In [1]:
```
import pandas as pd

# Load the hitting and pitching data
hitting = pd.read_csv('hittingstats.csv')
pitching = pd.read_csv('pitchingstats.csv')

# Quick look at the data
print(hitting.head())
print(pitching.head())
```

```
                    Tm  #Bat  BatAge   R/G    G    PA    AB    R    H   2B  \
0  Arizona Diamondbacks    51    28.6  3.47  162  6284  5522  886  1452  271
1         Atlanta Braves    55    29.3  4.35  162  6075  5481  704  1333  273
2      Baltimore Orioles    60    26.9  4.85  162  6176  5567  786  1391  262
3        Boston Red Sox    57    27.3  4.44  162  6192  5577  751  1404  311
4         Chicago Cubs    56    27.8  4.54  162  6116  5441  736  1318  253

   ...   SLG   OPS  OPS+    TB  GDP  HBP  SH  SF  IBB  LOB
0  ...  0.448  0.777  115  2430  112   84  34  66   17  1111
1  ...  0.415  0.724  100  2275  119   58   9  39   11  1088
2  ...  0.435  0.751  108  2424   75   66   6  45   14  1108
3  ...  0.462  0.761  104  2557  110   72   7  40   30  1116
4  ...  0.393  0.710  105  2139   89   70  17  41   19  1105

[5 rows x 29 columns]
                    Tm   #P  PAge  RA/G   W   L  W-L%   ERA   G  GS  ...  \
0  Arizona Diamondbacks   33  28.6  4.86  89  73  0.549  4.62  162  162  ...
1         Atlanta Braves   30  31.5  3.75  89  73  0.549  3.49  162  162  ...
2      Baltimore Orioles   34  29.6  4.31  91  71  0.562  3.94  162  162  ...
3        Boston Red Sox   35  29.5  4.41  81  81  0.500  4.04  162  162  ...
4         Chicago Cubs   36  29.2  4.13  83  79  0.512  3.78  162  162  ...

   BF  ERA+  FIP  WHIP   H9  HR9  BB9  SO9  SO/W  LOB
0  6195   91  4.09  1.355  9.2  1.1  3.0  8.2  2.73  1096
1  5978  120  3.44  1.196  8.0  0.9  2.8  9.7  3.46  1043
2  6071   96  3.93  1.237  8.1  1.1  3.0  8.6  2.87  1062
3  6154  106  4.10  1.256  8.4  1.2  2.9  8.4  2.93  1080
4  6052  106  4.04  1.248  8.2  1.3  3.0  8.5  2.78  1080

[5 rows x 30 columns]
```

In [2]:
```
# Merge datasets on team name
data = pd.merge(hitting, pitching, on='Tm', suffixes=('_bat', '_pit'))

# Drop unneeded columns
data = data.drop(columns=['#Bat', '#P'])

# Preview merged data
print(data.head())
```

```
                    Tm  BatAge  R/G  G_bat   PA    AB  R_bat  H_bat  2B  \
0  Arizona Diamondbacks    28.6  3.47    162  6284  5522    886   1452  271
1         Atlanta Braves    29.3  4.35    162  6075  5481    704   1333  273
2      Baltimore Orioles    26.9  4.85    162  6176  5567    786   1391  262
3        Boston Red Sox    27.3  4.44    162  6192  5577    751   1404  311
4         Chicago Cubs    27.8  4.54    162  6116  5441    736   1318  253

   3B  ...   BF  ERA+  FIP  WHIP   H9  HR9  BB9  SO9  SO/W  LOB_pit
0  37  ...  6195    91  4.09  1.355  9.2  1.1  3.0  8.2  2.73     1096
1  15  ...  5978   120  3.44  1.196  8.0  0.9  2.8  9.7  3.46     1043
2  33  ...  6071    96  3.93  1.237  8.1  1.1  3.0  8.6  2.87     1062
3  30  ...  6154   106  4.10  1.256  8.4  1.2  2.9  8.4  2.93     1080
4  29  ...  6052   106  4.04  1.248  8.2  1.3  3.0  8.5  2.78     1080

[5 rows x 62 columns]
```

In [3]:
```
# Target variable
y = data['W']

# Drop columns we won't use as predictors
X = data.drop(columns=['Tm', 'W', 'L', 'W-L%'])

# Optional: Normalize or standardize here

# Handle categorical data if any (not much here)
```

In [4]:
```
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LinearRegression
from sklearn.metrics import mean_absolute_error, r2_score

# Train-test split
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)

# Fit a linear regression model
model = LinearRegression()
model.fit(X_train, y_train)

# Predict
y_pred = model.predict(X_test)

# Evaluate
print("R2 Score:", r2_score(y_test, y_pred))
print("MAE:", mean_absolute_error(y_test, y_pred))
```

```
R2 Score: -0.2882260856497397
MAE: 9.600009082171349
```

