

```
In [1]: # Load the Data
import numpy as np
import pandas as pd
```

```
In [2]: from sklearn.tree import DecisionTreeClassifier
from sklearn.model_selection import train_test_split
import matplotlib.pyplot as plt
from sklearn.tree import plot_tree
```

```
In [3]: from sklearn.metrics import confusion_matrix
from sklearn.metrics import precision_score
from sklearn.metrics import recall_score
from sklearn.metrics import accuracy_score
```

```
In [4]: Team1vsTeam2 = pd.read_excel(r'Team1vsTeam2_2019-2023.xlsx')
df = Team1vsTeam2
df
```

```
Out[4]:
```

| | index | WR | KD | CKPM | GPR | GSPD | EGR | MLR | FB | FT | ... |
|-----|---------------------|-----------|-------|-------|-------|--------|------|-------|-------|-------|-----|
| 0 | T1 | 0.119447 | 0.13 | -0.02 | 0.37 | 0.014 | 1.1 | 10.9 | -0.01 | 0.05 | ... |
| 1 | Cloud9 | 0.196685 | 0.35 | -0.18 | 0.33 | 0.045 | 4.0 | 15.2 | 0.00 | -0.07 | ... |
| 2 | Gen.G | 0.030134 | 0.25 | -0.23 | 0.27 | 0.025 | 7.0 | -4.0 | 0.15 | 0.12 | ... |
| 3 | Team BDS | -0.236601 | -0.55 | -0.05 | -1.07 | -0.089 | -9.7 | -16.2 | 0.01 | -0.07 | ... |
| 4 | G2 Esports | 0.099203 | 0.01 | 0.35 | 0.27 | 0.030 | -0.4 | 10.4 | -0.10 | 0.08 | ... |
| ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... |
| 363 | J Team | -0.109790 | -0.30 | -0.20 | -0.47 | -0.027 | -5.8 | -5.7 | -0.12 | -0.09 | ... |
| 364 | Dplus KIA | -0.070833 | -0.09 | -0.02 | -0.10 | 0.019 | -7.0 | -0.1 | -0.15 | -0.03 | ... |
| 365 | Invictus Gaming | 0.103757 | -0.03 | 0.38 | 0.17 | 0.022 | 1.5 | 9.5 | -0.23 | 0.04 | ... |
| 366 | Royal Never Give Up | 0.177327 | 0.30 | 0.12 | 0.60 | 0.019 | 8.0 | 11.2 | 0.03 | -0.01 | ... |
| 367 | Fnatic | -0.037646 | -0.28 | 0.23 | -0.09 | -0.013 | -0.8 | -3.5 | 0.01 | -0.02 | ... |

368 rows × 21 columns

```
In [5]: df.IsWin.value_counts()
df.describe()
```

Out[5]:

| | WR | KD | CKPM | GPR | GSPD | EGR |
|--------------|------------|------------|------------|------------|------------|------------|
| count | 368.000000 | 368.000000 | 368.000000 | 368.000000 | 368.000000 | 368.000000 |
| mean | 0.005476 | 0.011848 | 0.002174 | 0.021630 | 0.001641 | 0.260598 |
| std | 0.112255 | 0.268764 | 0.158620 | 0.527456 | 0.039215 | 7.766588 |
| min | -0.293907 | -0.940000 | -0.430000 | -1.640000 | -0.125000 | -18.500000 |
| 25% | -0.078084 | -0.170000 | -0.110000 | -0.330000 | -0.024250 | -5.825000 |
| 50% | 0.007489 | 0.020000 | -0.010000 | 0.065000 | 0.002500 | 0.300000 |
| 75% | 0.089830 | 0.180000 | 0.110000 | 0.360000 | 0.027000 | 6.425000 |
| max | 0.279326 | 0.890000 | 0.430000 | 1.640000 | 0.125000 | 18.500000 |

In []:

Dummy Encoding

Encoding scheme to 'index' variable

```
In [6]: df_enc = pd.get_dummies(df, columns = ['index'])  
df_enc2 = pd.get_dummies(df, columns = ['index'], drop_first = True)
```

```
In [7]: df_enc.info()  
df_enc.head()
```

