

```
In [1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
import xgboost as xgb
```

## Data Cleaning and Analysis, Feature Engineering

```
In [2]: df = pd.read_csv('Global_Space_Exploration_Dataset.csv')
pd.set_option('display.max_columns', None)
```

```
In [3]: df.head(20)
```

Out[3]:

	Country	Year	Mission Name	Mission Type	Launch Site	Satellite Type	Budget (in Billion \$)
0	China	2008	Sharable tertiary superstructure	Manned	Sheilatown	Communication	16.20
1	Japan	2018	Re-engineered composite flexibility	Manned	New Ericfurt	Communication	29.04
2	Israel	2013	Reactive disintermediate projection	Manned	Port Kaitlynstad	Communication	28.73
3	UAE	2010	Grass-roots 6thgeneration implementation	Unmanned	Mariastad	Spy	37.27
4	India	2006	Balanced discrete orchestration	Manned	North Jasonborough	Weather	18.95
5	USA	2011	Down-sized holistic methodology	Unmanned	North Kevin	Research	22.76
6	Germany	2011	Adaptive coherent definition	Manned	Wilsonburgh	Spy	9.33
7	India	2012	Innovative 6thgeneration algorithm	Unmanned	South William	Weather	6.62
8	Israel	2024	Business-focused exuding contingency	Manned	Edwardstad	Navigation	13.25
9	Israel	2011	Cross-group incremental function	Unmanned	Port Carla	Communication	23.76
10	France	2014	Reactive heuristic pricing structure	Unmanned	South Sarathon	Research	43.67
11	UK	2007	Innovative client-server matrix	Manned	Marcusborough	Spy	42.47
12	Russia	2023	Up-sized bifurcated conglomeration	Unmanned	North Shannon	Communication	11.68
13	China	2024	Public-key disintermediate matrix	Manned	Kathrynmouth	Research	29.52

	Country	Year	Mission Name	Mission Type	Launch Site	Satellite Type	Budget (in Billion \$)
14	UAE	2002	Vision-oriented fresh-thinking pricing structure	Manned	Whiteside	Spy	37.86
15	India	2020	Enterprise-wide heuristic knowledge user	Unmanned	Rodriguezshire	Communication	25.79
16	Russia	2017	Innovative zero tolerance workforce	Unmanned	West Katherineville	Spy	21.53
17	USA	2013	Digitized intangible encryption	Manned	New CassandraSide	Navigation	4.22
18	Germany	2019	Organic tertiary access	Manned	Lamville	Spy	47.41
19	India	2020	Phased context-sensitive intranet	Unmanned	Popehaven	Research	35.59

In [4]: `df.shape`

Out[4]: (3000, 12)

In [5]: `df.isna().sum()`

Out[5]:

Country	0
Year	0
Mission Name	0
Mission Type	0
Launch Site	0
Satellite Type	0
Budget (in Billion \$)	0
Success Rate (%)	0
Technology Used	0
Environmental Impact	0
Collaborating Countries	0
Duration (in Days)	0

`dtype: int64`

In [6]: `df.duplicated().sum()`

Out[6]: 0

In [7]: `df['Collaborating Countries'] = df['Collaborating Countries'].str.split(', ')`

In [8]: `df = df.explode('Collaborating Countries')`

```
In [9]: df = df.rename(columns={'Country': 'Main Country', 'Collaborating Countries': 'Co
```

```
In [10]: df.head(10)
```

Out[10]:

	Main Country	Year	Mission Name	Mission Type	Launch Site	Satellite Type	Budget (in Billion \$)	St
0	China	2008	Sharable tertiary superstructure	Manned	Sheilatown	Communication	16.20	
0	China	2008	Sharable tertiary superstructure	Manned	Sheilatown	Communication	16.20	
0	China	2008	Sharable tertiary superstructure	Manned	Sheilatown	Communication	16.20	
1	Japan	2018	Re-engineered composite flexibility	Manned	New Ericfurt	Communication	29.04	
1	Japan	2018	Re-engineered composite flexibility	Manned	New Ericfurt	Communication	29.04	
2	Israel	2013	Reactive disintermediate projection	Manned	Port Kaitlynstad	Communication	28.73	
2	Israel	2013	Reactive disintermediate projection	Manned	Port Kaitlynstad	Communication	28.73	
2	Israel	2013	Reactive disintermediate projection	Manned	Port Kaitlynstad	Communication	28.73	
3	UAE	2010	Grass-roots 6thgeneration implementation	Unmanned	Mariastad		Spy	37.27
4	India	2006	Balanced discrete orchestration	Manned	North Jasonborough	Weather		18.95

```
In [11]: df['Mission Name'].drop_duplicates()
print("There are 3000 rows and 3000 individual mission names with drop duplicates
      "Each mission has a unique name so therefore, 'Mission Name' won't be useful in
      "our machine learning project and can be dropped\n"
      )
```

There are 3000 rows and 3000 individual mission names with drop duplicates.  
Each mission has a unique name so therefore, 'Mission Name' won't be useful in our machine learning project and can be dropped

```
In [12]: df = df.drop('Mission Name', axis=1)
```

```
In [13]: df.head()
```

Out[13]:

	Main Country	Year	Mission Type	Launch Site	Satellite Type	Budget (in Billion \$)	Success Rate (%)	Technology Used	E
0	China	2008	Manned	Sheilatown	Communication	16.20	90	Nuclear Propulsion	
0	China	2008	Manned	Sheilatown	Communication	16.20	90	Nuclear Propulsion	
0	China	2008	Manned	Sheilatown	Communication	16.20	90	Nuclear Propulsion	
1	Japan	2018	Manned	New Ericfurt	Communication	29.04	99	Solar Propulsion	
1	Japan	2018	Manned	New Ericfurt	Communication	29.04	99	Solar Propulsion	



```
In [14]: df['Launch Site'] = df['Launch Site'].str.strip()
df['Launch Site'].drop_duplicates()
```

```
Out[14]: 0              Sheilatown
1              New Ericfurt
2            Port Kaitlynstad
3             Mariastad
4        North Jasonborough
...
2994      East Elaineburgh
2995      East Shawna
2996    Douglasborough
2997      Bellhaven
2998   Deniseview
Name: Launch Site, Length: 2702, dtype: object
```

```
In [15]: print(
    "There are 3,000 rows and 2,999 unique mission names after dropping duplicates."
    "All missions except one have a unique launch site, so 'Launch Site' is not
    "for our machine learning project and can be dropped."
)
```

There are 3,000 rows and 2,999 unique mission names after dropping duplicates.  
All missions except one have a unique launch site, so 'Launch Site' is not useful  
for our machine learning project and can be dropped.

```
In [16]: df = df.drop('Launch Site', axis=1)
```

```
In [17]: df.head()
```

Out[17]:

	Main Country	Year	Mission Type	Satellite Type	Budget (in Billion \$)	Success Rate (%)	Technology Used	Environmental Impact
0	China	2008	Manned	Communication	16.20	90	Nuclear Propulsion	Medium
0	China	2008	Manned	Communication	16.20	90	Nuclear Propulsion	Medium
0	China	2008	Manned	Communication	16.20	90	Nuclear Propulsion	Medium
1	Japan	2018	Manned	Communication	29.04	99	Solar Propulsion	High
1	Japan	2018	Manned	Communication	29.04	99	Solar Propulsion	High

In [18]: `df['Duration (in Years)'] = (df['Duration (in Days)'] / 365.25)`In [19]: `df['Budget per Year (in Billion $)'] = (df['Budget (in Billion $)'] / df['Duration (in Years)'])`In [20]: `df.head()`

Out[20]:

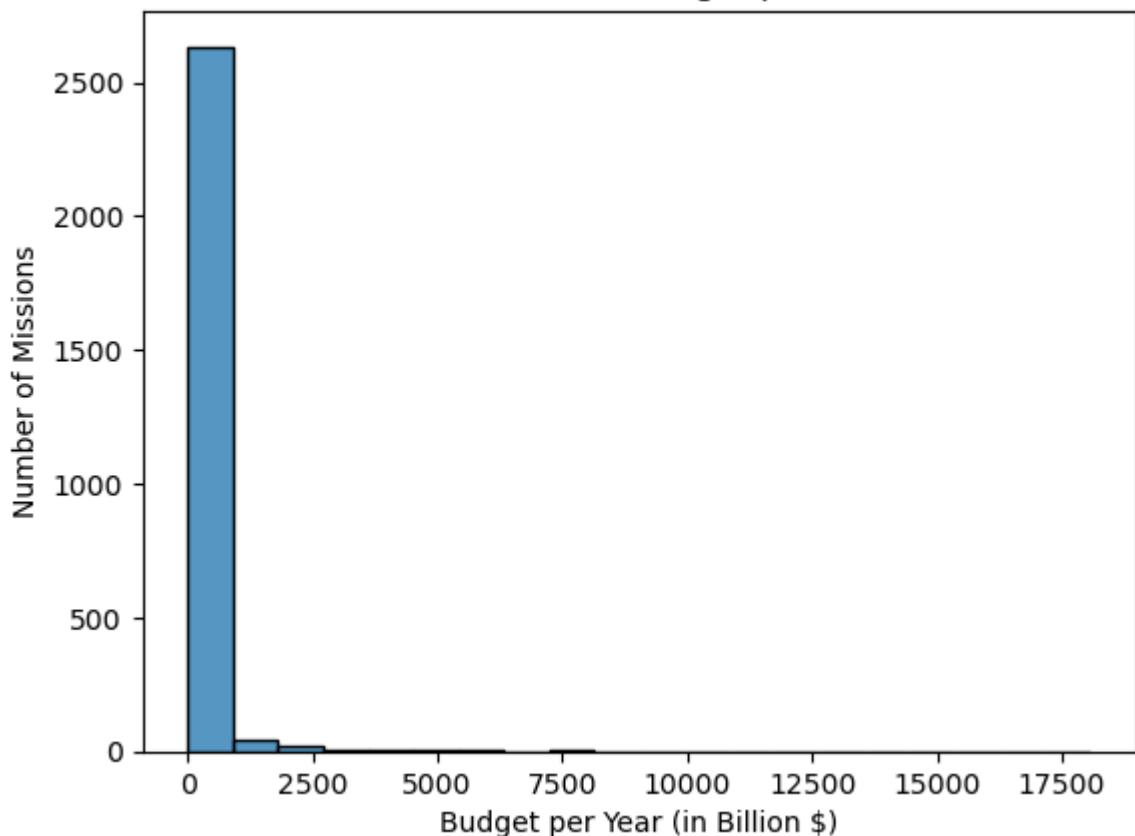
	Main Country	Year	Mission Type	Satellite Type	Budget (in Billion \$)	Success Rate (%)	Technology Used	Environmental Impact
0	China	2008	Manned	Communication	16.20	90	Nuclear Propulsion	Medium
0	China	2008	Manned	Communication	16.20	90	Nuclear Propulsion	Medium
0	China	2008	Manned	Communication	16.20	90	Nuclear Propulsion	Medium
1	Japan	2018	Manned	Communication	29.04	99	Solar Propulsion	High
1	Japan	2018	Manned	Communication	29.04	99	Solar Propulsion	High



## Exploratory Data Analysis

In [21]: `sns.histplot(df['Budget per Year (in Billion $)'].drop_duplicates(), bins=20)  
plt.xlabel('Budget per Year (in Billion $)')  
plt.ylabel('Number of Missions')  
plt.title('Distribution of Budget per Year')  
plt.show()`

### Distribution of Budget per Year



```
In [22]: df['Budget per Year (in Billion $)'].max()
```

```
Out[22]: 18072.57
```

```
In [23]: df['Budget per Year (in Billion $)'].min()
```

```
Out[23]: 0.62
```

```
In [24]: df['Budget Category'] = pd.cut(df['Budget per Year (in Billion $)'], bins=[0, 10000, 20000, 30000, 40000, 50000, 60000, 70000, 80000, 90000, 100000, 110000, 120000, 130000, 140000, 150000, 160000, 170000, 180000, 190000, 200000, 210000, 220000, 230000, 240000, 250000, 260000, 270000, 280000, 290000, 300000, 310000, 320000, 330000, 340000, 350000, 360000, 370000, 380000, 390000, 400000, 410000, 420000, 430000, 440000, 450000, 460000, 470000, 480000, 490000, 500000, 510000, 520000, 530000, 540000, 550000, 560000, 570000, 580000, 590000, 600000, 610000, 620000, 630000, 640000, 650000, 660000, 670000, 680000, 690000, 700000, 710000, 720000, 730000, 740000, 750000, 760000, 770000, 780000, 790000, 800000, 810000, 820000, 830000, 840000, 850000, 860000, 870000, 880000, 890000, 900000, 910000, 920000, 930000, 940000, 950000, 960000, 970000, 980000, 990000, 1000000])
```

```
In [25]: df.head()
```

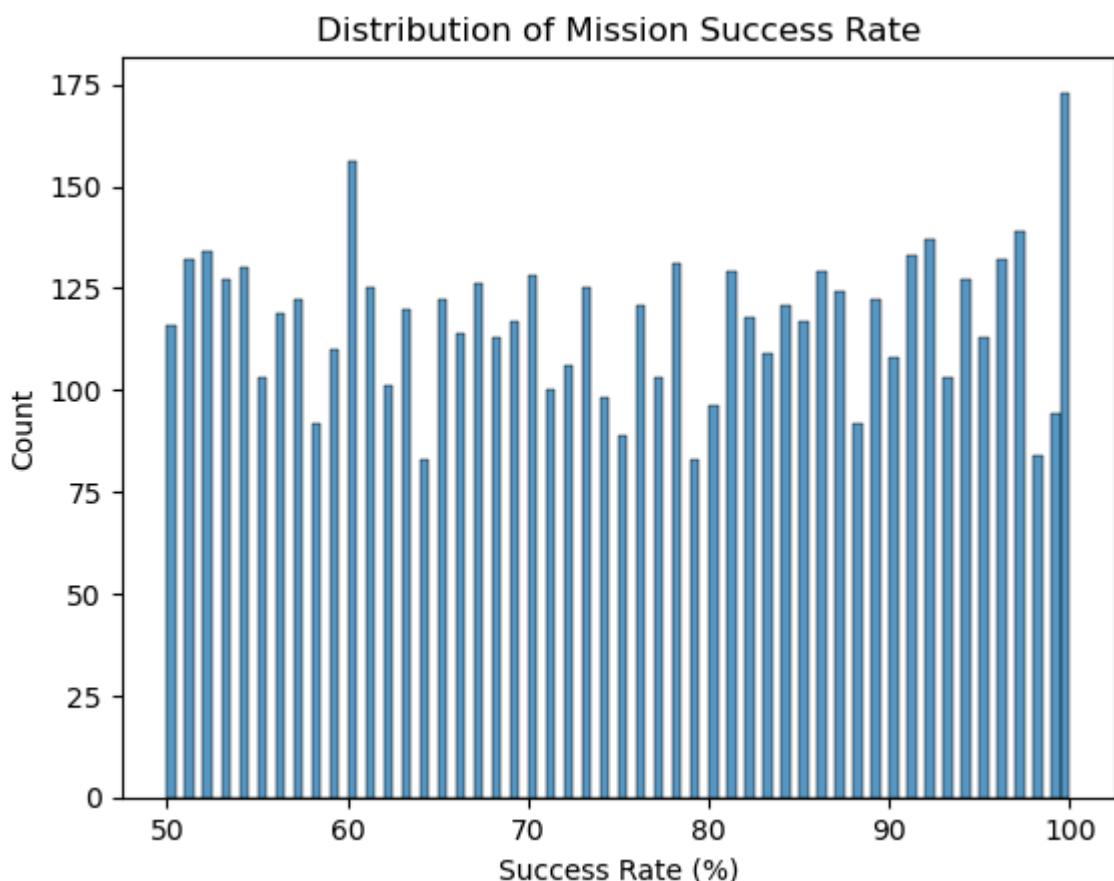
Out[25]:

Main Country	Year	Mission Type	Satellite Type	Budget (in Billion \$)	Success Rate (%)	Technology Used	Environmental Impact
0 China	2008	Manned	Communication	16.20	90	Nuclear Propulsion	Medium
0 China	2008	Manned	Communication	16.20	90	Nuclear Propulsion	Medium
0 China	2008	Manned	Communication	16.20	90	Nuclear Propulsion	Medium
1 Japan	2018	Manned	Communication	29.04	99	Solar Propulsion	High
1 Japan	2018	Manned	Communication	29.04	99	Solar Propulsion	High



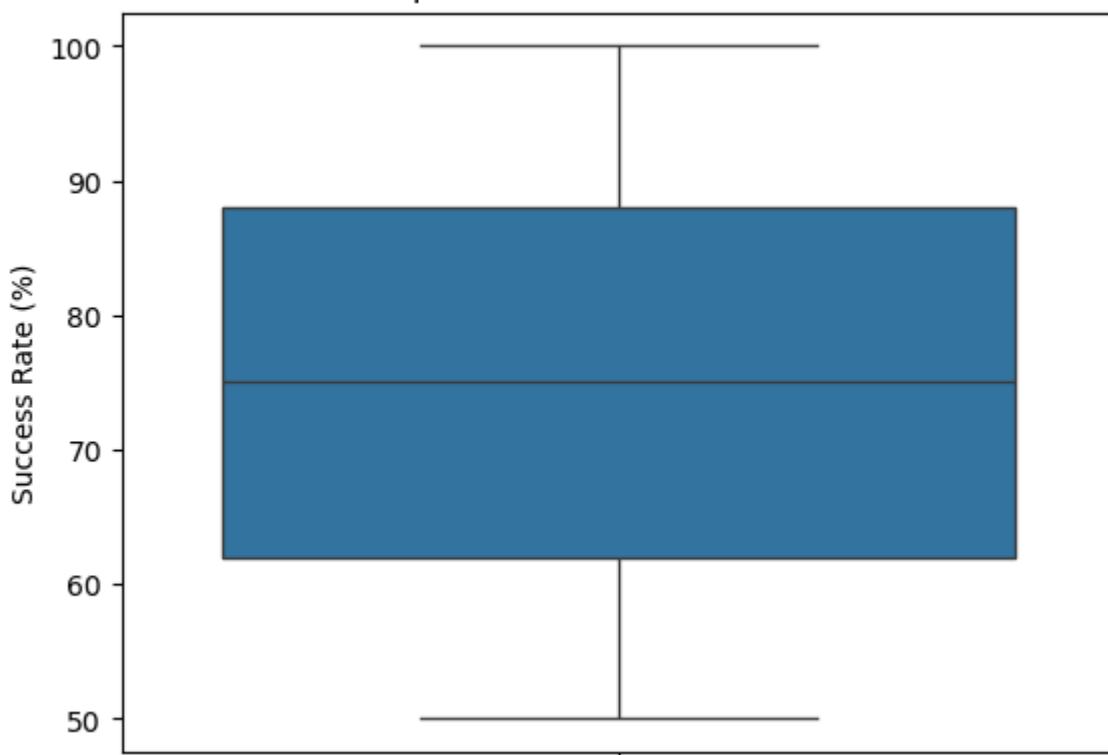
In [26]:

```
sns.histplot(df['Success Rate (%)'], bins=100)
plt.title('Distribution of Mission Success Rate')
plt.show()
```

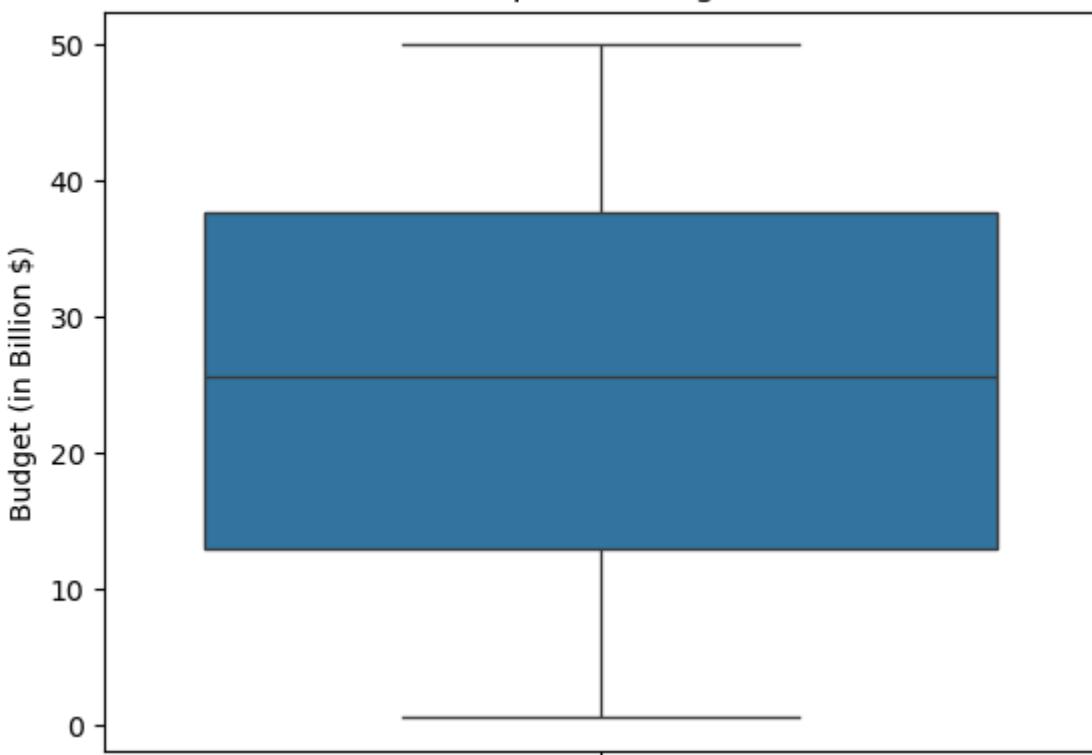


In [27]:

```
sns.boxplot(df['Success Rate (%)'])
plt.title('Boxplot of Mission Success Rate')
plt.show()
```