In [6]:
```
# Select only numeric columns
numeric_data = data.select_dtypes(include='number')

# Correlation with wins
correlation = numeric_data.corr()['W'].sort_values(ascending=False)

print(correlation)

# Plot the correlations
import seaborn as sns
import matplotlib.pyplot as plt

plt.figure(figsize=(10, 6))
sns.barplot(x=correlation.values, y=correlation.index)
plt.title('Feature Correlation with Wins')
plt.show()
```

```
W          1.000000
W-L%       0.999928
R/G        0.814450
R_bat      0.811557
OPS+       0.801851
             ...
WHIP      -0.733830
ERA       -0.735495
R_pit     -0.746551
HR/G      -0.748835
L         -0.999003
Name: W, Length: 61, dtype: float64
```


In [7]:
```
selected_features = ['R/G', 'R_bat', 'OPS+', 'RA/G', 'R_pit', 'ERA', 'WHIP']
X_selected = data[selected_features]

from sklearn.model_selection import train_test_split
from sklearn.linear_model import LinearRegression
from sklearn.metrics import r2_score, mean_absolute_error

# Split again with selected features
X_train, X_test, y_train, y_test = train_test_split(X_selected, y, test_size=0.2, random_state=42)

# Linear Regression
model = LinearRegression()
model.fit(X_train, y_train)
y_pred = model.predict(X_test)

# Evaluate
print("R2 Score:", r2_score(y_test, y_pred))
print("MAE:", mean_absolute_error(y_test, y_pred))
```

```
R2 Score: 0.5990828815415422
MAE: 6.990891266572347
```

In [9]:
```
# Try Random Forest
from sklearn.ensemble import RandomForestRegressor

rf = RandomForestRegressor(random_state=42)
rf.fit(X_train, y_train)
y_pred_rf = rf.predict(X_test)

print("Random Forest R2:", r2_score(y_test, y_pred_rf))
print("Random Forest MAE:", mean_absolute_error(y_test, y_pred_rf))
```

```
Random Forest R2: 0.9837573619631902
Random Forest MAE: 1.1766666666666634
```

In [10]:
```
# Predict wins for all teams in the dataset
predicted_wins = rf.predict(X_selected)

# Add predictions to the original DataFrame
data['Predicted_Wins'] = predicted_wins.round(1)  # Round to 1 decimal for clarity

# Sort by most predicted wins
data_sorted = data[['Tm', 'Predicted_Wins']].sort_values(by='Predicted_Wins', ascending=False)

print(data_sorted)
```

```
                    Tm  Predicted_Wins
13    Los Angeles Dodgers            96.1
20  Philadelphia Phillies            93.7
18     New York Yankees            93.0
2       Baltimore Orioles            91.7
22       San Diego Padres            91.6
15      Milwaukee Brewers            91.2
7     Cleveland Guardians            90.6
17        New York Mets            89.2
1          Atlanta Braves            88.9
10        Houston Astros            88.6
11    Kansas City Royals            86.8
23     Seattle Mariners            85.3
9        Detroit Tigers            84.9
4          Chicago Cubs            84.6
16     Minnesota Twins            83.5
0   Arizona Diamondbacks            82.2
3        Boston Red Sox            81.9
24    St. Louis Cardinals            81.2
24  San Francisco Giants            79.6
6      Cincinnati Reds            79.0
26     Tampa Bay Rays            78.8
25      Texas Rangers            78.2
28    Toronto Blue Jays            75.6
21   Pittsburgh Pirates            75.4
29  Washington Nationals            71.1
19      Oakland Athletics            69.6
12    Los Angeles Angels            64.4
8     Colorado Rockies            63.4
14       Miami Marlins            60.3
5      Chicago White Sox            49.8
```

In [11]:
```
import matplotlib.pyplot as plt

plt.figure(figsize=(12, 10))
plt.barh(data_sorted['Tm'], data_sorted['Predicted_Wins'], color='skyblue')
plt.xlabel("Predicted Wins")
plt.title("Predicted MLB Wins (out of 162 games)")
plt.gca().invert_yaxis()  # Highest wins on top
plt.grid(axis='x', linestyle='--', alpha=0.7)
plt.tight_layout()
plt.show()
```


In [12]:
```
plt.figure(figsize=(12, 10))
bars = plt.barh(data_sorted['Tm'], data_sorted['Predicted_Wins'], color='skyblue')
plt.axvline(90, color='red', linestyle='--', label='Playoff Line (90 Wins)')
plt.xlabel("Predicted Wins")
plt.title("Predicted MLB Wins with Playoff Threshold")
plt.gca().invert_yaxis()
plt.grid(axis='x', linestyle='--', alpha=0.7)
plt.legend()
plt.tight_layout()
plt.show()
```


In [ ]:
```

```