_entropy	sld_entropy	path_entropy	domain_token_count	\
0	3.863465	2.721928	2.235926	2.521641	2	
1	4.208925	3.629220	3.419382	2.521641	2	
2	4.452820	3.947703	3.000000	2.584963	3	
3	4.760096	3.664498	3.121928	-0.000000	2	
4	4.241729	3.836592	3.000000	2.584963	3	

  

	path_token_count	total_tokens	avg_token_length	class
0	1	3	5.00	1
1	1	3	9.00	1
2	1	4	5.25	1
3	1	3	8.50	1
4	1	4	5.25	1

[5 rows x 30 columns]

```
[28]: print(f'The combined dataset consists of {df.shape[0]} rows and {df.shape[1]}_
      ↪columns')
```

The combined dataset consists of 253051 rows and 30 columns

```
[29]: df.columns
```

```
[29]: Index(['has_https', 'url_len', 'domain_len', 'path_len', 'query_len',
          'url_depth', 'subdomain_count', 'tld_len', 'url_has_ipv4',
          'url_has_port', 'sld_len', 'sld_has_digit', 'sld_has_hyphen',
          'sld_token_count', 'dot_count_domain', 'hyphen_count_domain_path',
          'underscore_count_path_query', 'slash_count', 'digit_count',
          'alphabet_count', 'spl_char_count', 'url_entropy', 'domain_entropy',
          'sld_entropy', 'path_entropy', 'domain_token_count', 'path_token_count',
          'total_tokens', 'avg_token_length', 'class'],
          dtype='object')
```

```
[30]: df.dtypes
```

```
[30]: has_https      int64
      url_len        int64
      domain_len     int64
      path_len       int64
```

query_len	int64
url_depth	int64
subdomain_count	int64
tld_len	int64
url_has_ipv4	int64
url_has_port	int64
sld_len	int64
sld_has_digit	int64
sld_has_hyphen	int64
sld_token_count	int64
dot_count_domain	int64
hyphen_count_domain_path	int64
underscore_count_path_query	int64
slash_count	int64
digit_count	int64
alphabet_count	int64
spl_char_count	int64
url_entropy	float64
domain_entropy	float64
sld_entropy	float64
path_entropy	float64
domain_token_count	int64
path_token_count	int64
total_tokens	int64
avg_token_length	float64
class	int64
dtype:	object

```
[31]: df.duplicated().sum()
```

```
[31]: np.int64(29531)
```

```
[32]: df.drop_duplicates(keep='first', ignore_index=True, inplace=True)
```

```
[33]: # Saving the dataset

df.to_csv(r'data/processed/processed_data.csv', index=False)
```