

Taemin Huh DATA1030 Midterm Project

October 21, 2022

1 DATA1030 Project: Pokemon Battle Analysis

1.1 Exploratory Data Analysis

```
[1]: import pandas as pd
import numpy as np
from matplotlib import pylab as plt

pkmn = pd.read_csv(r"C:\Users\User\Desktop\DSI\DATA1030-Fall2022\Taemin Huh\
↳DATA1030 Project\data\pokemon.csv")
pkmn = pkmn.rename(index=str, columns={"#": "ID"})
battle = pd.read_csv(r"C:\Users\User\Desktop\DSI\DATA1030-Fall2022\Taemin Huh\
↳DATA1030 Project\data\combats.csv")
pkmn.head()
```

```
[1]:
```

	ID	Name	Type 1	Type 2	HP	Attack	Defense	Sp. Atk	Sp. Def	\
0	1	Bulbasaur	Grass	Poison	45	49	49	65	65	
1	2	Ivysaur	Grass	Poison	60	62	63	80	80	
2	3	Venusaur	Grass	Poison	80	82	83	100	100	
3	4	Mega Venusaur	Grass	Poison	80	100	123	122	120	
4	5	Charmander	Fire	NaN	39	52	43	60	50	

	Speed	Generation	Legendary
0	45	1	False
1	60	1	False
2	80	1	False
3	80	1	False
4	65	1	False

```
[2]: battle.head()
```

```
[2]:
```

	First_pokemon	Second_pokemon	Winner
0	266	298	298
1	702	701	701
2	191	668	668
3	237	683	683
4	151	231	151

```
[3]: print("Shape of Pokemon data: ", (pkmn.shape))
      print("Shape of Battle data: ", (battle.shape))
```

```
Shape of Pokemon data: (800, 12)
Shape of Battle data: (50000, 3)
```

```
[4]: # Identifying missing values
      pkmn.isnull().sum()
```

```
[4]: ID          0
      Name        1
      Type 1      0
      Type 2     386
      HP          0
      Attack      0
      Defense     0
      Sp. Atk     0
      Sp. Def     0
      Speed       0
      Generation  0
      Legendary   0
      dtype: int64
```

```
[5]: battle.isnull().sum()
```

```
[5]: First_pokemon    0
      Second_pokemon  0
      Winner          0
      dtype: int64
```

```
[6]: # Singling out Pokemone with missing name
      pkmn[pkmn['Name'].isnull()]
```

```
[6]:      ID Name      Type 1 Type 2  HP  Attack  Defense  Sp. Atk  Sp. Def  Speed  \
62  63  NaN  Fighting    NaN  65    105      60      60      70    95

      Generation  Legendary
62              1      False
```

```
[7]: # By looking at the online Pokemon database (https://www.serebii.net/pokemon/type/fighting/), able to match the missing Pokemon name to "Primeape", which
      ↪ has exactly the above stat profile among the pool of 7 pure Fighting-type (i.
      ↪ e. no Type 2) Pokemons in Generation 1.
      pkmn['Name'][62] = "Primeape"
```

```
C:\Users\User\AppData\Local\Temp\ipykernel_27768\2549842310.py:2:
SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame
```

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy

```
pkmn['Name'][62] = "Primeape"
```

```
[8]: # Feature engineering & merging datasets
winTotal = battle.Winner.value_counts()
winTotal = winTotal.sort_index()
firstAtk_Count = battle.First_pokemon.value_counts()
secondAtk_Count = battle.Second_pokemon.value_counts()
battleTotal = firstAtk_Count + secondAtk_Count
```

```
[9]: featEng = pd.DataFrame()
featEng['Total Win Count'] = winTotal
featEng['Total Battle Count'] = battleTotal
# featEng['First Attack Count'] = firstAtk_Count
# featEng['First Attack Rate'] = firstAtk_Count/battleTotal
featEng['Win Rate'] = winTotal/battleTotal
pkmnWR = pd.merge(pkmn, featEng, right_index = True, left_on='ID')
pkmnWR
```

```
[9]:
```

	ID	Name	Type 1	Type 2	HP	Attack	Defense	Sp. Atk	\
0	1	Bulbasaur	Grass	Poison	45	49	49	65	
1	2	Ivysaur	Grass	Poison	60	62	63	80	
2	3	Venusaur	Grass	Poison	80	82	83	100	
3	4	Mega Venusaur	Grass	Poison	80	100	123	122	
4	5	Charmander	Fire	NaN	39	52	43	60	
..	
795	796	Diancie	Rock	Fairy	50	100	150	100	
796	797	Mega Diancie	Rock	Fairy	50	160	110	160	
797	798	Hoopla Confined	Psychic	Ghost	80	110	60	150	
798	799	Hoopla Unbound	Psychic	Dark	80	160	60	170	
799	800	Volcanion	Fire	Water	80	110	120	130	

	Sp. Def	Speed	Generation	Legendary	Total Win Count	\
0	65	45	1	False	37	
1	80	60	1	False	46	
2	100	80	1	False	89	
3	120	80	1	False	70	
4	50	65	1	False	55	
..	
795	150	50	6	True	39	
796	110	110	6	True	116	
797	130	70	6	True	60	
798	130	80	6	True	89	
799	90	70	6	True	75	

Total Battle Count Win Rate

```

0          133  0.278195
1          121  0.380165
2          132  0.674242
3          125  0.560000
4          112  0.491071
..         ...      ...
795        105  0.371429
796        131  0.885496
797        119  0.504202
798        144  0.618056
799        121  0.619835

```

[783 rows x 15 columns]

```
[10]: pkmnAll = pd.merge(pkmn, featEng, left_on='ID', right_index = True, how='left')
pkmnAll[pkmnAll['Win Rate'].isnull()]
```

```
[10]:
```