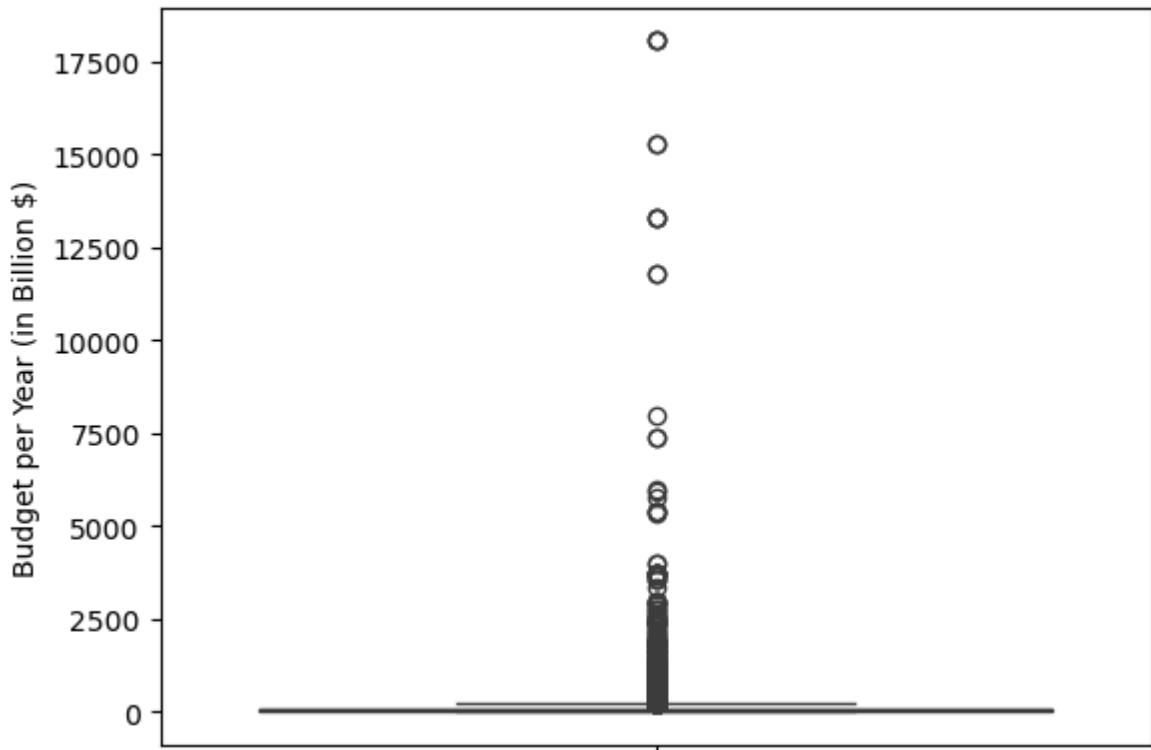
**Boxplot of Mission Success Rate**

```
In [28]: sns.boxplot(df['Budget (in Billion $)'])  
plt.title('Boxplot of Budget')  
plt.show()
```

**Boxplot of Budget**

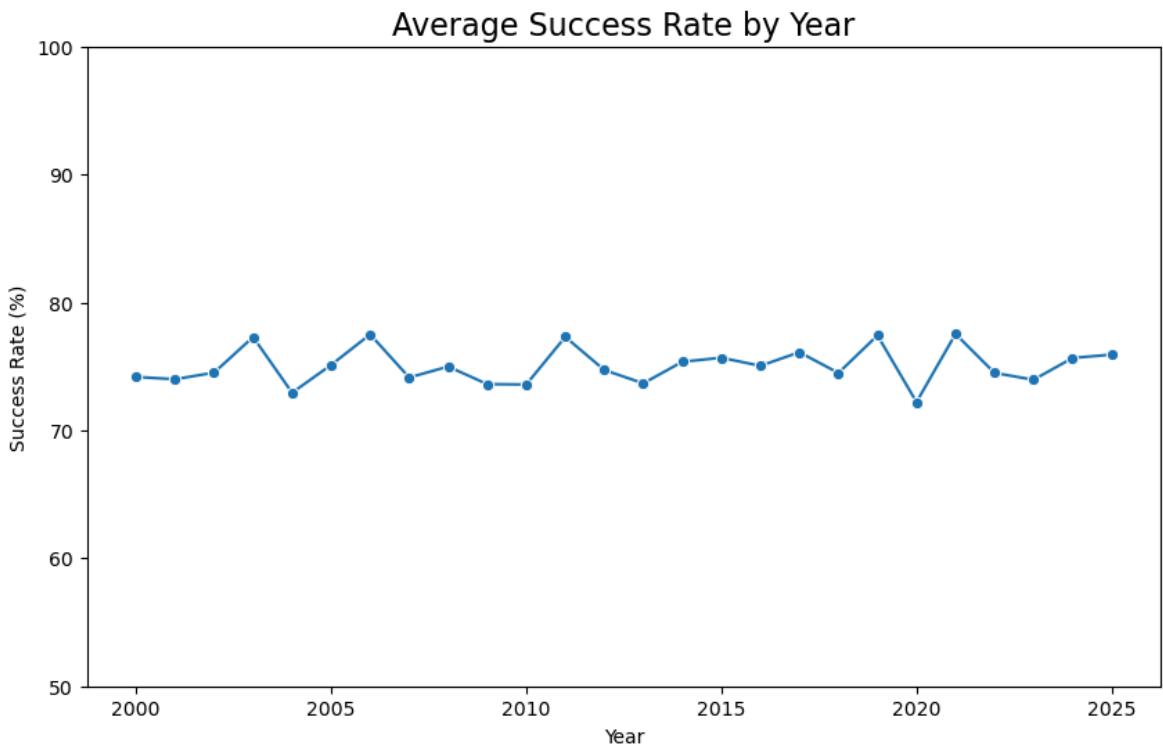
```
In [29]: sns.boxplot(df['Budget per Year (in Billion $)'])  
plt.title('Boxplot of Budget per year')  
plt.show()
```

Boxplot of Budget per year



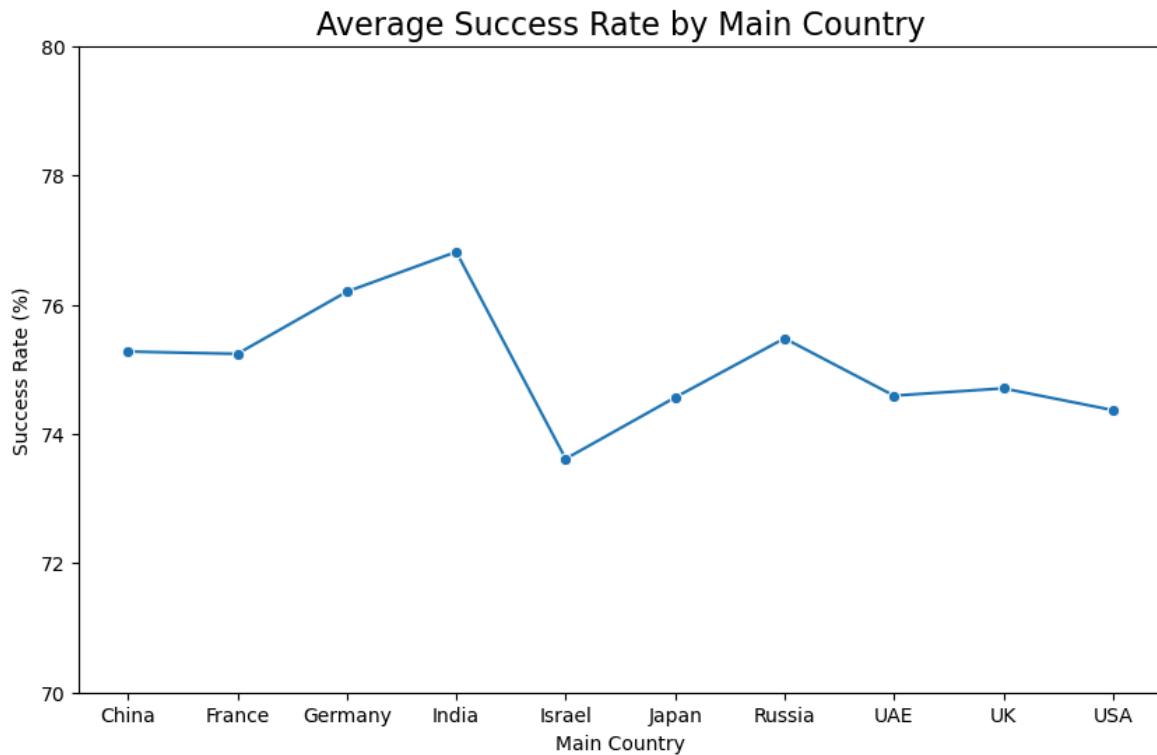
```
In [30]: avg_success_year = df.groupby('Year')['Success Rate (%)'].mean().reset_index()
```

```
In [31]: plt.figure(figsize=(10,6))
sns.lineplot(data=avg_success_year, x='Year', y='Success Rate (%)', marker='o')
plt.title('Average Success Rate by Year', fontsize=16)
plt.ylim(50, 100)
plt.show()
```



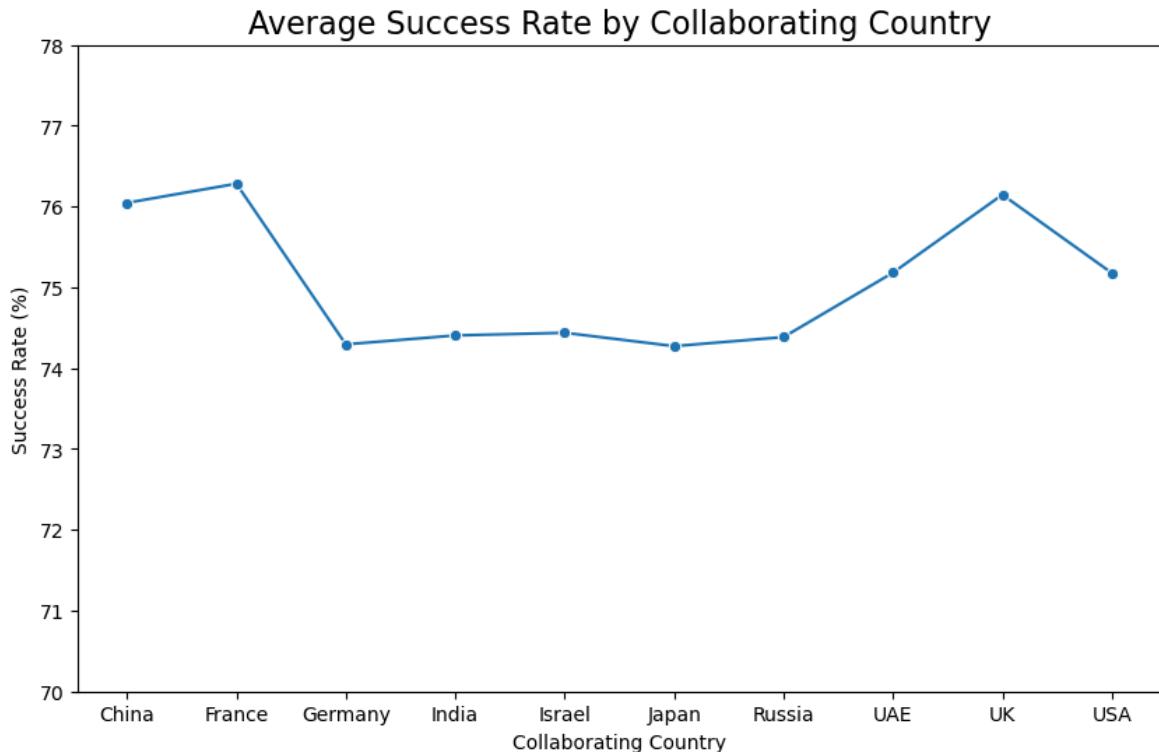
```
In [32]: avg_success_country = df.groupby('Main Country')['Success Rate (%)'].mean().reset_index()
```

```
In [33]: plt.figure(figsize=(10,6))
sns.lineplot(data=avg_success_country , x='Main Country', y='Success Rate (%)',
plt.title('Average Success Rate by Main Country', fontsize=16)
plt.ylim(70, 80)
plt.show()
```



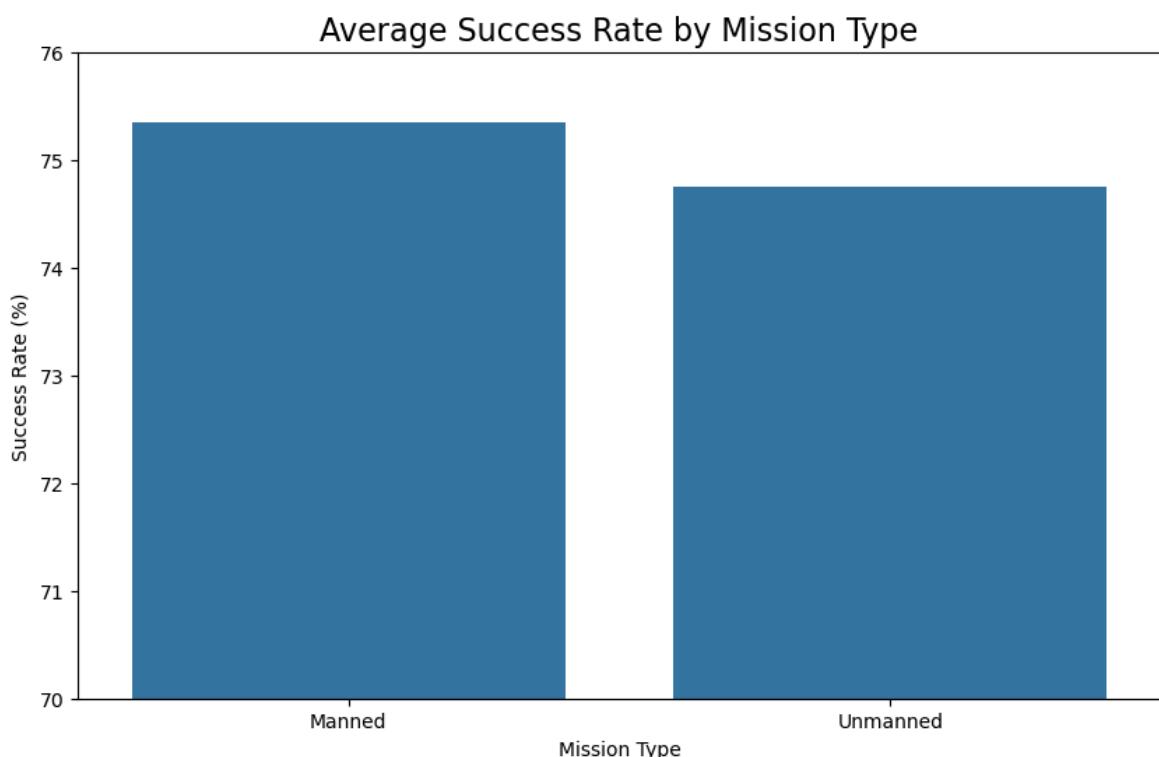
```
In [34]: avg_success_col_country = df.groupby('Collaborating Country')['Success Rate (%)']
```

```
In [35]: plt.figure(figsize=(10,6))
sns.lineplot(data=avg_success_col_country , x='Collaborating Country', y='Success Rate (%)',
plt.title('Average Success Rate by Collaborating Country', fontsize=16)
plt.ylim(70, 78)
plt.show()
```



```
In [36]: avg_success_type = df.groupby('Mission Type')['Success Rate (%)'].mean().reset_index()
```

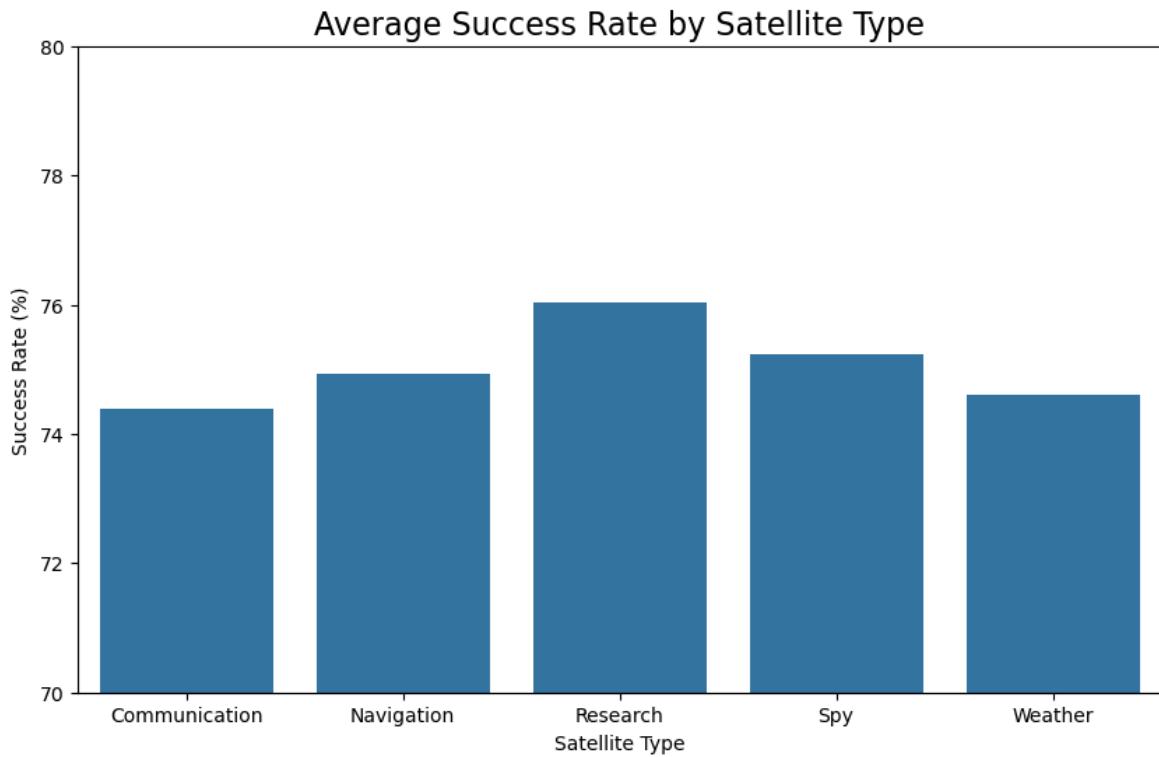
```
In [37]: plt.figure(figsize=(10,6))
sns.barplot(data=avg_success_type , x='Mission Type', y='Success Rate (%)')
plt.title('Average Success Rate by Mission Type', fontsize=16)
plt.ylim(70, 76)
plt.show()
```