<class 'pandas.core.frame.DataFrame'>

RangeIndex: 368 entries, 0 to 367

Data columns (total 66 columns):

| # | Column | Non-Null Count | Dtype |
|----|---------------------------|----------------|---------|
| 0 | WR | 368 non-null | float64 |
| 1 | KD | 368 non-null | float64 |
| 2 | CKPM | 368 non-null | float64 |
| 3 | GPR | 368 non-null | float64 |
| 4 | GSPD | 368 non-null | float64 |
| 5 | EGR | 368 non-null | float64 |
| 6 | MLR | 368 non-null | float64 |
| 7 | FB | 368 non-null | float64 |
| 8 | FT | 368 non-null | float64 |
| 9 | F3T | 368 non-null | float64 |
| 10 | HLD | 368 non-null | float64 |
| 11 | FD | 368 non-null | float64 |
| 12 | DRG | 368 non-null | float64 |
| 13 | ELD | 368 non-null | float64 |
| 14 | BN | 368 non-null | float64 |
| 15 | LNE | 368 non-null | float64 |
| 16 | JNG | 368 non-null | float64 |
| 17 | WPM | 368 non-null | float64 |
| 18 | CWPM | 368 non-null | float64 |
| 19 | IsWin | 368 non-null | int64 |
| 20 | index_100 Thieves | 368 non-null | uint8 |
| 21 | index_Bilibili Gaming | 368 non-null | uint8 |
| 22 | index_CTBC Flying Oyster | 368 non-null | uint8 |
| 23 | index_Cloud9 | 368 non-null | uint8 |
| 24 | index_Clutch Gaming | 368 non-null | uint8 |
| 25 | index_DRX | 368 non-null | uint8 |
| 26 | index_DWG KIA | 368 non-null | uint8 |
| 27 | index_DetonatioN FocusMe | 368 non-null | uint8 |
| 28 | index_Dplus KIA | 368 non-null | uint8 |
| 29 | index_Dplus Kia | 368 non-null | uint8 |
| 30 | index_EDward Gaming | 368 non-null | uint8 |
| 31 | index_Evil Geniuses | 368 non-null | uint8 |
| 32 | index_FlyQuest | 368 non-null | uint8 |
| 33 | index_Fnatic | 368 non-null | uint8 |
| 34 | index_FunPlus Phoenix | 368 non-null | uint8 |
| 35 | index_G2 Esports | 368 non-null | uint8 |
| 36 | index_GAM Esports | 368 non-null | uint8 |
| 37 | index_GEN.G | 368 non-null | uint8 |
| 38 | index_Gen.G | 368 non-null | uint8 |
| 39 | index_Griffin | 368 non-null | uint8 |
| 40 | index_Hanwha Life Esports | 368 non-null | uint8 |
| 41 | index_Hong Kong Attitude | 368 non-null | uint8 |
| 42 | index_Invictus Gaming | 368 non-null | uint8 |
| 43 | index_J Team | 368 non-null | uint8 |
| 44 | index_JD Gaming | 368 non-null | uint8 |
| 45 | index_KT Rolster | 368 non-null | uint8 |
| 46 | index_LGD Gaming | 368 non-null | uint8 |
| 47 | index_LNG Esports | 368 non-null | uint8 |
| 48 | index_MAD Lions | 368 non-null | uint8 |
| 49 | index_Machi Esports | 368 non-null | uint8 |
| 50 | index_NRG | 368 non-null | uint8 |

```

51 index_PSG Talon          368 non-null  uint8
52 index_RNG                368 non-null  uint8
53 index_Rogue              368 non-null  uint8
54 index_Royal Never Give Up 368 non-null  uint8
55 index_SK Telecom T1      368 non-null  uint8
56 index_Splyce             368 non-null  uint8
57 index_Suning             368 non-null  uint8
58 index_T1                 368 non-null  uint8
59 index_TSM                368 non-null  uint8
60 index_Team BDS           368 non-null  uint8
61 index_Team Liquid        368 non-null  uint8
62 index_Top Esports        368 non-null  uint8
63 index_Unicorns of Love.CIS 368 non-null  uint8
64 index_Weibo Gaming       368 non-null  uint8
65 index_ahq eSports Club   368 non-null  uint8
dtypes: float64(19), int64(1), uint8(46)
memory usage: 74.2 KB

```

Out[7]:

| | WR | KD | CKPM | GPR | GSPD | EGR | MLR | FB | FT | F3T | ... | inde |
|----------|-----------|-------|-------|-------|--------|------|-------|-------|-------|-------|-----|------|
| 0 | 0.119447 | 0.13 | -0.02 | 0.37 | 0.014 | 1.1 | 10.9 | -0.01 | 0.05 | 0.03 | ... | |
| 1 | 0.196685 | 0.35 | -0.18 | 0.33 | 0.045 | 4.0 | 15.2 | 0.00 | -0.07 | -0.03 | ... | |
| 2 | 0.030134 | 0.25 | -0.23 | 0.27 | 0.025 | 7.0 | -4.0 | 0.15 | 0.12 | 0.06 | ... | |
| 3 | -0.236601 | -0.55 | -0.05 | -1.07 | -0.089 | -9.7 | -16.2 | 0.01 | -0.07 | -0.19 | ... | |
| 4 | 0.099203 | 0.01 | 0.35 | 0.27 | 0.030 | -0.4 | 10.4 | -0.10 | 0.08 | 0.03 | ... | |

5 rows × 66 columns

```

In [8]: df_enc2.info()
df_enc2.head()

```

<class 'pandas.core.frame.DataFrame'>

RangeIndex: 368 entries, 0 to 367

Data columns (total 65 columns):

| # | Column | Non-Null Count | Dtype |
|----|---------------------------|----------------|---------|
| 0 | WR | 368 non-null | float64 |
| 1 | KD | 368 non-null | float64 |
| 2 | CKPM | 368 non-null | float64 |
| 3 | GPR | 368 non-null | float64 |
| 4 | GSPD | 368 non-null | float64 |
| 5 | EGR | 368 non-null | float64 |
| 6 | MLR | 368 non-null | float64 |
| 7 | FB | 368 non-null | float64 |
| 8 | FT | 368 non-null | float64 |
| 9 | F3T | 368 non-null | float64 |
| 10 | HLD | 368 non-null | float64 |
| 11 | FD | 368 non-null | float64 |
| 12 | DRG | 368 non-null | float64 |
| 13 | ELD | 368 non-null | float64 |
| 14 | BN | 368 non-null | float64 |
| 15 | LNE | 368 non-null | float64 |
| 16 | JNG | 368 non-null | float64 |
| 17 | WPM | 368 non-null | float64 |
| 18 | CWPM | 368 non-null | float64 |
| 19 | IsWin | 368 non-null | int64 |
| 20 | index_Bilibili Gaming | 368 non-null | uint8 |
| 21 | index_CTBC Flying Oyster | 368 non-null | uint8 |
| 22 | index_Cloud9 | 368 non-null | uint8 |
| 23 | index_Clutch Gaming | 368 non-null | uint8 |
| 24 | index_DRX | 368 non-null | uint8 |
| 25 | index_DWG KIA | 368 non-null | uint8 |
| 26 | index_DetonatioN FocusMe | 368 non-null | uint8 |
| 27 | index_Dplus KIA | 368 non-null | uint8 |
| 28 | index_Dplus Kia | 368 non-null | uint8 |
| 29 | index_EDward Gaming | 368 non-null | uint8 |
| 30 | index_Evil Geniuses | 368 non-null | uint8 |
| 31 | index_FlyQuest | 368 non-null | uint8 |
| 32 | index_Fnatic | 368 non-null | uint8 |
| 33 | index_FunPlus Phoenix | 368 non-null | uint8 |
| 34 | index_G2 Esports | 368 non-null | uint8 |
| 35 | index_GAM Esports | 368 non-null | uint8 |
| 36 | index_GEN.G | 368 non-null | uint8 |
| 37 | index_Gen.G | 368 non-null | uint8 |
| 38 | index_Griffin | 368 non-null | uint8 |
| 39 | index_Hanwha Life Esports | 368 non-null | uint8 |
| 40 | index_Hong Kong Attitude | 368 non-null | uint8 |
| 41 | index_Invictus Gaming | 368 non-null | uint8 |
| 42 | index_J Team | 368 non-null | uint8 |
| 43 | index_JD Gaming | 368 non-null | uint8 |
| 44 | index_KT Rolster | 368 non-null | uint8 |
| 45 | index_LGD Gaming | 368 non-null | uint8 |
| 46 | index_LNG Esports | 368 non-null | uint8 |
| 47 | index_MAD Lions | 368 non-null | uint8 |
| 48 | index_Machi Esports | 368 non-null | uint8 |
| 49 | index_NRG | 368 non-null | uint8 |
| 50 | index_PSG Talon | 368 non-null | uint8 |

```

51 index_RNG          368 non-null   uint8
52 index_Rogue        368 non-null   uint8
53 index_Royal Never Give Up  368 non-null   uint8
54 index_SK Telecom T1 368 non-null   uint8
55 index_Splyce        368 non-null   uint8
56 index_Suning        368 non-null   uint8
57 index_T1           368 non-null   uint8
58 index_TSM           368 non-null   uint8
59 index_Team BDS      368 non-null   uint8
60 index_Team Liquid   368 non-null   uint8
61 index_Top Esports   368 non-null   uint8
62 index_Unicorns of Love.CIS 368 non-null   uint8
63 index_Weibo Gaming  368 non-null   uint8
64 index_ahq eSports Club 368 non-null   uint8

```

dtypes: float64(19), int64(1), uint8(45)

memory usage: 73.8 KB

Out[8]:

| | WR | KD | CKPM | GPR | GSPD | EGR | MLR | FB | FT | F3T | ... | inde |
|----------|-----------|-------|-------|-------|--------|------|-------|-------|-------|-------|-----|------|
| 0 | 0.119447 | 0.13 | -0.02 | 0.37 | 0.014 | 1.1 | 10.9 | -0.01 | 0.05 | 0.03 | ... | |
| 1 | 0.196685 | 0.35 | -0.18 | 0.33 | 0.045 | 4.0 | 15.2 | 0.00 | -0.07 | -0.03 | ... | |
| 2 | 0.030134 | 0.25 | -0.23 | 0.27 | 0.025 | 7.0 | -4.0 | 0.15 | 0.12 | 0.06 | ... | |
| 3 | -0.236601 | -0.55 | -0.05 | -1.07 | -0.089 | -9.7 | -16.2 | 0.01 | -0.07 | -0.19 | ... | |
| 4 | 0.099203 | 0.01 | 0.35 | 0.27 | 0.030 | -0.4 | 10.4 | -0.10 | 0.08 | 0.03 | ... | |

5 rows × 65 columns

```

In [9]: X = df_enc2[['WR', 'KD', 'CKPM', 'GPR', 'GSPD', 'EGR', 'MLR', 'FB', 'FT', 'F3T', 'HL
Y = df_enc2[['IsWin']]

```

```

In [10]: model = DecisionTreeClassifier()
model.fit(X,Y)

feature_importance = model.feature_importances_

feature_importance_df_enc2 = pd.DataFrame({'Feature': X.columns, 'Importance
feature_importance_df_enc2 = feature_importance_df_enc2.sort_values(by='Impo
feature_importance_df_enc2.head(25)

```

Out[10]:

| | Feature | Importance |
|-----------|---------|------------|
| 17 | WPM | 0.201932 |
| 13 | ELD | 0.094971 |
| 18 | CWPM | 0.080782 |
| 2 | CKPM | 0.080069 |
| 7 | FB | 0.076451 |
| 1 | KD | 0.065964 |
| 0 | WR | 0.065082 |
| 11 | FD | 0.052038 |
| 16 | JNG | 0.050143 |
| 4 | GSPD | 0.050075 |
| 10 | HLD | 0.047507 |
| 6 | MLR | 0.035129 |
| 14 | BN | 0.031032 |
| 8 | FT | 0.025621 |
| 5 | EGR | 0.019369 |
| 12 | DRG | 0.013958 |
| 15 | LNE | 0.004968 |
| 9 | F3T | 0.004907 |
| 3 | GPR | 0.000000 |

```
In [24]: features_to_remove = ['F3T', 'LNE', 'GPR']
df1=df.drop(features_to_remove, axis=1)
df1
```

Out[24]:

| | index | WR | KD | CKPM | GSPD | EGR | MLR | FB | FT | HLD | F |
|-----|---------------------|-----------|-------|-------|--------|------|-------|-------|-------|-------|------|
| 0 | T1 | 0.119447 | 0.13 | -0.02 | 0.014 | 1.1 | 10.9 | -0.01 | 0.05 | 0.02 | 0.0 |
| 1 | Cloud9 | 0.196685 | 0.35 | -0.18 | 0.045 | 4.0 | 15.2 | 0.00 | -0.07 | 0.12 | 0.1 |
| 2 | Gen.G | 0.030134 | 0.25 | -0.23 | 0.025 | 7.0 | -4.0 | 0.15 | 0.12 | -0.05 | -0.0 |
| 3 | Team BDS | -0.236601 | -0.55 | -0.05 | -0.089 | -9.7 | -16.2 | 0.01 | -0.07 | -0.03 | 0.0 |
| 4 | G2 Esports | 0.099203 | 0.01 | 0.35 | 0.030 | -0.4 | 10.4 | -0.10 | 0.08 | 0.10 | -0.0 |
| ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... |
| 363 | J Team | -0.109790 | -0.30 | -0.20 | -0.027 | -5.8 | -5.7 | -0.12 | -0.09 | -0.05 | -0.1 |
| 364 | Dplus KIA | -0.070833 | -0.09 | -0.02 | 0.019 | -7.0 | -0.1 | -0.15 | -0.03 | 0.04 | -0.1 |
| 365 | Invictus Gaming | 0.103757 | -0.03 | 0.38 | 0.022 | 1.5 | 9.5 | -0.23 | 0.04 | -0.10 | -0.1 |
| 366 | Royal Never Give Up | 0.177327 | 0.30 | 0.12 | 0.019 | 8.0 | 11.2 | 0.03 | -0.01 | -0.10 | 0.1 |
| 367 | Fnatic | -0.037646 | -0.28 | 0.23 | -0.013 | -0.8 | -3.5 | 0.01 | -0.02 | -0.15 | -0.0 |

368 rows × 18 columns

```
In [25]: X = df1[['WR', 'KD', 'CKPM', 'GSPD', 'EGR', 'MLR', 'FT', 'FB', 'HLD', 'FD', 'DRG', 'ELD']]
Y = df1[['IsWin']]
```

```
In [26]: # Split Data

X_train, X_test, Y_train, Y_test = train_test_split(X, Y, test_size=0.3, random_state=42)
print(X_train.shape, X_test.shape)
```

Out[26]: ((257, 16), (111, 16))

```
In [27]: # BASELINE
negative = np.sum(Y_train == 0)
positive = np.sum(Y_train == 1)
print(pd.Series({'0': negative, '1': positive}))

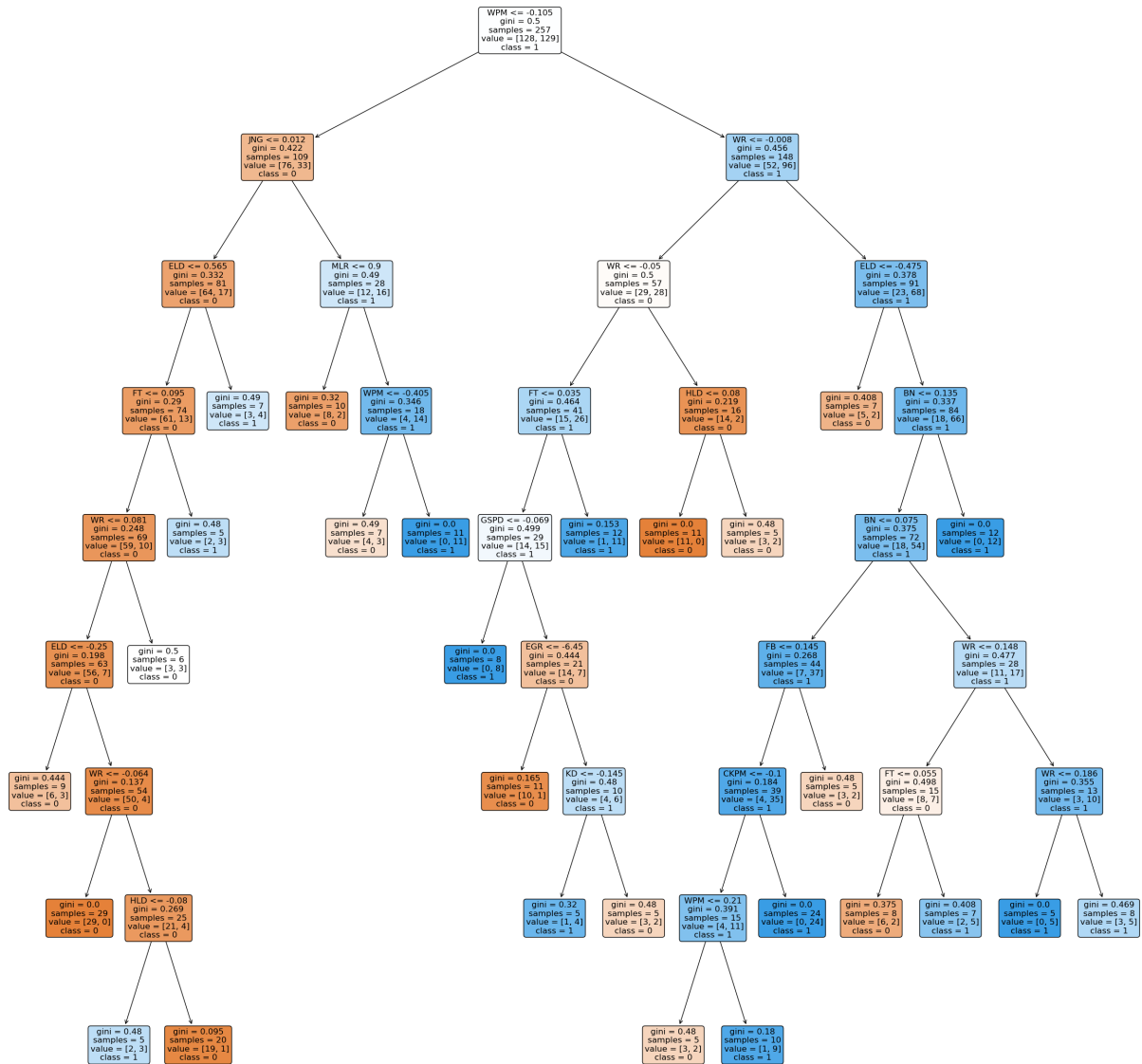
0    IsWin    128
dtype: int64
1    IsWin    129
dtype: int64
dtype: object
```

```
In [28]: dtc = DecisionTreeClassifier(min_samples_leaf=5, ccp_alpha=0.003, random_state=42)
dtc = dtc.fit(X_train, Y_train)
```



```
In [29]: print('Node count =', dtc.tree_.node_count)
plt.figure(figsize=(30,30))
plot_tree(dtc,
          feature_names=list(X_train.columns),
          class_names=['0', '1'],
          filled=True,
          impurity=True,
          rounded=True,
          fontsize=12)
plt.show()
```