	ID	Name	Type 1	Type 2	HP	Attack	Defense	\
11	12	Blastoise	Water	NaN	79	83	100	
32	33	Sandshrew	Ground	NaN	50	75	85	
45	46	Wigglytuff	Normal	Fairy	140	70	45	
65	66	Poliwag	Water	NaN	40	50	40	
77	78	Victreebel	Grass	Poison	80	105	65	
89	90	Magnetron	Electric	Steel	50	60	95	
143	144	Ditto	Normal	NaN	48	48	48	
182	183	Ariados	Bug	Poison	70	90	70	
230	231	Shuckle	Bug	Rock	20	10	230	
235	236	Ursaring	Normal	NaN	90	130	75	
321	322	Hariyama	Fighting	NaN	144	120	60	
418	419	Mega Latias	Dragon	Psychic	80	100	120	
478	479	Honchkrow	Dark	Flying	100	125	52	
555	556	Servine	Grass	NaN	60	60	75	
617	618	Maractus	Grass	NaN	75	86	67	
654	655	Jellicent	Water	Ghost	100	60	70	
781	782	Pumpkaboo Small Size	Ghost	Grass	44	66	70	

	Sp. Atk	Sp. Def	Speed	Generation	Legendary	Total Win Count	\
11	85	105	78	1	False	NaN	
32	20	30	40	1	False	NaN	
45	85	50	45	1	False	NaN	
65	40	40	90	1	False	NaN	
77	100	70	70	1	False	NaN	
89	120	70	70	1	False	NaN	
143	48	48	48	1	False	NaN	
182	60	60	40	2	False	NaN	
230	10	230	5	2	False	NaN	
235	75	75	55	2	False	NaN	

321	40	60	50	3	False	NaN
418	140	150	110	3	True	NaN
478	105	52	71	4	False	NaN
555	60	75	83	5	False	NaN
617	106	67	60	5	False	NaN
654	85	105	60	5	False	NaN
781	44	55	56	6	False	NaN

	Total Battle Count	Win Rate
11	NaN	NaN
32	NaN	NaN
45	NaN	NaN
65	NaN	NaN
77	NaN	NaN
89	NaN	NaN
143	NaN	NaN
182	NaN	NaN
230	NaN	NaN
235	NaN	NaN
321	NaN	NaN
418	NaN	NaN
478	NaN	NaN
555	NaN	NaN
617	NaN	NaN
654	NaN	NaN
781	NaN	NaN

```
[11]: pkmnWR.describe()
```

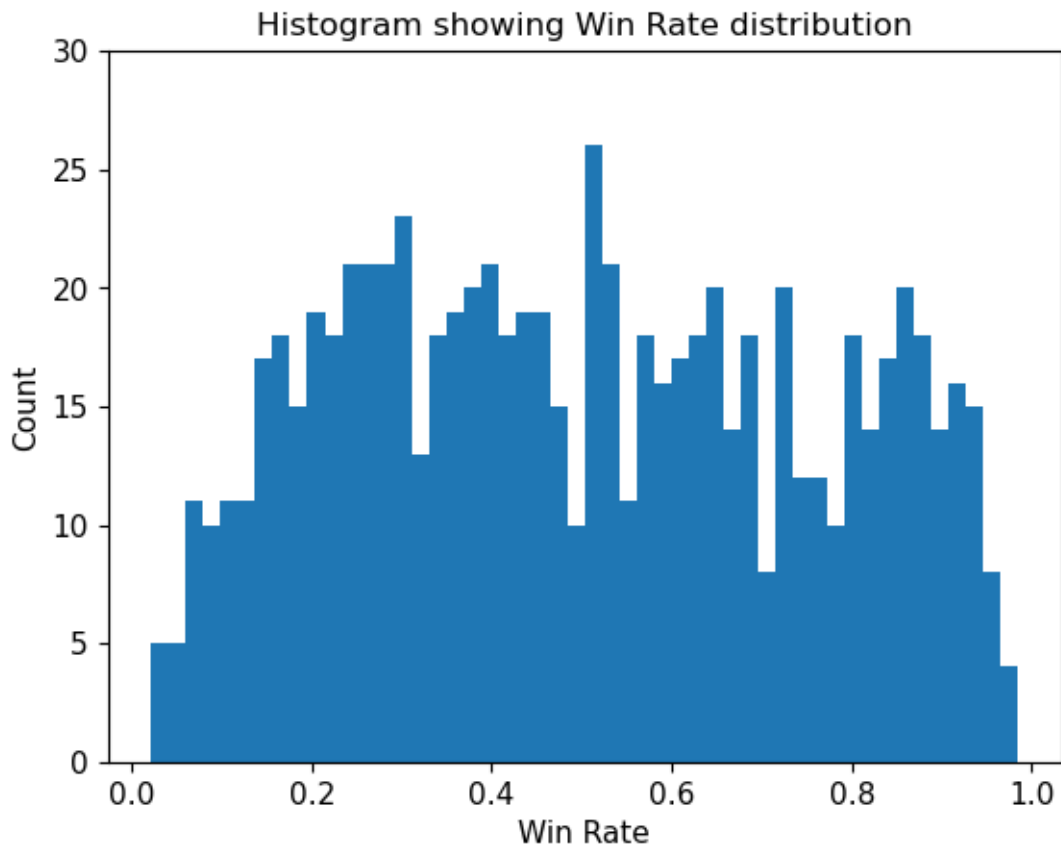
```
[11]:
```

	ID	HP	Attack	Defense	Sp. Atk	Sp. Def \
count	783.000000	783.000000	783.000000	783.000000	783.000000	783.000000
mean	402.873563	69.140485	79.007663	73.699872	72.839080	71.749681
std	230.297452	25.348783	32.502566	30.879737	32.672868	27.248993
min	1.000000	1.000000	5.000000	5.000000	10.000000	20.000000
25%	204.500000	50.000000	55.000000	50.000000	50.000000	50.000000
50%	403.000000	65.000000	75.000000	70.000000	65.000000	70.000000
75%	601.500000	80.000000	100.000000	90.000000	95.000000	90.000000
max	800.000000	255.000000	190.000000	230.000000	194.000000	200.000000

	Speed	Generation	Total Win Count	Total Battle Count	Win Rate
count	783.000000	783.000000	783.000000	783.000000	783.000000
mean	68.443167	3.339719	63.856960	127.541507	0.501538
std	29.158076	1.656435	32.925941	11.397402	0.254993
min	5.000000	1.000000	3.000000	92.000000	0.021739
25%	45.000000	2.000000	36.000000	120.000000	0.284228
50%	65.000000	3.000000	62.000000	128.000000	0.491071
75%	90.000000	5.000000	91.000000	135.000000	0.717644

max 180.000000 6.000000 152.000000 164.000000 0.984496

```
[12]: pkmnWR['Win Rate'].plot.hist(bins = 50,fontsize=11)
ax = plt.gca()
ax.set_ylim([0, 30])
plt.xlabel('Win Rate',fontsize=11)
plt.ylabel('Count',fontsize=11)
# arial = {'fontname':'Arial'}
# txt="Figure 1: Histogram showing Win Rate distribution."
# plt.figtext(0.5, -0.05, txt, wrap=True, horizontalalignment='center',
# ↪ fontsize=10)
plt.title('Histogram showing Win Rate distribution',fontsize=12)
plt.savefig(r"C:\Users\User\Desktop\DSI\DATA1030-Fall2022\Taemin Huh DATA1030_
↪ Project\figures\WR Histogram.png")
plt.show()
```



```
[13]: # Pearson correlation table & heatmap
col = ['HP', 'Attack', 'Defense', 'Sp. Atk', 'Sp. Def', 'Speed', 'Legendary',
↪ 'Win Rate']
```

```
pkmnWR.loc[:,col].corr()
```

```
[13]:
```

	HP	Attack	Defense	Sp. Atk	Sp. Def	Speed	\
HP	1.000000	0.417427	0.265230	0.363244	0.409110	0.179423	
Attack	0.417427	1.000000	0.464539	0.395211	0.288078	0.382310	
Defense	0.265230	0.464539	1.000000	0.237592	0.490118	0.025762	
Sp. Atk	0.363244	0.395211	0.237592	1.000000	0.529276	0.470548	
Sp. Def	0.409110	0.288078	0.490118	0.529276	1.000000	0.276715	
Speed	0.179423	0.382310	0.025762	0.470548	0.276715	1.000000	
Legendary	0.280265	0.348391	0.247921	0.448339	0.367030	0.323420	
Win Rate	0.258006	0.500181	0.129426	0.478940	0.324218	0.937742	

	Legendary	Win Rate
HP	0.280265	0.258006
Attack	0.348391	0.500181
Defense	0.247921	0.129426
Sp. Atk	0.448339	0.478940
Sp. Def	0.367030	0.324218
Speed	0.323420	0.937742
Legendary	1.000000	0.325007
Win Rate	0.325007	1.000000

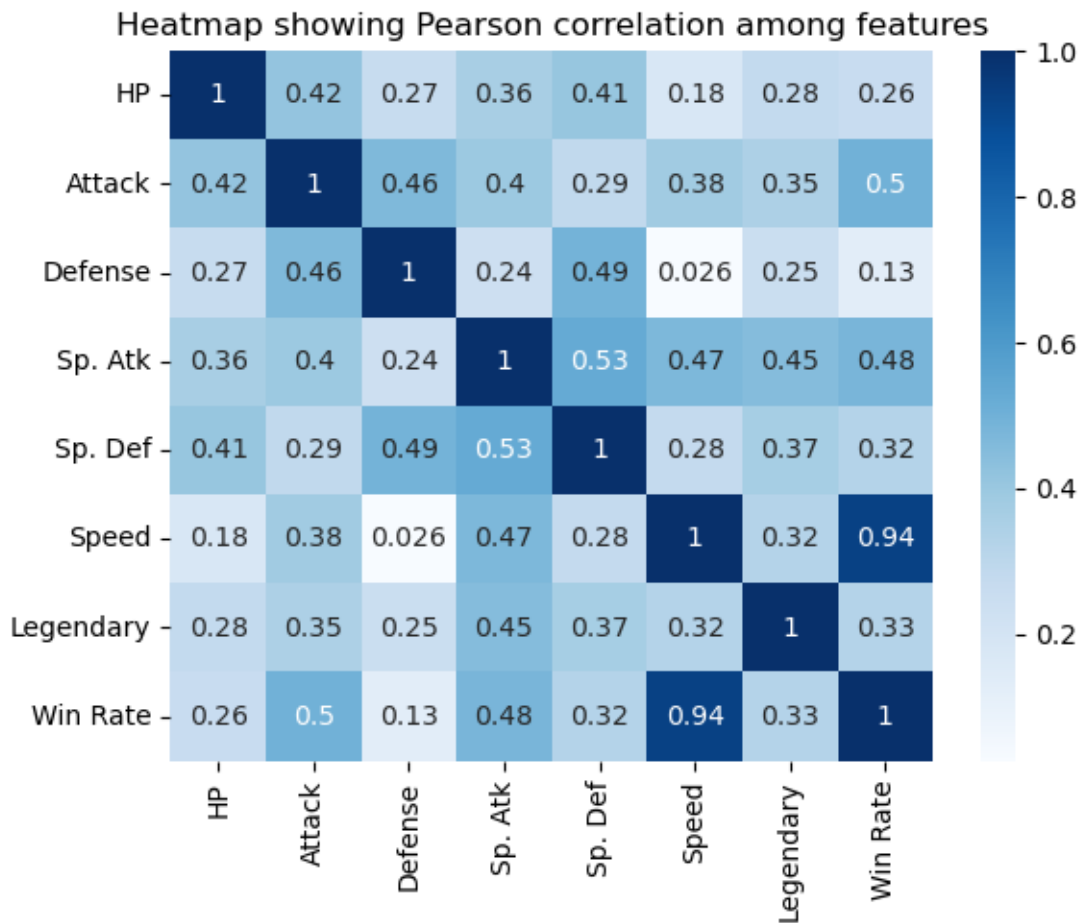
```
[14]: # Pearson correlation table & heatmap
import seaborn as sns

corr = pkmnWR.loc[:,col].corr(method='pearson', min_periods=1)
print(pkmnWR.loc[:,col].corr(method='pearson', min_periods=1))
sns.heatmap(corr, cmap="Blues", annot=True)
plt.title('Heatmap showing Pearson correlation among features',fontsize=12)
plt.savefig(r"C:\Users\User\Desktop\DSI\DATA1030-Fall12022\Taemin Huh DATA1030_
↳Project\figures\Pearson Correlation Heatmap.png")
```