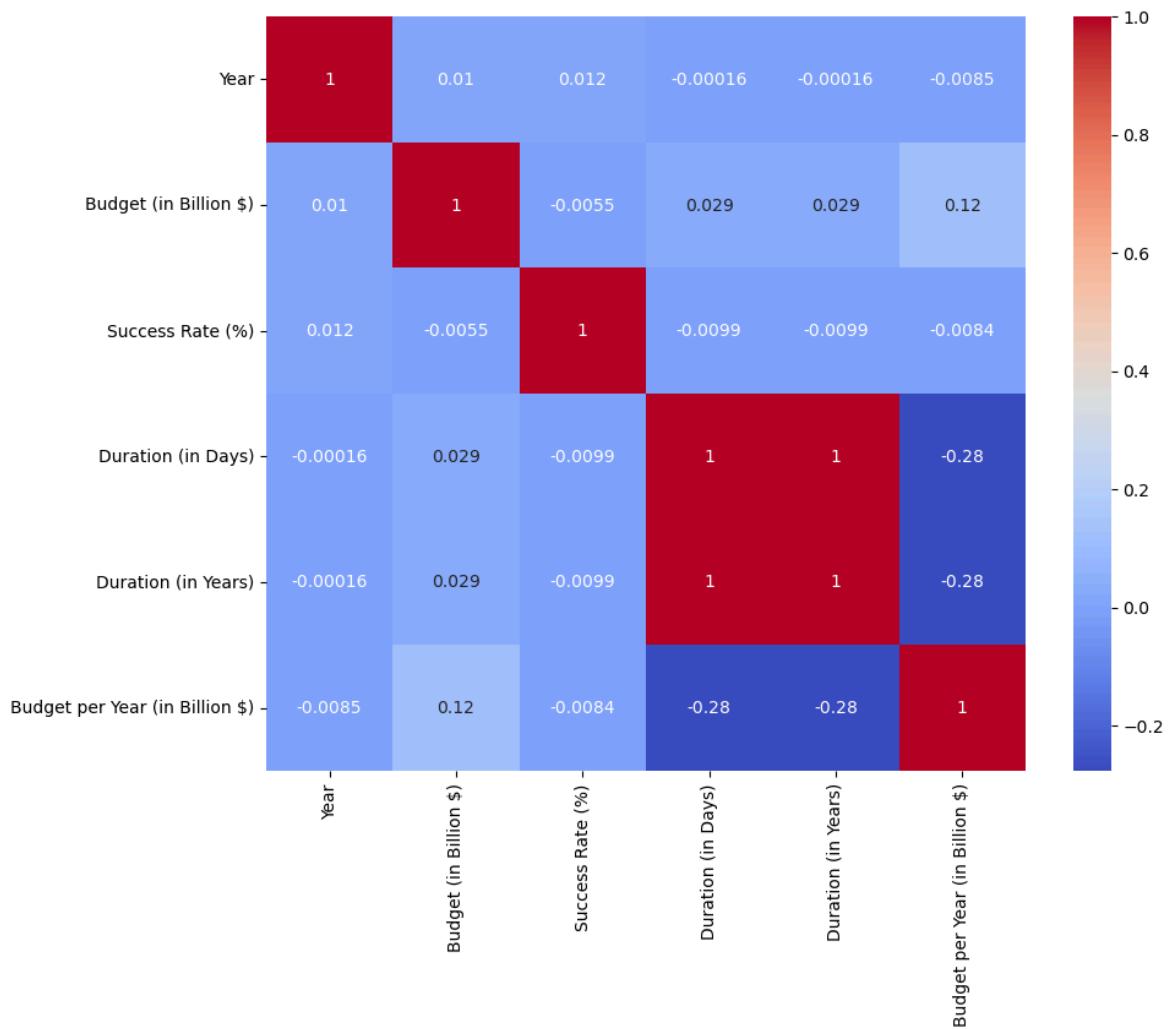
```
In [38]: avg_success_satellite = df.groupby('Satellite Type')['Success Rate (%)'].mean().reset_index()
```

```
In [39]: plt.figure(figsize=(10,6))
sns.barplot(data=avg_success_satellite , x='Satellite Type', y='Success Rate (%)')
```

```
plt.title('Average Success Rate by Satellite Type', fontsize=16)
plt.ylim(70, 80)
plt.show()
```



```
In [40]: numeric_df = df.select_dtypes(include='number')
plt.figure(figsize=(10,8))
sns.heatmap(numeric_df.corr(), annot=True, cmap='coolwarm')
plt.show()
```



In [41]: `df.head()`

Out[41]:

	Main Country	Year	Mission Type	Satellite Type	Budget (in Billion \$)	Success Rate (%)	Technology Used	Environmental Impact
0	China	2008	Manned	Communication	16.20	90	Nuclear Propulsion	Medium
0	China	2008	Manned	Communication	16.20	90	Nuclear Propulsion	Medium
0	China	2008	Manned	Communication	16.20	90	Nuclear Propulsion	Medium
1	Japan	2018	Manned	Communication	29.04	99	Solar Propulsion	High
1	Japan	2018	Manned	Communication	29.04	99	Solar Propulsion	High

## Preprocessing

```
In [42]: print("im going to drop the new columns i made because those are only to make th
```

im going to drop the new columns i made because those are only to make the eda look better and it just adds more filler to the machine

```
In [43]: df = df.drop('Budget per Year (in Billion $)', axis=1)
```

```
In [44]: df = df.drop('Duration (in Years)', axis=1)
```

```
In [45]: df = df.drop('Budget Category', axis=1)
```

```
In [46]: nominal_cols = df[['Main Country','Mission Type','Satellite Type','Technology Us  
ordinal_cols = df[['Environmental Impact']]
```

```
In [47]: from sklearn.preprocessing import OneHotEncoder, OrdinalEncoder
```

```
In [48]: one_encoder = OneHotEncoder(sparse_output=False)  
nominal = one_encoder.fit_transform(nominal_cols)
```

```
In [49]: ohe_f_names = one_encoder.get_feature_names_out(nominal_cols.columns)
```

```
In [50]: ohe_df = pd.DataFrame(nominal,columns=ohe_f_names,index=df.index)
```

```
In [51]: ohe_df.head()
```

Out[51]:

	Main Country_China	Main Country_France	Main Country_Germany	Main Country_India	Main Country_Israel	Co
<b>0</b>	1.0	0.0	0.0	0.0	0.0	0.0
<b>0</b>	1.0	0.0	0.0	0.0	0.0	0.0
<b>0</b>	1.0	0.0	0.0	0.0	0.0	0.0
<b>1</b>	0.0	0.0	0.0	0.0	0.0	0.0
<b>1</b>	0.0	0.0	0.0	0.0	0.0	0.0



```
In [52]: df_encoded = pd.concat([df,ohe_df],axis=1)
```

```
In [53]: df_encoded.head()
```

Out[53]:

	Main Country	Year	Mission Type	Satellite Type	Budget (in Billion \$)	Success Rate (%)	Technology Used	Environmental Impact
0	China	2008	Manned	Communication	16.20	90	Nuclear Propulsion	Medium
0	China	2008	Manned	Communication	16.20	90	Nuclear Propulsion	Medium
0	China	2008	Manned	Communication	16.20	90	Nuclear Propulsion	Medium
1	Japan	2018	Manned	Communication	29.04	99	Solar Propulsion	High
1	Japan	2018	Manned	Communication	29.04	99	Solar Propulsion	High



In [54]:

```
ord_encoder = OrdinalEncoder()
ordinal = ord_encoder.fit_transform(ordinal_cols)
```

In [55]:

```
or_df = pd.DataFrame(ordinal,
                     columns=[col + '_ord' for col in ordinal_cols],
                     index=df.index)
or_df
```

Out[55]:

**Environmental Impact\_ord**

0	2.0
0	2.0
0	2.0
1	0.0
1	0.0
...	...
2997	1.0
2998	1.0
2998	1.0
2998	1.0
2999	2.0

5946 rows × 1 columns

In [56]:

```
df_final = pd.concat([df_encoded, or_df], axis=1)
```

In [57]:

```
df_final.head()
```

Out[57]:

	Main Country	Year	Mission Type	Satellite Type	Budget (in Billion \$)	Success Rate (%)	Technology Used	Environmental Impact
0	China	2008	Manned	Communication	16.20	90	Nuclear Propulsion	Medium
0	China	2008	Manned	Communication	16.20	90	Nuclear Propulsion	Medium
0	China	2008	Manned	Communication	16.20	90	Nuclear Propulsion	Medium
1	Japan	2018	Manned	Communication	29.04	99	Solar Propulsion	High
1	Japan	2018	Manned	Communication	29.04	99	Solar Propulsion	High



In [58]: df2 = df\_final.select\_dtypes(include='number')

In [59]: df2.head()

Out[59]:

	Year	Budget (in Billion \$)	Success Rate (%)	Duration (in Days)	Main Country_China	Main Country_France	Main Country_Germany
0	2008	16.20	90	112	1.0	0.0	0.0
0	2008	16.20	90	112	1.0	0.0	0.0
0	2008	16.20	90	112	1.0	0.0	0.0
1	2018	29.04	99	236	0.0	0.0	0.0
1	2018	29.04	99	236	0.0	0.0	0.0



## Machine Learning (Regression)

In [60]: X = df2.drop(columns=['Success Rate (%)'])  
y = df2['Success Rate (%)']In [61]: from sklearn.model\_selection import train\_test\_split, GridSearchCV  
from sklearn.tree import DecisionTreeClassifier, plot\_tree  
from sklearn.metrics import accuracy\_score, classification\_report, confusion\_matrix  
from sklearn.metrics import ConfusionMatrixDisplay as cmd  
from sklearn.ensemble import AdaBoostRegressor  
from sklearn.ensemble import AdaBoostClassifier  
from sklearn.ensemble import StackingRegressor

In [62]: X\_train,X\_test, y\_train, y\_test = train\_test\_split(X,y,test\_size=0.3,random\_state=42)

In [63]: from sklearn.linear\_model import ElasticNet

```
In [64]: from sklearn.linear_model import LinearRegression, Lasso
from sklearn.ensemble import RandomForestRegressor, GradientBoostingRegressor
from sklearn.metrics import mean_squared_error, mean_absolute_error, r2_score
import joblib
```

```
In [65]: base_model = LinearRegression()
base_model2 = Lasso()
base_model3 = ElasticNet()
rf_model = RandomForestRegressor()
gr_model = GradientBoostingRegressor()
xg_model = xgb.XGBRegressor()
ada_model = AdaBoostRegressor()
```

```
In [66]: estimators = [
    ('rf', RandomForestRegressor(random_state=42)),
    ('gbr', GradientBoostingRegressor(random_state=42)),
    ('xgb', xgb.XGBRegressor(random_state=42))
]

stack_model = StackingRegressor(
    estimators=estimators,
    final_estimator= ElasticNet(),
    n_jobs=-1
)
```

```
In [67]: base_model.fit(X_train, y_train)
base_model2.fit(X_train, y_train)
base_model3.fit(X_train, y_train)
rf_model.fit(X_train, y_train)
gr_model.fit(X_train, y_train)

xg_model.fit(X_train, y_train)
ada_model.fit(X_train, y_train)
```

Out[67]:

- ▼ AdaBoostRegressor [i](#) [?](#)
- ▶ Parameters

```
In [68]: stack_model.fit(X_train, y_train)
```

Out[68]:

- ▶ StackingRegressor [i](#) [?](#)
- rf
- gbr
- ▶ RandomForestRegressor [?](#)
- ▶ GradientBoostingRegressor [?](#) ▶
- final\_estimator
- ▶ ElasticNet [?](#)

```
In [69]: def evaluate_model(model,X_test,y_test,name='Model'):
    y_pred = model.predict(X_test)
    mse = mean_squared_error(y_test, y_pred)
    rmse = np.sqrt(mse)
```

```
mae = mean_absolute_error(y_test,y_pred)
r2 = r2_score(y_test,y_pred)

print(f"--{name}--")
print(f"MAE:{mae:.2f}")
print(f"MSE:{mse:.2f}")
print(f"RSME:{rmse:.2f}")
print(f"R2:{r2:.2f}")
print("-"*20)
return r2
```

In [70]: `evaluate_model(base_model3,X_test,y_test,name='base_model')`

```
--base_model--
MAE:13.23
MSE:228.52
RSME:15.12
R2:-0.00
-----
```

Out[70]: -0.00043158171515034205

In [71]: `evaluate_model(base_model2,X_test,y_test,name='base_model')`

```
--base_model--
MAE:13.23
MSE:228.51
RSME:15.12
R2:-0.00
-----
```

Out[71]: -0.00039739725100518264

In [72]: `evaluate_model(base_model,X_test,y_test,name='base_model')`

```
--base_model--
MAE:13.17
MSE:228.67
RSME:15.12
R2:-0.00
-----
```

Out[72]: -0.0010606418656966543

In [73]: `evaluate_model(rf_model,X_test,y_test,name='rf_model')`

```
--rf_model--
MAE:8.36
MSE:110.90
RSME:10.53
R2:0.51
-----
```

Out[73]: 0.5144859404196307

In [74]: `evaluate_model(gr_model,X_test,y_test,name='gr_model')`

```
--gr_model--
MAE:12.69
MSE:213.67
RSME:14.62
R2:0.06
-----
```

```
Out[74]: 0.06458110747783996
```

```
In [75]: evaluate_model(xg_model,X_test,y_test,name='xg_model')
```

```
--xg_model--  
MAE:8.38  
MSE:125.64  
RSME:11.21  
R2:0.45  
-----
```

```
Out[75]: 0.44996678829193115
```

```
In [76]: evaluate_model(ada_model,X_test,y_test,name='ada_model')
```

```
--ada_model--  
MAE:13.20  
MSE:228.53  
RSME:15.12  
R2:-0.00  
-----
```

```
Out[76]: -0.0004876585161428526
```

```
In [77]: evaluate_model(stack_model,X_test,y_test,name='stack_model')
```

```
--stack_model--  
MAE:6.79  
MSE:95.80  
RSME:9.79  
R2:0.58  
-----
```

```
Out[77]: 0.5805908223144238
```

```
In [78]: print("rf_model is the best model, we'll use that then")
```

```
rf_model is the best model, we'll use that then
```

```
In [79]: model = stack_model
```

```
In [80]: y_pred = model.predict(X_test)  
y_pred
```

```
Out[80]: array([50.36508188, 70.4953313 , 68.2512062 , ..., 55.39049312,  
 73.01738208, 77.37253188])
```

```
In [81]: param_grid = {  
    'rf_n_estimators': [100, 200],  
    'rf_max_depth': [None, 10],  
    'gbr_n_estimators': [100, 200],  
    'gbr_learning_rate': [0.05, 0.1],  
    'xgb_n_estimators': [100, 200],  
    'xgb_max_depth': [3, 5],  
}  
  
grid_search = GridSearchCV(estimator=model, param_grid=param_grid, cv=5, n_jobs=-1)  
grid_search.fit(X_train, y_train)  
  
best_model = grid_search.best_estimator_  
evaluate_model(best_model,X_test,y_test,name='grid_model')
```

```
Fitting 5 folds for each of 64 candidates, totalling 320 fits
--grid_model--
MAE:6.56
MSE:91.63
RSME:9.57
R2:0.60
-----
```

Out[81]: 0.5988585577808523

In [82]: `evaluate_model(best_model,X_test,y_test,name='stack')`

```
--stack--
MAE:6.56
MSE:91.63
RSME:9.57
R2:0.60
-----
```

Out[82]: 0.5988585577808523

## Deep learning (alternate path)

In [83]: `# training data  
X_train_full, X_test, y_train_full, y_test = train_test_split(X,y,test_size=0.2,  
  
# validation data  
X_train, X_val, y_train, y_val = train_test_split(X_train_full,y_train_full,test`

In [84]: `from sklearn.preprocessing import StandardScaler  
scaler = StandardScaler()  
X_train_scaled = scaler.fit_transform(X_train)  
X_val_scaled = scaler.transform(X_val)  
X_test_scaled = scaler.transform(X_test)`

In [85]: `from sklearn.neural_network import MLPRegressor`

In [86]: `# regr = MLPRegressor(random_state=1, max_iter=2000, tol=0.1)  
# regr.fit(X_train, y_train)  
# MLPRegressor(max_iter=10000, random_state=1, tol=0.1)  
# regr.predict(X_test[:2])  
# # array([ 28.98, -291])  
# regr.score(X_test, y_test)`

In [87]: `from tensorflow.keras.models import Sequential  
from tensorflow.keras.layers import Dense, BatchNormalization, Dropout, Activation  
from tensorflow.keras.layers import Dense, Input  
  
model = Sequential([  
 Input(shape=(X_train_scaled.shape[1],)),  
 # Dense(512, activation='relu'),  
 # Dense(64, activation='relu'),  
 # Dense(128, activation='relu'),  
 # # Dense(32, activation='relu'),  
 # Dense(256, activation='relu'),  
 # # Dense(32, activation='relu'),  
 # Dense(32, activation='relu'),  
 # Dense(32, activation='relu'),  
 Dense(32, activation='relu'),`

```
Dense(32, activation='relu'),
Dense(1)
])
model.summary()
```

C:\Users\Predator\anaconda3\envs\tensorflow\_env\Lib\site-packages\keras\src\export\tf2onnx\_lib.py:8: FutureWarning: In the future `np.object` will be defined as the corresponding NumPy scalar.  
if not hasattr(np, "object"):  
Model: "sequential"

Layer (type)	Output Shape
dense (Dense)	(None, 32)
dense_1 (Dense)	(None, 32)
dense_2 (Dense)	(None, 1)

Total params: 2,273 (8.88 KB)  
Trainable params: 2,273 (8.88 KB)  
Non-trainable params: 0 (0.00 B)

In [88]: # model.compile(  
# loss='mean\_squared\_error',  
# optimizer='adam',  
# metrics=['mean\_absolute\_error'])  
# )  
  
# history = model.fit(  
# X\_train\_scaled, y\_train,  
# epochs=100,  
# validation\_data=(X\_val\_scaled, y\_val),  
# batch\_size=32,  
# verbose=1  
# )

In [89]: # plt.plot(history.history['val\_loss'], label='validation loss')  
# plt.plot(history.history['Loss'], label='training loss')  
# plt.legend()  
# plt.show()

In [90]: # y\_pred = model.predict(X\_test)  
  
# y\_pred = y\_pred.ravel()

In [91]: # from sklearn.metrics import mean\_absolute\_error, mean\_squared\_error, r2\_score  
  
# evaluate\_model(model, X\_test, y\_test, name='deep')

## Evaluate

In [92]: best\_model = grid\_search.best\_estimator\_  
evaluate\_model(best\_model, X\_test, y\_test, name='grid\_model')

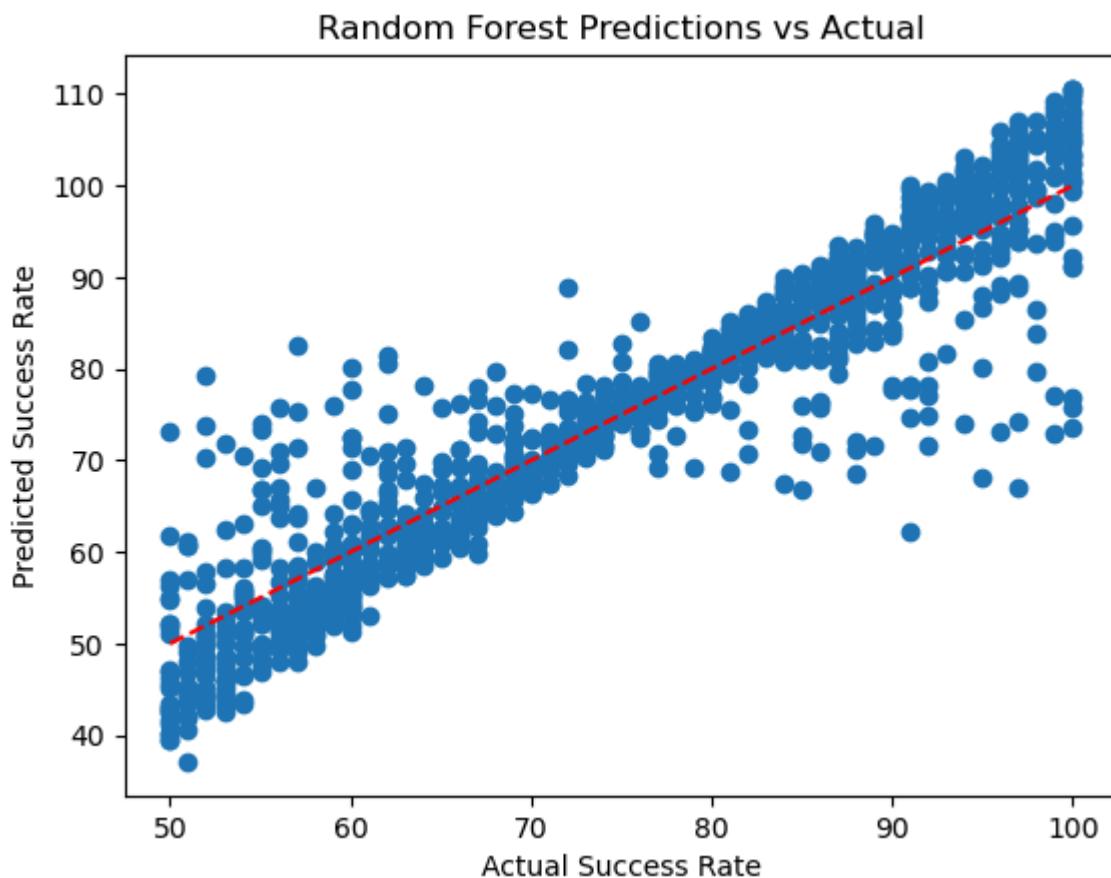
```
--grid_model--  
MAE:4.23  
MSE:35.88  
RSME:5.99  
R2:0.84  
-----
```

```
Out[92]: 0.8401982770579272
```

```
In [93]: y_pred = best_model.predict(X_test)  
y_pred
```

```
Out[93]: array([78.96866928, 73.17602993, 77.04699092, ..., 81.13129435,  
69.68906884, 51.0381872 ])
```

```
In [94]: import matplotlib.pyplot as plt  
  
plt.scatter(y_test, y_pred)  
plt.plot([y_test.min(), y_test.max()], [y_test.min(), y_test.max()], 'r--')  
plt.xlabel('Actual Success Rate')  
plt.ylabel('Predicted Success Rate')  
plt.title('Random Forest Predictions vs Actual')  
plt.show()
```



## Saving The Model

```
In [95]: joblib.dump(best_model, 'regression.joblib')
```

```
Out[95]: ['regression.joblib']
```

```
In [96]: joblib.dump(best_model, 'regression.joblib')
```

```
Out[96]: ['regression.joblib']
```

```
In [97]: joblib.dump(one_encoder, 'one_encoder.joblib')  
joblib.dump(ord_encoder, 'ord_encoder.joblib')
```

```
Out[97]: ['ord_encoder.joblib']
```

## Gradio App

```
In [1]: pip install gradio
```

```
Collecting gradio
  Downloading gradio-6.2.0-py3-none-any.whl.metadata (16 kB)
Collecting aiofiles<25.0,>=22.0 (from gradio)
  Using cached aiofiles-24.1.0-py3-none-any.whl.metadata (10 kB)
Requirement already satisfied: aiofiles<25.0,>=22.0 in c:\users\predator\anaconda3\envs\tensorflow_env\lib\site-packages (from gradio) (4.12.0)
Requirement already satisfied: brotli>=1.1.0 in c:\users\predator\anaconda3\envs\tensorflow_env\lib\site-packages (from gradio) (1.2.0)
Collecting fastapi<1.0,>=0.115.2 (from gradio)
  Downloading fastapi-0.128.0-py3-none-any.whl.metadata (30 kB)
Collecting ffcmpy (from gradio)
  Using cached ffcmpy-1.0.0-py3-none-any.whl.metadata (3.0 kB)
Collecting gradio-client==2.0.2 (from gradio)
  Downloading gradio_client-2.0.2-py3-none-any.whl.metadata (7.1 kB)
Collecting groovy~=0.1 (from gradio)
  Using cached groovy-0.1.2-py3-none-any.whl.metadata (6.1 kB)
Requirement already satisfied: httpx<1.0,>=0.24.1 in c:\users\predator\anaconda3\envs\tensorflow_env\lib\site-packages (from gradio) (0.28.1)
Collecting huggingface-hub<2.0,>=0.33.5 (from gradio)
  Downloading huggingface_hub-1.2.4-py3-none-any.whl.metadata (13 kB)
Requirement already satisfied: jinja2<4.0 in c:\users\predator\anaconda3\envs\tensorflow_env\lib\site-packages (from gradio) (3.1.6)
Requirement already satisfied: markupsafe<4.0,>=2.0 in c:\users\predator\anaconda3\envs\tensorflow_env\lib\site-packages (from gradio) (3.0.3)
Requirement already satisfied: numpy<3.0,>=1.0 in c:\users\predator\anaconda3\envs\tensorflow_env\lib\site-packages (from gradio) (1.26.4)
Collecting orjson~=3.0 (from gradio)
  Using cached orjson-3.11.5-cp312-cp312-win_amd64.whl.metadata (42 kB)
Requirement already satisfied: packaging in c:\users\predator\anaconda3\envs\tensorflow_env\lib\site-packages (from gradio) (25.0)
Requirement already satisfied: pandas<3.0,>=1.0 in c:\users\predator\anaconda3\envs\tensorflow_env\lib\site-packages (from gradio) (2.3.3)
Requirement already satisfied: pillow<13.0,>=8.0 in c:\users\predator\anaconda3\envs\tensorflow_env\lib\site-packages (from gradio) (12.0.0)
Collecting pydantic<=3.0,>=2.0 (from gradio)
  Downloading pydantic-2.12.5-py3-none-any.whl.metadata (90 kB)
Collecting pydub (from gradio)
  Using cached pydub-0.25.1-py2.py3-none-any.whl.metadata (1.4 kB)
Collecting python-multipart>=0.0.18 (from gradio)
  Downloading python_multipart-0.0.21-py3-none-any.whl.metadata (1.8 kB)
Requirement already satisfied: pyyaml<7.0,>=5.0 in c:\users\predator\anaconda3\envs\tensorflow_env\lib\site-packages (from gradio) (6.0.3)
Collecting safehttpx<0.2.0,>=0.1.7 (from gradio)
  Using cached safehttpx-0.1.7-py3-none-any.whl.metadata (4.2 kB)
Collecting semantic-version~=2.0 (from gradio)
  Using cached semantic_version-2.10.0-py2.py3-none-any.whl.metadata (9.7 kB)
Collecting starlette<1.0,>=0.40.0 (from gradio)
  Using cached starlette-0.50.0-py3-none-any.whl.metadata (6.3 kB)
Collecting tomlkit<0.14.0,>=0.12.0 (from gradio)
  Using cached tomlkit-0.13.3-py3-none-any.whl.metadata (2.8 kB)
Collecting typer<1.0,>=0.12 (from gradio)
  Downloading typer-0.21.1-py3-none-any.whl.metadata (16 kB)
Requirement already satisfied: typing-extensions~=4.0 in c:\users\predator\anaconda3\envs\tensorflow_env\lib\site-packages (from gradio) (4.15.0)
Collecting uvicorn>=0.14.0 (from gradio)
  Downloading uvicorn-0.40.0-py3-none-any.whl.metadata (6.7 kB)
Collecting fsspec (from gradio-client==2.0.2->gradio)
  Downloading fsspec-2025.12.0-py3-none-any.whl.metadata (10 kB)
Requirement already satisfied: idna>=2.8 in c:\users\predator\anaconda3\envs\tensorflow_env\lib\site-packages (from anyio<5.0,>=3.0->gradio) (3.11)
```

```
Collecting annotated-doc>=0.0.2 (from fastapi<1.0,>=0.115.2->gradio)
  Using cached annotated_doc-0.0.4-py3-none-any.whl.metadata (6.6 kB)
Requirement already satisfied: certifi in c:\users\predator\anaconda3\envs\tensorflow_env\lib\site-packages (from httpx<1.0,>=0.24.1->gradio) (2025.11.12)
Requirement already satisfied: httpcore==1.* in c:\users\predator\anaconda3\envs\tensorflow_env\lib\site-packages (from httpx<1.0,>=0.24.1->gradio) (1.0.9)
Requirement already satisfied: h11>=0.16 in c:\users\predator\anaconda3\envs\tensorflow_env\lib\site-packages (from httpcore==1.*->httpx<1.0,>=0.24.1->gradio) (0.16.0)
Collecting filelock (from huggingface-hub<2.0,>=0.33.5->gradio)
  Downloading filelock-3.20.2-py3-none-any.whl.metadata (2.1 kB)
Collecting hf-xet<2.0.0,>=1.2.0 (from huggingface-hub<2.0,>=0.33.5->gradio)
  Using cached hf_xet-1.2.0-cp37-abi3-win_amd64.whl.metadata (5.0 kB)
Collecting shellingham (from huggingface-hub<2.0,>=0.33.5->gradio)
  Downloading shellingham-1.5.4-py2.py3-none-any.whl.metadata (3.5 kB)
Collecting tqdm>=4.42.1 (from huggingface-hub<2.0,>=0.33.5->gradio)
  Using cached tqdm-4.67.1-py3-none-any.whl.metadata (57 kB)
Collecting typer-slim (from huggingface-hub<2.0,>=0.33.5->gradio)
  Downloading typer_slim-0.21.1-py3-none-any.whl.metadata (16 kB)
Requirement already satisfied: python-dateutil>=2.8.2 in c:\users\predator\anaconda3\envs\tensorflow_env\lib\site-packages (from pandas<3.0,>=1.0->gradio) (2.9.0.post0)
Requirement already satisfied: pytz>=2020.1 in c:\users\predator\anaconda3\envs\tensorflow_env\lib\site-packages (from pandas<3.0,>=1.0->gradio) (2025.2)
Requirement already satisfied: tzdata>=2022.7 in c:\users\predator\anaconda3\envs\tensorflow_env\lib\site-packages (from pandas<3.0,>=1.0->gradio) (2025.3)
Collecting annotated-types>=0.6.0 (from pydantic<=3.0,>=2.0->gradio)
  Downloading annotated_types-0.7.0-py3-none-any.whl.metadata (15 kB)
Collecting pydantic-core==2.41.5 (from pydantic<=3.0,>=2.0->gradio)
  Downloading pydantic_core-2.41.5-cp312-cp312-win_amd64.whl.metadata (7.4 kB)
Collecting typing-inspection>=0.4.2 (from pydantic<=3.0,>=2.0->gradio)
  Downloading typing_inspection-0.4.2-py3-none-any.whl.metadata (2.6 kB)
Collecting click>=8.0.0 (from typer<1.0,>=0.12->gradio)
  Using cached click-8.3.1-py3-none-any.whl.metadata (2.6 kB)
Requirement already satisfied: rich>=10.11.0 in c:\users\predator\anaconda3\envs\tensorflow_env\lib\site-packages (from typer<1.0,>=0.12->gradio) (14.2.0)
Requirement already satisfied: colorama in c:\users\predator\anaconda3\envs\tensorflow_env\lib\site-packages (from click>=8.0.0->typer<1.0,>=0.12->gradio) (0.4.6)
Requirement already satisfied: six>=1.5 in c:\users\predator\anaconda3\envs\tensorflow_env\lib\site-packages (from python-dateutil>=2.8.2->pandas<3.0,>=1.0->gradio) (1.17.0)
Requirement already satisfied: markdown-it-py>=2.2.0 in c:\users\predator\anaconda3\envs\tensorflow_env\lib\site-packages (from rich>=10.11.0->typer<1.0,>=0.12->gradio) (4.0.0)
Requirement already satisfied: pygments<3.0.0,>=2.13.0 in c:\users\predator\anaconda3\envs\tensorflow_env\lib\site-packages (from rich>=10.11.0->typer<1.0,>=0.12->gradio) (2.19.2)
Requirement already satisfied: mdurl~0.1 in c:\users\predator\anaconda3\envs\tensorflow_env\lib\site-packages (from markdown-it-py>=2.2.0->rich>=10.11.0->typer<1.0,>=0.12->gradio) (0.1.2)
Downloading gradio-6.2.0-py3-none-any.whl (23.0 MB)
----- 0.0/23.0 MB ? eta ------
----- 2.4/23.0 MB 19.2 MB/s eta 0:00:02
----- 2.6/23.0 MB 10.8 MB/s eta 0:00:02
----- 3.4/23.0 MB 5.4 MB/s eta 0:00:04
----- 3.9/23.0 MB 5.0 MB/s eta 0:00:04
----- 4.2/23.0 MB 4.4 MB/s eta 0:00:05
----- 5.0/23.0 MB 4.0 MB/s eta 0:00:05
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----- 6.0/23.0 MB 3.5 MB/s eta 0:00:05
```

```
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----- 20.2/23.0 MB 2.7 MB/s eta 0:00:02
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----- 21.5/23.0 MB 2.6 MB/s eta 0:00:01
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----- 22.5/23.0 MB 2.6 MB/s eta 0:00:01
----- 23.0/23.0 MB 2.6 MB/s 0:00:08

Downloading gradio_client-2.0.2-py3-none-any.whl (55 kB)
Using cached aiofiles-24.1.0-py3-none-any.whl (15 kB)
Downloading fastapi-0.128.0-py3-none-any.whl (103 kB)
Using cached groovy-0.1.2-py3-none-any.whl (14 kB)
Downloading huggingface_hub-1.2.4-py3-none-any.whl (520 kB)
Using cached hf_xet-1.2.0-cp37-abi3-win_amd64.whl (2.9 MB)
Using cached orjson-3.11.5-cp312-cp312-win_amd64.whl (133 kB)
Downloading pydantic-2.12.5-py3-none-any.whl (463 kB)
Downloading pydantic_core-2.41.5-cp312-cp312-win_amd64.whl (2.0 MB)
----- 0.0/2.0 MB ? eta -:::-
----- 0.5/2.0 MB 2.8 MB/s eta 0:00:01
----- 1.0/2.0 MB 2.2 MB/s eta 0:00:01
----- 1.6/2.0 MB 2.3 MB/s eta 0:00:01
----- 1.8/2.0 MB 2.3 MB/s eta 0:00:01
----- 2.0/2.0 MB 2.2 MB/s 0:00:00

Using cached safehttpx-0.1.7-py3-none-any.whl (9.0 kB)
Using cached semantic_version-2.10.0-py2.py3-none-any.whl (15 kB)
Using cached starlette-0.50.0-py3-none-any.whl (74 kB)
Using cached tomlkit-0.13.3-py3-none-any.whl (38 kB)
Downloading typer-0.21.1-py3-none-any.whl (47 kB)
Using cached annotated_doc-0.0.4-py3-none-any.whl (5.3 kB)
Downloading annotated_types-0.7.0-py3-none-any.whl (13 kB)
Using cached click-8.3.1-py3-none-any.whl (108 kB)
Downloading fsspec-2025.12.0-py3-none-any.whl (201 kB)
Downloading python_multipart-0.0.21-py3-none-any.whl (24 kB)
Downloading shellingham-1.5.4-py2.py3-none-any.whl (9.8 kB)
Using cached tqdm-4.67.1-py3-none-any.whl (78 kB)
```

```
Downloading typing_inspection-0.4.2-py3-none-any.whl (14 kB)
Downloading uvicorn-0.40.0-py3-none-any.whl (68 kB)
Using cached ffmpeg-1.0.0-py3-none-any.whl (5.6 kB)
Downloading filelock-3.20.2-py3-none-any.whl (16 kB)
Using cached pydub-0.25.1-py2.py3-none-any.whl (32 kB)
Downloading typer_slim-0.21.1-py3-none-any.whl (47 kB)
Installing collected packages: pydub, typing-inspection, tqdm, tomlkit, shellingham, semantic-version, python-multipart, pydantic-core, orjson, hf-xet, groovy, fsspec, filelock, ffmpeg, click, annotated-types, annotated-doc, aiofiles, uvicorn, typer-slim, starlette, pydantic, typer, safehttpx, huggingface-hub, fastapi, gradio-client, gradio
```

```
- 1/28 [typing-inspection]
-- 2/28 [tqdm]
--- 2/28 [tqdm]
---- 3/28 [tomlkit]
----- 5/28 [semantic-version]
----- 7/28 [pydantic-core]
----- 7/28 [pydantic-core]
----- 9/28 [hf-xet]
----- 9/28 [hf-xet]
----- 11/28 [fsspec]
----- 11/28 [fsspec]
----- 11/28 [fsspec]
----- 11/28 [fsspec]
----- 12/28 [filelock]
----- 14/28 [click]
----- 14/28 [click]
----- 17/28 [aiofiles]
----- 18/28 [uvicorn]
----- 18/28 [uvicorn]
----- 19/28 [typer-slim]
----- 20/28 [starlette]
----- 20/28 [starlette]
----- 21/28 [pydantic]
----- 22/28 [typer]
----- 24/28 [huggingface-hub]
----- 25/28 [fastapi]
----- 25/28 [fastapi]
----- 25/28 [fastapi]
----- 26/28 [gradio-client]
----- 27/28 [gradio]
----- 27/28 [gradio]
----- 27/28 [gradio]
----- 27/28 [gradio]
```





```
- 27/28 [gradio]
----- - 27/28 [gradio]
```

Successfully installed aiofiles-24.1.0 annotated-doc-0.0.4 annotated-types-0.7.0 click-8.3.1 fastapi-0.128.0 ffmpeg-1.0.0 filelock-3.20.2 fsspec-2025.12.0 gradio-6.2.0 gradio-client-2.0.2 groovy-0.1.2 hf-xet-1.2.0 huggingface-hub-1.2.4 orjson-3.11.5 pydantic-2.12.5 pydantic-core-2.41.5 pydub-0.25.1 python-multipart-0.0.21 safehttpx-0.1.7 semantic-version-2.10.0 shellingham-1.5.4 starlette-0.50.0 tomlkit-0.13.3 tqdm-4.67.1 typer-0.21.1 typer-slim-0.21.1 typing-inspection-0.4.2 uvicorn-0.40.0

Note: you may need to restart the kernel to use updated packages.

```
In [2]: import gradio as gr
import joblib
```

```
In [3]: rf_model = joblib.load('regression.joblib')
one_encoder = joblib.load('one_encoder.joblib')
ord_encoder = joblib.load('ord_encoder.joblib')
```

```
In [4]: country_choices = ['UK', 'China', 'France', 'Israel', 'USA', 'UAE', 'India', 'Japan']
mission_type_choices = ['Unmanned', 'Manned']
satellite_type_choices = ['Research', 'Weather', 'Communication', 'Navigation', 'Military']
technology_choices = ['Traditional Rocket', 'Solar Propulsion', 'AI Navigation', 'Quantum Computing']
collab_country_choices = ['Germany', 'USA', 'Russia', 'Japan', 'UAE', 'Israel', 'Kuwait']
env_choices = ['Low', 'Medium', 'High']

nominal_cols = ['Main Country', 'Mission Type', 'Satellite Type', 'Technology Used']
ordinal_cols = ['Environmental Impact']
numeric_cols = ['Year', 'Budget (in Billion $)', 'Duration (in Days)']
```

```
# AI GENERATE STARTS FROM HERE BEEP BOOP
def predict_mission(Main_Country, Year, Mission_Type, Satellite_Type, Budget, Technology_Used, Environmental_Impact, Collaborating_Country, Duration_Days):

    # Create single-row DataFrame with numeric + single-valued nominal/ordinal columns
    user_df = pd.DataFrame({
        'Main Country': [Main_Country],
        'Year': [Year],
        'Mission Type': [Mission_Type],
        'Satellite Type': [Satellite_Type],
        'Budget (in Billion $)': [Budget],
        'Technology Used': [Technology_Used],
        'Environmental Impact': [Environmental_Impact],
        'Collaborating Country': [Collaborating_Country[0]], # temporary placeholder
        'Duration (in Days)': [Duration_Days]
    })

    # Handle multiple collaborating countries by duplicating the row for each selected country
    collab_rows = []
```

```

for country in Collaborating_Country:
    row_copy = user_df.copy()
    row_copy[ 'Collaborating Country' ] = country
    collab_rows.append(row_copy)

multi_collab_df = pd.concat(collab_rows, axis=0)

# Transform categorical columns
nominal_encoded = one_encoder.transform(multi_collab_df[nominal_cols])
ordinal_encoded = ord_encoder.transform(multi_collab_df[ordinal_cols])

# Extract numeric values
numeric = multi_collab_df[numeric_cols].values

# Combine all features
final_input = np.hstack([numeric, nominal_encoded, ordinal_encoded])

# Predict and average if multiple collaborators
predictions = rf_model.predict(final_input)
prediction = predictions.mean()

return round(prediction, 2)

# AI GENERATION ENDS HERE BEEP BOOP

inputs = [
    gr.Dropdown(label='Main Country', choices=country_choices),
    gr.Number(label='Year'),
    gr.Dropdown(label='Mission Type', choices=mission_type_choices),
    gr.Dropdown(label='Satellite Type', choices=satellite_type_choices),
    gr.Number(label='Budget (in Billion $)'),
    gr.Dropdown(label='Technology Used', choices=technology_choices),
    gr.Dropdown(label='Environmental Impact', choices=env_choices),
    gr.CheckboxGroup(label='Collaborating Country', choices=collab_country_choices),
    gr.Number(label='Duration (in Days)'),
]

outputs = gr.Number(label='Predicted Success Rate (%)')

app = gr.Interface(fn=predict_mission, inputs=inputs, outputs=outputs, title='Global Space Mission Predictor')
app.launch(share=True)

```

\* Running on local URL: <http://127.0.0.1:7860>

Could not create share link. Please check your internet connection or our status page: <https://status.gradio.app>.

Out[4]:

In [ ]: