

Importing Necessary Packages

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

Data Preprocessing

```
df=pd.read_csv(r"C:\Users\Arigala.Adarsh\Downloads\
csgo_round_snapshots.csv")
```

```
df.head()
```

	time_left	ct_score	t_score	map	bomb_planted	ct_health
0	175.00	0.0	0.0	de_dust2	False	500.0
1	156.03	0.0	0.0	de_dust2	False	500.0
2	96.03	0.0	0.0	de_dust2	False	391.0
3	76.03	0.0	0.0	de_dust2	False	391.0
4	174.97	1.0	0.0	de_dust2	False	500.0

	ct_armor	t_armor	ct_money	...	t_grenade_flashbang
0	0.0	0.0	4000.0	...	0.0
1	400.0	300.0	600.0	...	0.0
2	294.0	200.0	750.0	...	0.0
3	294.0	200.0	750.0	...	0.0
4	192.0	0.0	18350.0	...	0.0

	ct_grenade_smokegrenade	t_grenade_smokegrenade
0	0.0	0.0
1	0.0	2.0
2	0.0	2.0
3	0.0	0.0
4	0.0	0.0

	ct_grenade_incendiarygrenade	t_grenade_incendiarygrenade
0	0.0	0.0
1	0.0	0.0
2	0.0	0.0
3	0.0	0.0

4	0.0	0.0
---	-----	-----

	ct_grenade_molotovgrenade	t_grenade_molotovgrenade \
0	0.0	0.0
1	0.0	0.0
2	0.0	0.0
3	0.0	0.0
4	0.0	0.0

	ct_grenade_decoygrenade	t_grenade_decoygrenade	round_winner
0	0.0	0.0	CT
1	0.0	0.0	CT
2	0.0	0.0	CT
3	0.0	0.0	CT
4	0.0	0.0	CT

[5 rows x 97 columns]

df.shape

(122410, 97)

df.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 122410 entries, 0 to 122409
Data columns (total 97 columns):
```

#	Column	Non-Null Count	Dtype
0	time_left	122410 non-null	float64
1	ct_score	122410 non-null	float64
2	t_score	122410 non-null	float64
3	map	122410 non-null	object
4	bomb_planted	122410 non-null	bool
5	ct_health	122410 non-null	float64
6	t_health	122410 non-null	float64
7	ct_armor	122410 non-null	float64
8	t_armor	122410 non-null	float64
9	ct_money	122410 non-null	float64
10	t_money	122410 non-null	float64
11	ct_helmets	122410 non-null	float64
12	t_helmets	122410 non-null	float64
13	ct_defuse_kits	122410 non-null	float64
14	ct_players_alive	122410 non-null	float64
15	t_players_alive	122410 non-null	float64
16	ct_weapon_ak47	122410 non-null	float64
17	t_weapon_ak47	122410 non-null	float64
18	ct_weapon_aug	122410 non-null	float64
19	t_weapon_aug	122410 non-null	float64
20	ct_weapon_awp	122410 non-null	float64

21	t_weapon_awp	122410	non-null	float64
22	ct_weapon_bizon	122410	non-null	float64
23	t_weapon_bizon	122410	non-null	float64
24	ct_weapon_cz75auto	122410	non-null	float64
25	t_weapon_cz75auto	122410	non-null	float64
26	ct_weapon_elite	122410	non-null	float64
27	t_weapon_elite	122410	non-null	float64
28	ct_weapon_famas	122410	non-null	float64
29	t_weapon_famas	122410	non-null	float64
30	ct_weapon_g3sg1	122410	non-null	float64
31	t_weapon_g3sg1	122410	non-null	float64
32	ct_weapon_galilar	122410	non-null	float64
33	t_weapon_galilar	122410	non-null	float64
34	ct_weapon_glock	122410	non-null	float64
35	t_weapon_glock	122410	non-null	float64
36	ct_weapon_m249	122410	non-null	float64
37	t_weapon_m249	122410	non-null	float64
38	ct_weapon_m4a1s	122410	non-null	float64
39	t_weapon_m4a1s	122410	non-null	float64
40	ct_weapon_m4a4	122410	non-null	float64
41	t_weapon_m4a4	122410	non-null	float64
42	ct_weapon_mac10	122410	non-null	float64
43	t_weapon_mac10	122410	non-null	float64
44	ct_weapon_mag7	122410	non-null	float64
45	t_weapon_mag7	122410	non-null	float64
46	ct_weapon_mp5sd	122410	non-null	float64
47	t_weapon_mp5sd	122410	non-null	float64
48	ct_weapon_mp7	122410	non-null	float64
49	t_weapon_mp7	122410	non-null	float64
50	ct_weapon_mp9	122410	non-null	float64
51	t_weapon_mp9	122410	non-null	float64
52	ct_weapon_negev	122410	non-null	float64
53	t_weapon_negev	122410	non-null	float64
54	ct_weapon_nova	122410	non-null	float64
55	t_weapon_nova	122410	non-null	float64
56	ct_weapon_p90	122410	non-null	float64
57	t_weapon_p90	122410	non-null	float64
58	ct_weapon_r8revolver	122410	non-null	float64
59	t_weapon_r8revolver	122410	non-null	float64
60	ct_weapon_sawedoff	122410	non-null	float64
61	t_weapon_sawedoff	122410	non-null	float64
62	ct_weapon_scar20	122410	non-null	float64
63	t_weapon_scar20	122410	non-null	float64
64	ct_weapon_sg553	122410	non-null	float64
65	t_weapon_sg553	122410	non-null	float64
66	ct_weapon_ssg08	122410	non-null	float64
67	t_weapon_ssg08	122410	non-null	float64
68	ct_weapon_ump45	122410	non-null	float64
69	t_weapon_ump45	122410	non-null	float64

```

70 ct_weapon_xm1014      122410 non-null float64
71 t_weapon_xm1014      122410 non-null float64
72 ct_weapon_deagle      122410 non-null float64
73 t_weapon_deagle       122410 non-null float64
74 ct_weapon_fiveseven   122410 non-null float64
75 t_weapon_fiveseven    122410 non-null float64
76 ct_weapon_usps         122410 non-null float64
77 t_weapon_usps         122410 non-null float64
78 ct_weapon_p250         122410 non-null float64
79 t_weapon_p250         122410 non-null float64
80 ct_weapon_p2000        122410 non-null float64
81 t_weapon_p2000        122410 non-null float64
82 ct_weapon_tec9         122410 non-null float64
83 t_weapon_tec9         122410 non-null float64
84 ct_grenade_hegrenade   122410 non-null float64
85 t_grenade_hegrenade   122410 non-null float64
86 ct_grenade_flashbang   122410 non-null float64
87 t_grenade_flashbang   122410 non-null float64
88 ct_grenade_smokegrenade 122410 non-null float64
89 t_grenade_smokegrenade 122410 non-null float64
90 ct_grenade_incendiarygrenade 122410 non-null float64
91 t_grenade_incendiarygrenade 122410 non-null float64
92 ct_grenade_molotovgrenade 122410 non-null float64
93 t_grenade_molotovgrenade 122410 non-null float64
94 ct_grenade_decoygrenade 122410 non-null float64
95 t_grenade_decoygrenade 122410 non-null float64
96 round_winner          122410 non-null object

```

dtypes: bool(1), float64(94), object(2)

memory usage: 89.8+ MB

df.describe()

	time_left	ct_score	t_score	ct_health \
count	122410.000000	122410.000000	122410.000000	122410.000000
mean	97.886922	6.709239	6.780435	412.106568
std	54.465238	4.790362	4.823543	132.293290
min	0.010000	0.000000	0.000000	0.000000
25%	54.920000	3.000000	3.000000	350.000000
50%	94.910000	6.000000	6.000000	500.000000
75%	166.917500	10.000000	10.000000	500.000000
max	175.000000	32.000000	33.000000	500.000000

	t_health	ct_armor	t_armor	ct_money \
count	122410.000000	122410.000000	122410.000000	122410.000000
mean	402.714500	314.142121	298.444670	9789.023773
std	139.919033	171.029736	174.576545	11215.042286
min	0.000000	0.000000	0.000000	0.000000
25%	322.000000	194.000000	174.000000	1300.000000
50%	500.000000	377.000000	334.000000	5500.000000
75%	500.000000	486.000000	468.000000	14600.000000

max	600.000000	500.000000	500.000000	80000.000000
	t_money	ct_helmets	...	ct_grenade_flashbang \
count	122410.000000	122410.000000	...	122410.000000
mean	11241.036680	2.053901	...	1.853157
std	12162.806759	1.841470	...	1.772791
min	0.000000	0.000000	...	0.000000
25%	1550.000000	0.000000	...	0.000000
50%	7150.000000	2.000000	...	1.000000
75%	18000.000000	4.000000	...	3.000000
max	80000.000000	5.000000	...	7.000000

	t_grenade_flashbang	ct_grenade_smokegrenade
t_grenade_smokegrenade \		
count	122410.000000	122410.000000
122410.000000		
mean	1.858100	1.540814
1.627146		
std	1.794473	1.737804
1.829147		
min	0.000000	0.000000
0.000000		
25%	0.000000	0.000000
0.000000		
50%	1.000000	1.000000
1.000000		
75%	3.000000	3.000000
3.000000		
max	7.000000	6.000000
9.000000		

	ct_grenade_incendiarygrenade	t_grenade_incendiarygrenade \
count	122410.000000	122410.000000
mean	1.001969	0.019819
std	1.458084	0.143933
min	0.000000	0.000000
25%	0.000000	0.000000
50%	0.000000	0.000000
75%	2.000000	0.000000
max	5.000000	3.000000

	ct_grenade_molotovgrenade	t_grenade_molotovgrenade \
count	122410.000000	122410.000000
mean	0.048011	1.352095
std	0.227669	1.663246
min	0.000000	0.000000
25%	0.000000	0.000000
50%	0.000000	1.000000
75%	0.000000	2.000000
max	3.000000	5.000000

	ct_grenade_decoygrenade	t_grenade_decoygrenade
count	122410.000000	122410.000000
mean	0.027694	0.025750
std	0.169531	0.164162
min	0.000000	0.000000
25%	0.000000	0.000000
50%	0.000000	0.000000
75%	0.000000	0.000000
max	3.000000	2.000000

[8 rows x 94 columns]

```
df.isnull().sum()
```

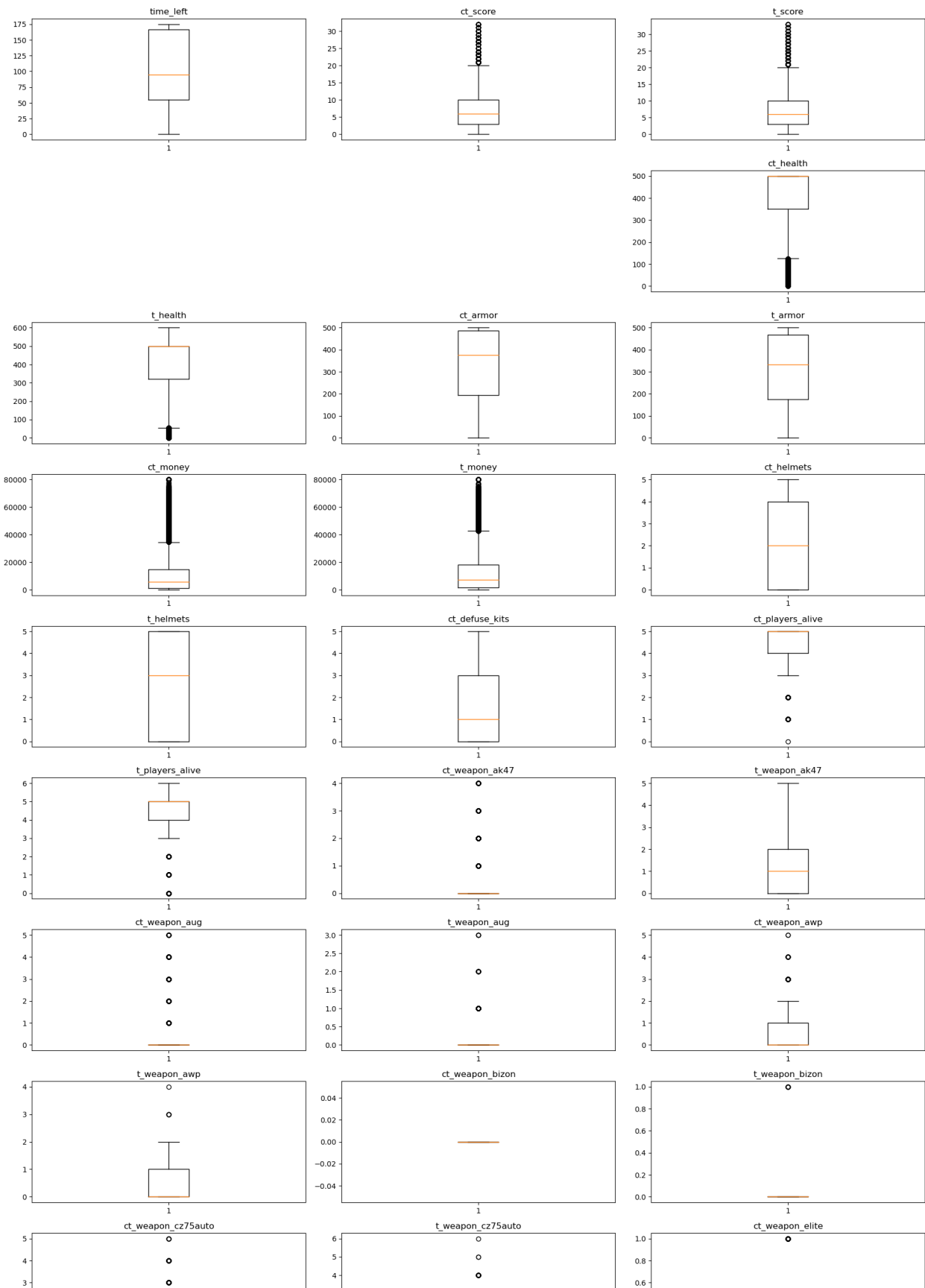
```
time_left      0
ct_score       0
t_score        0
map            0
bomb_planted   0
..
ct_grenade_molotovgrenade  0
t_grenade_molotovgrenade  0
ct_grenade_decoygrenade    0
t_grenade_decoygrenade    0
round_winner               0
Length: 97, dtype: int64
```

```
df.duplicated()
```

```
0      False
1      False
2      False
3      False
4      False
...
122405  False
122406  False
122407  False
122408  False
122409  False
Length: 122410, dtype: bool
```

```
plt.subplots(figsize=(18,140) )
length=len(df.columns)
for i,j in zip(range(length),df.columns):
    if(df[j].dtypes!="object" and df[j].dtypes!="bool") :
        plt.subplot(length//2,3,i+1)
        plt.boxplot(df[j])
        plt.title(j)
plt.tight_layout()
```

```
plt.show()
```




```
df.describe(include='0')
```

	map	round_winner
count	122410	122410
unique	8	2
top	de_inferno	T
freq	23811	62406

```
df["map"].value_counts()
```

de_inferno	23811
de_dust2	22144
de_nuke	19025
de_mirage	18576
de_overpass	14081
de_train	13491
de_vertigo	11137
de_cache	145

Name: map, dtype: int64

```
# lets see teams how they are successful in winning rounds
```

```
counts = df['map'].value_counts()
```

```
total = counts.sum()
```

```
percentages = counts / total * 100
```

```
counts.index
```

```
Index(['de_inferno', 'de_dust2', 'de_nuke', 'de_mirage',  
      'de_overpass',  
      'de_train', 'de_vertigo', 'de_cache'],  
      dtype='object')
```

```
for map_name, count, percent in zip(counts.index, counts.values,  
    percentages.values):
```

```
    print(f'{map_name}: {percent:.2f}%','/',count)
```

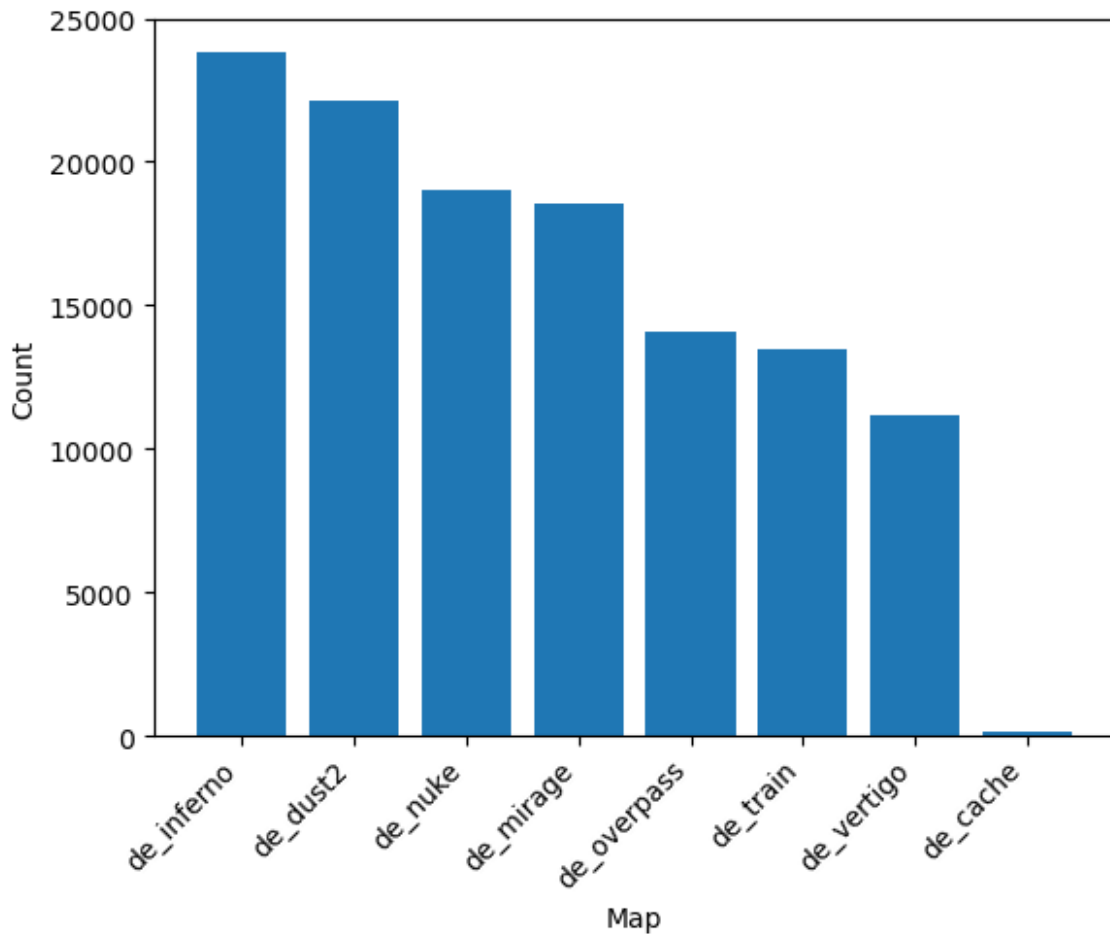
```
de_inferno: 19.45% / 23811  
de_dust2: 18.09% / 22144  
de_nuke: 15.54% / 19025  
de_mirage: 15.18% / 18576  
de_overpass: 11.50% / 14081  
de_train: 11.02% / 13491  
de_vertigo: 9.10% / 11137  
de_cache: 0.12% / 145
```

```
plt.bar(counts.index, counts.values)
```

```
plt.xticks(rotation=45, ha='right')  
plt.xlabel('Map')
```

```
plt.ylabel('Count')
```

```
Text(0, 0.5, 'Count')
```



```
for i in df.columns:
    if (df[i].dtypes=="object") | (df[i].dtypes=="bool"):
        print("Columns which have categorical values",i)
```

```
Columns which have categorical values map
Columns which have categorical values bomb_planted
Columns which have categorical values round_winner
```

```
df["bomb_planted"].value_counts()
```

```
False    108726
True       13684
Name: bomb_planted, dtype: int64
```

```
df["round_winner"].value_counts()
```

```
T         62406
CT        60004
Name: round_winner, dtype: int64
```

```
df["map"].value_counts()

de_inferno      23811
de_dust2        22144
de_nuke         19025
de_mirage       18576
de_overpass     14081
de_train        13491
de_vertigo      11137
de_cache        145
Name: map, dtype: int64

# Converting categorical features into a integer column
from sklearn.preprocessing import LabelEncoder
le=LabelEncoder()
df["bomb_planted"]=le.fit_transform(df["bomb_planted"])
df["map"]=le.fit_transform(df["map"])
df["round_winner"]=le.fit_transform(df["round_winner"])
```

Segregation of the data into dependent and independent columns

```
X=df.drop(columns=["round_winner"])
y=df[["round_winner"]]
```

Splitting the dataset into Train and Test datasets

```
from sklearn.model_selection import train_test_split
x_train,x_test,y_train,y_test=train_test_split(X,y,test_size=0.2,random_state=42)

# Scaling the data
from sklearn.preprocessing import StandardScaler
sc=StandardScaler()
x_train=sc.fit_transform(x_train)
x_test=sc.transform(x_test)
```

Choosing the model

```
from sklearn.discriminant_analysis import LinearDiscriminantAnalysis
lda=LinearDiscriminantAnalysis()
```

Training the model

```
lda.fit(x_train,y_train)

C:\Users\Arigala.Adarsh\anaconda3\lib\site-packages\sklearn\utils\
validation.py:1183: DataConversionWarning: 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().
  y = column_or_1d(y, warn=True)

LinearDiscriminantAnalysis()

lda.transform(x_test)

array([[ -0.31066701],
       [  0.16160545],
       [-2.19522227],
       ...,
       [  2.91456775],
       [-1.65102466],
       [  0.99528168]])

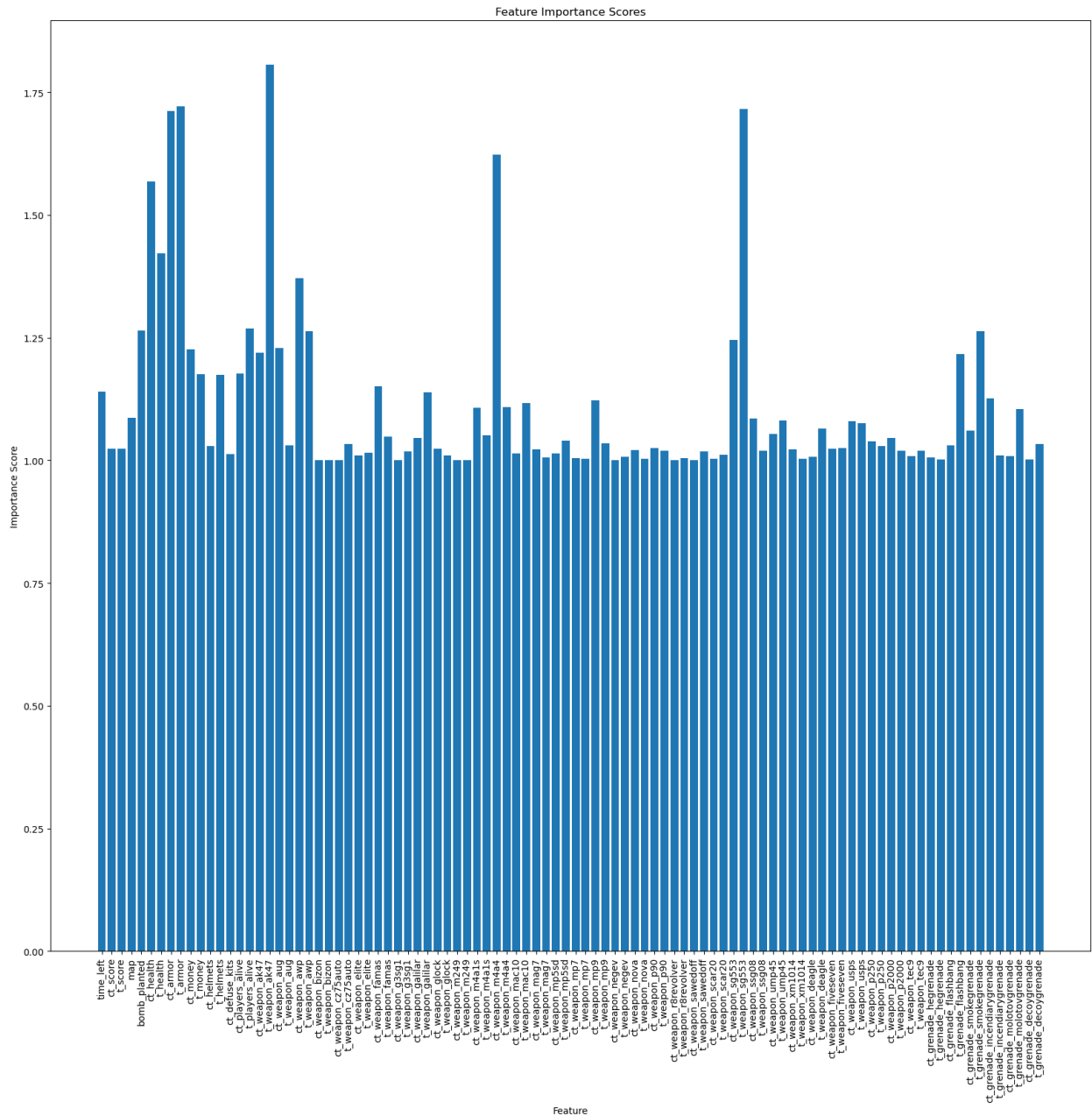
# Obtaining the LDA coefficients.This will give the importance scores
associated with each feature.
lda_coefficients=np.exp(np.abs(lda.coef_))
lda_coefficients= lda_coefficients.flatten()
lda_coefficients

array([1.14051375, 1.02348232, 1.02361012, 1.08711166, 1.26443741,
       1.56896428, 1.42226149, 1.71174795, 1.72149378, 1.22695841,
       1.17629264, 1.02925519, 1.17393359, 1.01233178, 1.17701202,
       1.26888831, 1.21975251, 1.80580698, 1.22872837, 1.03124048,
       1.37208174, 1.26351299, 1.          , 1.00130014, 1.00099319,
       1.03406233, 1.01079854, 1.01589366, 1.1511331 , 1.0481572 ,
       1.          , 1.01891369, 1.04559279, 1.13912974, 1.02386319,
       1.01009894, 1.          , 1.          , 1.10758456, 1.051198 ,
       1.62280372, 1.10928011, 1.01408145, 1.11733322, 1.02306813,
       1.00594113, 1.01383628, 1.04016588, 1.00546389, 1.00296978,
       1.1231047 , 1.0350133 , 1.          , 1.00757676, 1.02090408,
       1.00405543, 1.0252188 , 1.01962772, 1.          , 1.00493109,
       1.          , 1.01827676, 1.00401758, 1.01179667, 1.24546089,
       1.71649302, 1.08595734, 1.01928052, 1.05378886, 1.0817296 ,
       1.0223457 , 1.00392739, 1.00787732, 1.06560713, 1.02458755,
       1.02585421, 1.07987292, 1.07637588, 1.03850144, 1.02942838,
       1.0461802 , 1.02050308, 1.0093683 , 1.02047298, 1.00675662,
       1.0023777 , 1.03143135, 1.21721335, 1.06075101, 1.26371818,
       1.12626069, 1.01006317, 1.00852246, 1.10424102, 1.00231651,
       1.03293193])
```

```
num_features=X.shape[1]
feature_indices=np.arange(num_features)
feature_indices

feature_names=list(X.columns)

plt.figure(figsize=(20,18))
plt.bar(feature_indices,lda_coefficients)
plt.xticks(feature_indices,feature_names,rotation="vertical")
plt.xlabel('Feature')
plt.ylabel('Importance Score')
plt.title('Feature Importance Scores')
plt.show()
```



```
df_feature_score=pd.DataFrame({"Feature_names":feature_names,"feature_scores":lda_coefficients})
```

```
top_20_values=df_feature_score.nlargest(20,'feature_scores')
top_20_values.head(20)
```

	Feature_names	feature_scores
17	t_weapon_ak47	1.805807
8	t_armor	1.721494
65	t_weapon_sg553	1.716493
7	ct_armor	1.711748
40	ct_weapon_m4a4	1.622804

5	ct_health	1.568964
6	t_health	1.422261
20	ct_weapon_awp	1.372082
15	t_players_alive	1.268888
4	bomb_planted	1.264437
89	t_grenade_smokegrenade	1.263718
21	t_weapon_awp	1.263513
64	ct_weapon_sg553	1.245461
18	ct_weapon_aug	1.228728
9	ct_money	1.226958
16	ct_weapon_ak47	1.219753
87	t_grenade_flashbang	1.217213
14	ct_players_alive	1.177012
10	t_money	1.176293
12	t_helmets	1.173934

top_20_values.index

```
Int64Index([17, 8, 65, 7, 40, 5, 6, 20, 15, 4, 89, 21, 64, 18, 9, 16,
87, 14,
          10, 12],
          dtype='int64')
```

```
x_train=x_train[:,[17, 8, 65, 7, 40, 5, 6, 20, 15, 4, 89, 21, 64, 18,
9, 16, 87, 14,
                  10, 12]]
```

```
x_test=x_test[:,[17, 8, 65, 7, 40, 5, 6, 20, 15, 4, 89, 21, 64, 18, 9,
16, 87, 14,
                 10, 12]]
```

Logistic Regression

```
from sklearn.linear_model import LogisticRegression
lr=LogisticRegression()
lr.fit(x_train,y_train)
y_pred=lr.predict(x_test)
```

```
C:\Users\Arigala.Adarsh\anaconda3\lib\site-packages\sklearn\utils\
validation.py:1183: DataConversionWarning: 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().
  y = column_or_1d(y, warn=True)
```

Evolution of Model

```
from sklearn.metrics import accuracy_score,classification_report
accuracy_score(y_test,y_pred)
```

```
0.7516951229474717
```

```
classification_report(y_test,y_pred)
```

```
'          precision    recall  f1-score   support\n\n 0.74          0.76          0.75          12004\n 0.75          12478\n\n accuracy          0.75\n24482\n\n macro avg          0.75          0.75          24482\n\n weighted avg          0.75          0.75          24482\n'
```

```
# Decision Tree
```

```
from sklearn.tree import DecisionTreeClassifier
```

```
dtc=DecisionTreeClassifier()
```

```
dtc.fit(x_train,y_train)
```

```
y_pred=dtc.predict(x_test)
```

```
accuracy_score(y_test,y_pred)
```

```
0.8145984805162977
```

```
print(classification_report(y_test,y_pred))
```

	precision	recall	f1-score	support
0	0.81	0.81	0.81	12004
1	0.82	0.82	0.82	12478
accuracy			0.81	24482
macro avg	0.81	0.81	0.81	24482
weighted avg	0.81	0.81	0.81	24482

```
# Random Forest
```

```
from sklearn.ensemble import RandomForestClassifier
```

```
rfc=RandomForestClassifier()
```

```
rfc.fit(x_train,y_train)
```

```
y_pred=rfc.predict(x_test)
```

```
C:\Users\Arigala.Adarsh\anaconda3\lib\site-packages\sklearn\
base.py:1152: DataConversionWarning: 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().
```

```
    return fit_method(estimator, *args, **kwargs)
```

```
accuracy_score(y_test,y_pred)
```

```
0.8559758189690385
```

```
print(classification_report(y_test,y_pred))
```

	precision	recall	f1-score	support
0	0.85	0.86	0.85	12004
1	0.86	0.86	0.86	12478

accuracy			0.86	24482
macro avg	0.86	0.86	0.86	24482
weighted avg	0.86	0.86	0.86	24482