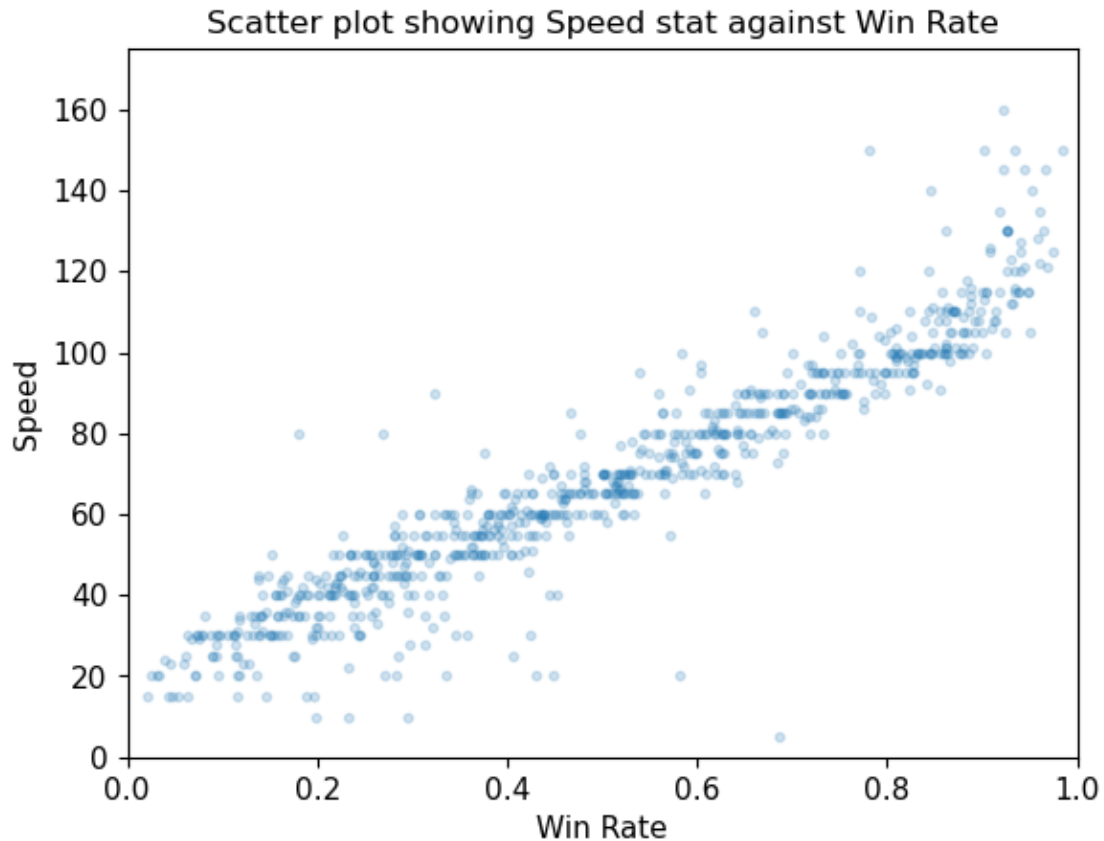
	HP	Attack	Defense	Sp. Atk	Sp. Def	Speed	\
HP	1.000000	0.417427	0.265230	0.363244	0.409110	0.179423	
Attack	0.417427	1.000000	0.464539	0.395211	0.288078	0.382310	
Defense	0.265230	0.464539	1.000000	0.237592	0.490118	0.025762	
Sp. Atk	0.363244	0.395211	0.237592	1.000000	0.529276	0.470548	
Sp. Def	0.409110	0.288078	0.490118	0.529276	1.000000	0.276715	
Speed	0.179423	0.382310	0.025762	0.470548	0.276715	1.000000	
Legendary	0.280265	0.348391	0.247921	0.448339	0.367030	0.323420	
Win Rate	0.258006	0.500181	0.129426	0.478940	0.324218	0.937742	

	Legendary	Win Rate
HP	0.280265	0.258006
Attack	0.348391	0.500181
Defense	0.247921	0.129426
Sp. Atk	0.448339	0.478940
Sp. Def	0.367030	0.324218

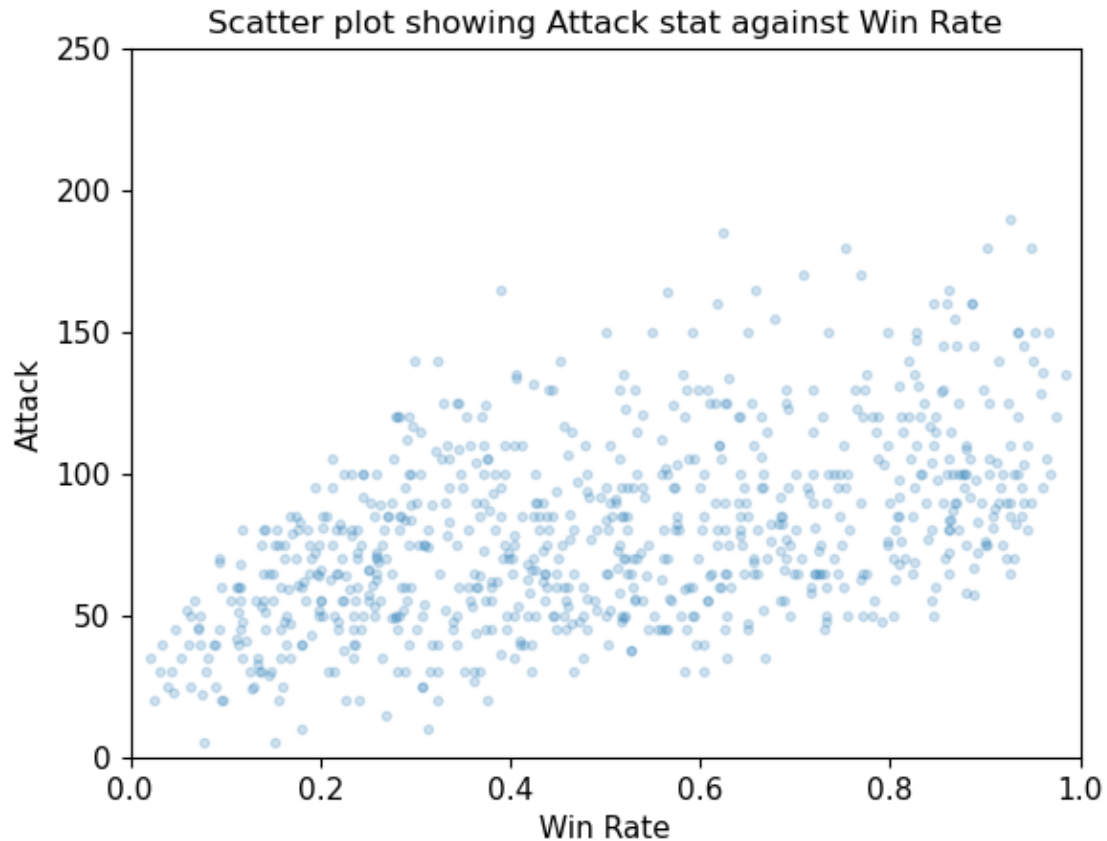
Speed	0.323420	0.937742
Legendary	1.000000	0.325007
Win Rate	0.325007	1.000000



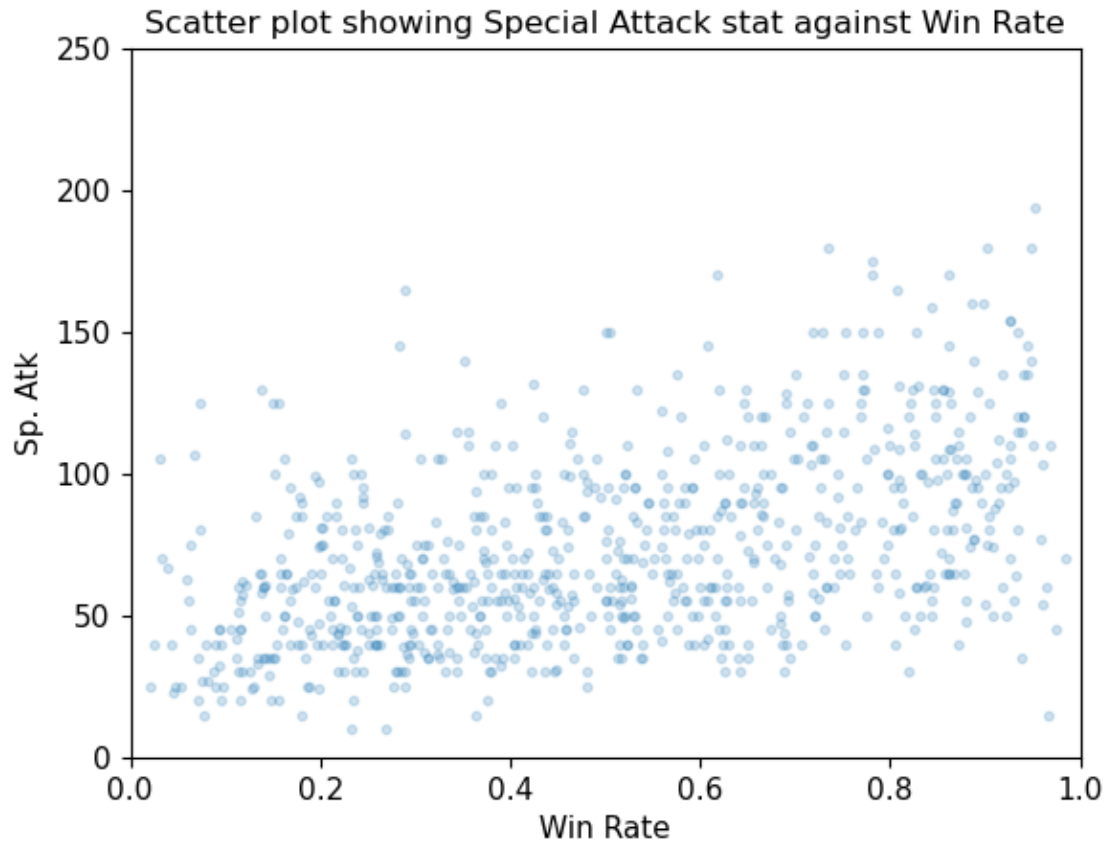
```
[15]: pkmnWR.plot.scatter('Win Rate', 'Speed',s=10,alpha=0.2,fontsize=11)
ax = plt.gca()
ax.set_xlim([0, 1])
ax.set_ylim([0, 175])
plt.xlabel('Win Rate',fontsize=11)
plt.ylabel('Speed',fontsize=11)
# txt="Scatter plot showing Speed stat against Win Rate."
# plt.figtext(0.5, -0.05, txt, wrap=True, horizontalalignment='center',
#             ↪ fontsize=10)
plt.title('Scatter plot showing Speed stat against Win Rate',fontsize=12)
plt.savefig(r"C:\Users\User\Desktop\DSI\DATA1030-Fall12022\Taemin Huh DATA1030_
            ↪ Project\figures\Speed-WR Scatter Plot.png")
plt.show()
```

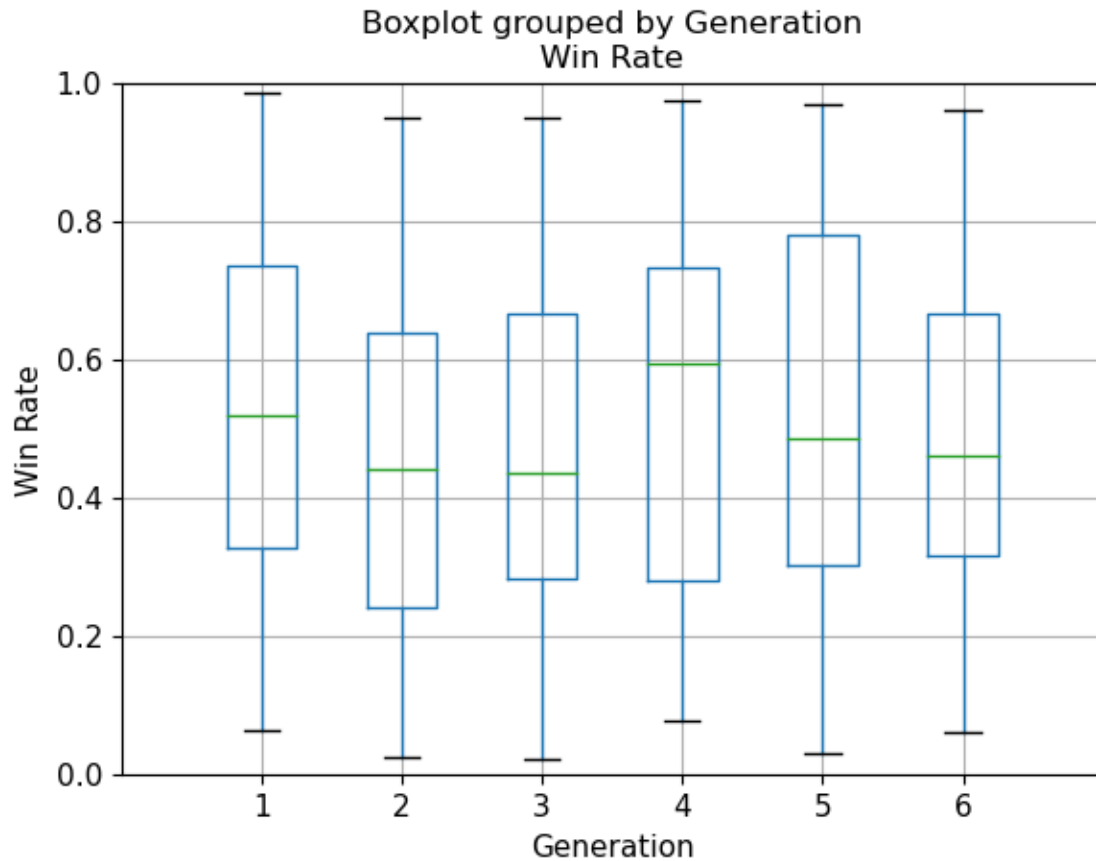
```
[16]: pkmnWR.plot.scatter('Win Rate', 'Attack',s=10,alpha=0.2,fontsize=11)
ax = plt.gca()
ax.set_xlim([0, 1])
ax.set_ylim([0, 250])
plt.xlabel('Win Rate',fontsize=11)
plt.ylabel('Attack',fontsize=11)
# txt="Scatter plot showing Attack stat against Win Rate."
# plt.figtext(0.5, -0.05, txt, wrap=True, horizontalalignment='center',
# ↪ fontsize=10)
plt.title('Scatter plot showing Attack stat against Win Rate', fontsize=12)
plt.savefig(r"C:\Users\User\Desktop\DSI\DATA1030-Fall2022\Taemin Huh DATA1030_
↪ Project\figures\Attack-WR Scatter Plot.png")
plt.show()
```



```
[17]: pkmnWR.plot.scatter('Win Rate', 'Sp. Atk',s=10,alpha=0.2,fontsize=11)
ax = plt.gca()
ax.set_xlim([0, 1])
ax.set_ylim([0, 250])
plt.xlabel('Win Rate',fontsize=11)
plt.ylabel('Sp. Atk',fontsize=11)
# txt="Scatter plot showing Special Attack stat against Win Rate."
# plt.figtext(0.5, -0.05, txt, wrap=True, horizontalalignment='center',
#             ↪fontsize=10)
plt.title('Scatter plot showing Special Attack stat against Win_
          ↪Rate',fontsize=12)
plt.savefig(r"C:\Users\User\Desktop\DSI\DATA1030-Fall2022\Taemin Huh DATA1030_
          ↪Project\figures\SpAtk-WR Scatter Plot.png")
plt.show()
```

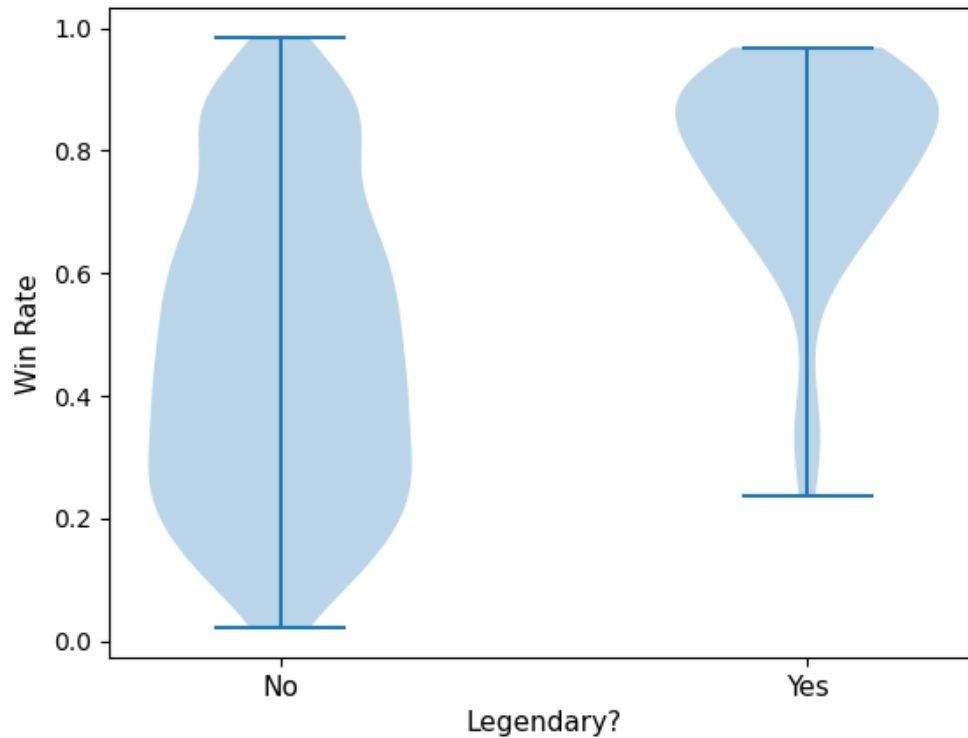


```
[18]: pkmnWR[['Win Rate', 'Generation']].boxplot(by='Generation', fontsize=11)
ax = plt.gca()
ax.set_xlim([0, 7])
ax.set_ylim([0, 1])
plt.ylabel('Win Rate', fontsize=11)
plt.xlabel('Generation', fontsize=11)
# txt="Box plot comparing Win Rate of Pokemons across 6 generations."
# plt.figtext(0.5, -0.05, txt, wrap=True, horizontalalignment='center',
# ↪ fontsize=10)
plt.savefig(r"C:\Users\User\Desktop\DSI\DATA1030-Fall2022\Taemin Huh DATA1030_
↪ Project\figures\Generation-WR Box Plot.png")
plt.show()
```



```
[19]: dataset = [pkmnWR[pkmnWR['Legendary']==False]['Win Rate'].values,
                pkmnWR[pkmnWR['Legendary']==True]['Win Rate'].values]
plt.violinplot(dataset = dataset)
plt.xticks([1,2],['No','Yes'],fontsize=11)
plt.ylabel('Win Rate',fontsize=11)
plt.xlabel('Legendary?',fontsize=11)
# txt="Violin plot comparing Win Rate of Legendary vs. non-Legendary Pokemons."
# plt.figtext(0.5, -0.05, txt, wrap=True, horizontalalignment='center',
#             ↪ fontsize=10)
plt.title('Violin plot comparing Win Rate of Legendary vs. non-Legendary
          ↪ Pokemons',fontsize=12)
plt.savefig(r"C:\Users\User\Desktop\DSI\DATA1030-Fall2022\Taemin Huh DATA1030
          ↪ Project\figures\Legendary Violin Plot.png")
plt.show()
```

Violin plot comparing Win Rate of Legendary vs. non-Legendary Pokemons



1.2 Preprocessing

```
[20]: from sklearn.model_selection import train_test_split
y = pkmnWR['Win Rate']
X = pkmnWR.loc[:, pkmnWR.columns != 'Win Rate']

print(X.head())
print(X.shape)
print(y)
```

	ID	Name	Type 1	Type 2	HP	Attack	Defense	Sp. Atk	Sp. Def	\
0	1	Bulbasaur	Grass	Poison	45	49	49	65	65	
1	2	Ivysaur	Grass	Poison	60	62	63	80	80	
2	3	Venusaur	Grass	Poison	80	82	83	100	100	
3	4	Mega Venusaur	Grass	Poison	80	100	123	122	120	
4	5	Charmander	Fire	NaN	39	52	43	60	50	

	Speed	Generation	Legendary	Total Win Count	Total Battle Count
0	45	1	False	37	133
1	60	1	False	46	121
2	80	1	False	89	132
3	80	1	False	70	125

```

4      65          1      False          55          112
(783, 14)
0      0.278195
1      0.380165
2      0.674242
3      0.560000
4      0.491071
...
795    0.371429
796    0.885496
797    0.504202
798    0.618056
799    0.619835
Name: Win Rate, Length: 783, dtype: float64

```

```

[21]: def basic_split(X,y,train_size,val_size,test_size,random_state):

        X_train, X_other, y_train, y_other = train_test_split(X,y,train_size =
↳train_size,random_state=random_state)
        print('training set:',X_train.shape, y_train.shape)
        print(X_other.shape, y_other.shape)

        X_val, X_test, y_val, y_test = train_test_split(X_other,y_other,train_size
↳= val_size/(val_size+test_size),random_state=random_state)
        print('validation set:',X_val.shape, y_val.shape)
        print('test set:',X_test.shape, y_test.shape)

        print(X_train.head())
        return X_train, y_train, X_val, y_val, X_test, y_test

print(basic_split(X,y,0.6,0.2,0.2,7))

```

```

training set: (469, 14) (469,)
(314, 14) (314,)
validation set: (157, 14) (157,)
test set: (157, 14) (157,)

```

	ID	Name	Type 1	Type 2	HP	Attack	Defense	Sp. Atk	\
95	96	Grimer	Poison	NaN	80	80	50	40	
221	222	Dunsparce	Normal	NaN	100	70	70	65	
139	140	Magikarp	Water	NaN	20	10	55	15	
541	542	Palkia	Water	Dragon	90	120	100	150	
7	8	Mega Charizard X	Fire	Dragon	78	130	111	130	

	Sp. Def	Speed	Generation	Legendary	Total Win Count	\
95	50	25	1	False	25	
221	65	45	2	False	26	
139	20	80	1	False	25	
541	120	100	4	True	115	

7	85	100	1	False	119
---	----	-----	---	-------	-----

Total Battle Count

95	142
221	137
139	139
541	149
7	139

(ID	Name	Type 1	Type 2	HP	Attack	Defense	Sp. Atk
\								
95	96	Grimer	Poison	NaN	80	80	50	40
221	222	Dunsparce	Normal	NaN	100	70	70	65
139	140	Magikarp	Water	NaN	20	10	55	15
541	542	Palkia	Water	Dragon	90	120	100	150
7	8	Mega Charizard X	Fire	Dragon	78	130	111	130
..
593	594	Gurdurr	Fighting	NaN	85	105	85	40
515	516	Rhyperior	Ground	Rock	115	140	130	55
550	551	Shaymin Land Forme	Grass	NaN	100	100	100	100
204	205	Jumpluff	Grass	Flying	75	55	70	55
183	184	Crobat	Poison	Flying	85	90	80	70

Sp.	Def	Speed	Generation	Legendary	Total Win Count	\
95	50	25	1	False	25	
221	65	45	2	False	26	
139	20	80	1	False	25	
541	120	100	4	True	115	
7	85	100	1	False	119	
..	
593	50	40	5	False	25	
515	55	40	4	False	47	
550	100	100	4	True	117	
204	95	110	2	False	114	
183	80	130	2	False	113	

Total Battle Count

95	142
221	137
139	139
541	149
7	139
..	...
593	118
515	104
550	136
204	135
183	122

[469 rows x 14 columns], 95 0.176056

221 0.189781
139 0.179856
541 0.771812
7 0.856115

...
593 0.211864
515 0.451923
550 0.860294
204 0.844444
183 0.926230

Name: Win Rate, Length: 469, dtype: float64, ID Name Type 1

Type 2 HP Attack Defense Sp. Atk \

767	768	Tyrantrum	Rock	Dragon	82	121	119	69
728	729	Diggersby	Normal	Ground	85	56	77	50
753	754	Aromatisse	Fairy	NaN	101	72	72	99
196	197	Mega Ampharos	Electric	Dragon	90	95	105	165
266	267	Pupitar	Rock	Ground	70	84	70	65
..
199	200	Azumarill	Water	Fairy	100	50	80	60
414	415	Regirock	Rock	NaN	80	100	200	50
261	262	Blissey	Normal	NaN	255	10	10	75
749	750	Doublade	Steel	Ghost	59	110	150	45
423	424	Groudon	Ground	NaN	100	150	140	100

	Sp. Def	Speed	Generation	Legendary	Total Win Count \
767	59	71	6	False	82
728	77	78	6	False	68
753	89	29	6	False	25
196	110	45	2	False	37
266	70	51	2	False	41
..
199	80	50	2	False	44
414	100	50	3	True	50
261	135	55	2	False	40
749	49	35	6	False	44
423	90	90	3	True	106

Total Battle Count

767	152
728	128
753	129
196	128
266	139
..	...
199	128
414	144
261	128

749 132
423 133

[157 rows x 14 columns], 767 0.539474

728 0.531250
753 0.193798
196 0.289062
266 0.294964

...
199 0.343750
414 0.347222
261 0.312500
749 0.333333
423 0.796992

Name: Win Rate, Length: 157, dtype: float64, ID Name

Type 1	Type 2	HP	Attack	Defense	\				
31	32		Raichu	Electric	NaN	60	90	55	
669	670		Lampent	Ghost	Fire	60	40	60	
345	346		Gulpin	Poison	NaN	70	43	53	
459	460	Wormadam	Sandy Cloak	Bug	Ground	60	79	105	
201	202		Politoed	Water	NaN	90	75	75	
..	
421	422		Kyogre	Water	NaN	100	100	90	
194	195		Flaaffy	Electric	NaN	70	55	55	
485	486		Bronzong	Steel	Psychic	67	89	116	
708	709	Landorus	Incarinate	Forme	Ground	Flying	89	125	90
676	677		Cryogonal	Ice	NaN	70	50	30	

	Sp. Atk	Sp. Def	Speed	Generation	Legendary	Total Win Count	\
31	90	80	110	1	False	105	
669	95	60	55	5	False	57	
345	43	53	40	3	False	21	
459	59	85	36	4	False	22	
201	90	100	70	2	False	65	
..	
421	150	140	90	3	True	92	
194	80	60	45	2	False	24	
485	79	116	33	4	False	37	
708	115	80	101	5	True	89	
676	95	135	105	5	False	119	

Total Battle Count

31	121
669	135
345	110
459	129
201	119
..	...

421	128
194	108
485	141
708	102
676	148

[157 rows x 14 columns], 31 0.867769

669	0.422222
345	0.190909
459	0.170543
201	0.546218

...	
421	0.718750
194	0.222222
485	0.262411
708	0.872549
676	0.804054

Name: Win Rate, Length: 157, dtype: float64)

```
[22]: from sklearn.compose import ColumnTransformer
from sklearn.pipeline import Pipeline
from sklearn.preprocessing import OneHotEncoder, MinMaxScaler

X_train, y_train, X_val, y_val, X_test, y_test = basic_split(X,y,0.6,0.2,0.2,7)

# Pre-processing with OneHotEncoder and MinMaxScaler

onehot_ftrs = ['Type 1', 'Type 2', 'Generation', 'Legendary']
minmax_ftrs = ['HP', 'Attack', 'Defense', 'Sp. Atk', 'Sp. Def', 'Speed']

preprocessor = ColumnTransformer(
    transformers=[
        ('onehot', OneHotEncoder(sparse=False,handle_unknown='ignore'),
↳onehot_ftrs),
        ('minmax', MinMaxScaler(), minmax_ftrs)])

clf = Pipeline(steps=[('preprocessor', preprocessor)])

X_train_prep = clf.fit_transform(X_train)
X_val_prep = clf.transform(X_val)
X_test_prep = clf.transform(X_test)

print('X_train shape:',X_train.shape)
print('X_train_prep shape:',X_train_prep.shape)
print(X_train_prep)
```

training set: (469, 14) (469,)
(314, 14) (314,)

validation set: (157, 14) (157,)

test set: (157, 14) (157,)

	ID	Name	Type 1	Type 2	HP	Attack	Defense	Sp. Atk	\
95	96	Grimer	Poison	NaN	80	80	50	40	
221	222	Dunsparce	Normal	NaN	100	70	70	65	
139	140	Magikarp	Water	NaN	20	10	55	15	
541	542	Palkia	Water	Dragon	90	120	100	150	
7	8	Mega Charizard X	Fire	Dragon	78	130	111	130	

	Sp. Def	Speed	Generation	Legendary	Total Win Count	\
95	50	25	1	False	25	
221	65	45	2	False	26	
139	20	80	1	False	25	
541	120	100	4	True	115	
7	85	100	1	False	119	

	Total Battle Count
95	142
221	137
139	139
541	149
7	139

X_train shape: (469, 14)

X_train_prep shape: (469, 51)

```
[[0. 0. 0. ... 0.16304348 0.16666667 0.13793103]
 [0. 0. 0. ... 0.29891304 0.25 0.27586207]
 [0. 0. 0. ... 0.02717391 0. 0.51724138]
 ...
 [0. 0. 0. ... 0.48913043 0.44444444 0.65517241]
 [0. 0. 0. ... 0.24456522 0.41666667 0.72413793]
 [0. 0. 0. ... 0.32608696 0.33333333 0.86206897]]
```