Node count = 53



```
In [30]: # MAKE PREDICTIONS
Y_pred = dtc.predict(X_test)
cm = confusion_matrix(Y_test, Y_pred)
Y_proba = dtc.predict_proba(X_test)
print ("Confusion Matrix : \n", cm)
print('Precision:', precision_score(Y_test, Y_pred))
print('Recall:', recall_score(Y_test, Y_pred))
```

```
print('Accuracy:',accuracy_score(Y_test, Y_pred))  
display(Y_proba)
```

Confusion Matrix :

[[24 25]

[27 35]]

Precision: 0.5833333333333334

Recall: 0.5645161290322581

Accuracy: 0.5315315315315315

```
array([[0.          , 1.          ],
       [0.42857143, 0.57142857],
       [0.2        , 0.8         ],
       [1.         , 0.          ],
       [0.1        , 0.9         ],
       [0.90909091, 0.09090909],
       [0.         , 1.          ],
       [0.         , 1.          ],
       [0.         , 1.          ],
       [0.375      , 0.625      ],
       [0.6        , 0.4         ],
       [0.6        , 0.4         ],
       [0.         , 1.          ],
       [0.         , 1.          ],
       [0.95       , 0.05       ],
       [0.         , 1.          ],
       [0.75       , 0.25       ],
       [0.         , 1.          ],
       [0.6        , 0.4         ],
       [0.2        , 0.8         ],
       [0.         , 1.          ],
       [0.         , 1.          ],
       [0.1        , 0.9         ],
       [0.         , 1.          ],
       [0.         , 1.          ],
       [0.28571429, 0.71428571],
       [0.2        , 0.8         ],
       [0.75       , 0.25       ],
       [1.         , 0.          ],
       [0.90909091, 0.09090909],
       [0.08333333, 0.91666667],
       [1.         , 0.          ],
       [0.6        , 0.4         ],
       [0.4        , 0.6         ],
       [0.95       , 0.05       ],
       [1.         , 0.          ],
       [1.         , 0.          ],
       [0.71428571, 0.28571429],
       [0.         , 1.          ],
       [0.2        , 0.8         ],
       [0.28571429, 0.71428571],
       [0.6        , 0.4         ],
       [0.375      , 0.625      ],
       [0.90909091, 0.09090909],
       [0.75       , 0.25       ],
       [0.         , 1.          ],
       [0.2        , 0.8         ],
       [0.28571429, 0.71428571],
       [0.8        , 0.2         ],
       [0.2        , 0.8         ],
       [1.         , 0.          ],
       [0.95       , 0.05       ],
       [0.08333333, 0.91666667],
       [1.         , 0.          ],
       [0.1        , 0.9         ],
       [0.         , 1.          ],
```

```
[0.5      , 0.5      ],
[0.42857143, 0.57142857],
[0.6      , 0.4      ],
[0.       , 1.       ],
[0.2      , 0.8      ],
[0.4      , 0.6      ],
[0.       , 1.       ],
[1.       , 0.       ],
[0.       , 1.       ],
[0.90909091, 0.09090909],
[0.       , 1.       ],
[0.90909091, 0.09090909],
[0.57142857, 0.42857143],
[0.       , 1.       ],
[0.       , 1.       ],
[0.95     , 0.05     ],
[1.       , 0.       ],
[0.66666667, 0.33333333],
[0.       , 1.       ],
[0.1      , 0.9      ],
[0.71428571, 0.28571429],
[1.       , 0.       ],
[0.       , 1.       ],
[0.       , 1.       ],
[0.       , 1.       ],
[0.4      , 0.6      ],
[0.2      , 0.8      ],
[0.75     , 0.25     ],
[1.       , 0.       ],
[0.90909091, 0.09090909],
[0.       , 1.       ],
[0.8      , 0.2      ],
[1.       , 0.       ],
[0.2      , 0.8      ],
[0.75     , 0.25     ],
[1.       , 0.       ],
[0.       , 1.       ],
[0.6      , 0.4      ],
[0.75     , 0.25     ],
[0.       , 1.       ],
[0.08333333, 0.91666667],
[1.       , 0.       ],
[0.       , 1.       ],
[0.6      , 0.4      ],
[0.1      , 0.9      ],
[0.6      , 0.4      ],
[0.6      , 0.4      ],
[1.       , 0.       ],
[0.       , 1.       ],
[0.       , 1.       ],
[0.75     , 0.25     ],
[0.2      , 0.8      ],
[0.71428571, 0.28571429],
[0.66666667, 0.33333333],
[0.       , 1.       ]])
```

In []:

Predict the Winner

8-in-4

```
In [31]: year2023StatForWorldsTeam = pd.read_excel(r'year2023StatForWorldsTeam(1).xls')
year2023StatForWorldsTeam.set_index("team", inplace = True)
year2023StatForWorldsTeam.rename(columns={"win rate": "WR"}, inplace=True)

selected_columns = year2023StatForWorldsTeam[['WR', 'KD', 'CKPM', 'GSPD', 'EGR',
selected_columns
```

Out[31]:

| | WR | KD | CKPM | GSPD | EGR | MLR | FT | FB | HLD | FD | DRG |
|-----------------|----------|------|------|--------|------|-------|------|------|------|------|------|
| team | | | | | | | | | | | |
| Gen.G | 0.726562 | 1.53 | 0.75 | 0.075 | 60.8 | 11.9 | 0.66 | 0.61 | 0.53 | 0.44 | 0.59 |
| T1 | 0.611511 | 1.10 | 0.77 | 0.040 | 59.8 | 1.4 | 0.72 | 0.51 | 0.61 | 0.61 | 0.58 |
| KT Rolster | 0.694215 | 1.52 | 0.69 | 0.045 | 57.5 | 11.9 | 0.61 | 0.55 | 0.45 | 0.55 | 0.60 |
| Dplus KIA | 0.606061 | 1.35 | 0.67 | 0.039 | 60.5 | 0.1 | 0.53 | 0.58 | 0.54 | 0.62 | 0.62 |
| JD Gaming | 0.777778 | 1.58 | 0.86 | 0.083 | 58.5 | 21.5 | 0.55 | 0.52 | 0.43 | 0.50 | 0.60 |
| Bilibili Gaming | 0.641379 | 1.21 | 0.91 | 0.053 | 58.7 | 6.9 | 0.52 | 0.57 | 0.48 | 0.59 | 0.56 |
| LNG Esports | 0.651786 | 1.39 | 0.81 | 0.045 | 61.4 | -1.4 | 0.40 | 0.61 | 0.60 | 0.60 | 0.54 |
| Weibo Gaming | 0.607843 | 1.18 | 0.82 | 0.044 | 52.3 | -18.9 | 0.67 | 0.45 | 0.50 | 0.67 | 0.51 |
| G2 Esports | 0.705263 | 1.36 | 1.02 | 0.069 | 60.1 | 10.5 | 0.61 | 0.48 | 0.64 | 0.60 | 0.59 |
| Fnatic | 0.492754 | 0.97 | 0.92 | -0.009 | 46.6 | 1.9 | 0.41 | 0.50 | 0.36 | 0.47 | 0.50 |
| MAD Lions | 0.483871 | 0.93 | 0.98 | -0.002 | 52.2 | -3.3 | 0.53 | 0.49 | 0.44 | 0.57 | 0.53 |
| Team BDS | 0.541176 | 1.03 | 0.81 | -0.006 | 48.8 | 5.3 | 0.48 | 0.53 | 0.40 | 0.55 | 0.56 |
| NRG | 0.560976 | 1.14 | 0.84 | -0.007 | 43.1 | 13.0 | 0.44 | 0.44 | 0.49 | 0.51 | 0.51 |
| Cloud9 | 0.680556 | 1.28 | 0.80 | 0.043 | 56.2 | 11.9 | 0.46 | 0.49 | 0.56 | 0.67 | 0.59 |
| Team Liquid | 0.492063 | 0.97 | 0.79 | 0.026 | 58.7 | -9.5 | 0.67 | 0.52 | 0.59 | 0.56 | 0.55 |
| GAM Esports | 0.696429 | 1.28 | 0.98 | 0.050 | 53.8 | 15.9 | 0.54 | 0.46 | 0.58 | 0.49 | 0.59 |

```
In [32]: # GenG in Blue
GenG_vs_BLG = selected_columns.loc[["Gen.G"]].sub(selected_columns.loc[["Bil
# BLG in Blue
BLG_vs_GenG = selected_columns.loc[["Bilibili Gaming"]].sub(selected_columns
```

```
In [33]: # Predict the Winning Rate
GenG_WR = dtc.predict_proba(GenG_vs_BLG)
BLG_WR = dtc.predict_proba(BLG_vs_GenG)
print('The Winning Rate for Gen.G in the blue side is ', GenG_WR,
      '; the Winning Rate for BLG in the blue side is ', BLG_WR)
```

The Winning Rate for Gen.G in the blue side is [[0. 1.]] ; the Winning Rate for BLG in the blue side is [[0.2 0.8]]

GenG win

```
In [37]: # NRG in Blue
NRG_vs_WBG = selected_columns.loc[["NRG"]].sub(selected_columns.loc[["Weibo
# WBG in Blue
WBG_vs_NRG = selected_columns.loc[["Weibo Gaming"]].sub(selected_columns.loc

# Predict the Winning Rate
NRG_WR = dtc.predict_proba(NRG_vs_WBG)
WBG_WR = dtc.predict_proba(WBG_vs_NRG)

print('The Winning Rate for NRG in the blue side is ', NRG_WR, '; the Winnin
```

The Winning Rate for NRG in the blue side is [[1. 0.]] ; the Winning Rate for WBG in the blue side is [[0.8 0.2]]

WBG win

```
In [38]: # JDG in Blue
JDG_vs_KT = selected_columns.loc[["JD Gaming"]].sub(selected_columns.loc[["K
# KT in Blue
KT_vs_JDG = selected_columns.loc[["KT Rolster"]].sub(selected_columns.loc[["

# Predict the Winning Rate
JDG_WR = dtc.predict(JDG_vs_KT)
KT_WR = dtc.predict(KT_vs_JDG)

print('The Winning Rate for JDG in the blue side is ', JDG_WR, '; the Winnin
```

The Winning Rate for JDG in the blue side is [0] ; the Winning Rate for KT in the blue side is [1]

KT win

```
In [41]: # LNG in Blue
LNG_vs_T1 = selected_columns.loc[["LNG Esports"]].sub(selected_columns.loc[["
# T1 in Blue
T1_vs_LNG = selected_columns.loc[["T1"]].sub(selected_columns.loc[["LNG Espc

# Predict the Winning Rate
LNG_WR = dtc.predict_proba(LNG_vs_T1)
```

```
T1_WR = dtc.predict_proba(T1_vs_LNG)
print('The Winning Rate for LNG in the blue side is ', LNG_WR, '; the Winnin
```

The Winning Rate for LNG in the blue side is [[0.66666667 0.33333333]] ; the Winning Rate for T1 in the blue side is [[1. 0.]]

LNG win

In []:

Semifinals

```
In [43]: # GenG in Blue
GenG_vs_WBG = selected_columns.loc[["Gen.G"]].sub(selected_columns.loc[["Wei
# WBG in Blue
WBG_vs_GenG = selected_columns.loc[["Weibo Gaming"]].sub(selected_columns.loc
# Predict the Winning Rate
GenG_WR = dtc.predict_proba(GenG_vs_WBG)
WBG_WR = dtc.predict_proba(WBG_vs_GenG)
print('The Winning Rate for GenG in the blue side is ', GenG_WR, '; the Winn
```

The Winning Rate for GenG in the blue side is [[0.6 0.4]] ; the Winning Rate for WBG in the blue side is [[0.90909091 0.09090909]]

GenG win

```
In [44]: # KT in Blue
KT_vs_JDG = selected_columns.loc[["KT Rolster"]].sub(selected_columns.loc[["
# LNG in Blue
LNG_vs_T1 = selected_columns.loc[["LNG Esports"]].sub(selected_columns.loc[["
# Predict the Winning Rate
KT_WR = dtc.predict(KT_vs_JDG)
LNG_WR = dtc.predict(LNG_vs_T1)
print('The Winning Rate for KT in the blue side is ', KT_WR,
      '; the Winning Rate for LNG in the blue side is ', LNG_WR)
```

The Winning Rate for KT in the blue side is [1] ; the Winning Rate for LNG in the blue side is [0]

KT win

In []:

Final

```
In [47]: # GenG in Blue
GenG_vs_KT = selected_columns.loc[["Gen.G"]].sub(selected_columns.loc[["KT R
# KT in Blue
KT_vs_GenG = selected_columns.loc[["KT Rolster"]].sub(selected_columns.loc[["
# Predict the Winning Rate
GenG_WR = dtc.predict(GenG_vs_KT)
KT_WR = dtc.predict(KT_vs_GenG)
```

```
print('The Winning Rate for GenG in the blue side is ', GenG_WR,  
      '; the Winning Rate for KT in the blue side is ', KT_WR)
```

The Winning Rate for GenG in the blue side is [1] ; the Winning Rate for KT in the blue side is [0]

By CART model, the winner is GenG

In []:

In []: