Bring in Data and Import Libraries

```
!pip install nba_api
!pip install geopy
!pip install folium
!pip install scikit-learn
import folium
import pandas as pd

from sklearn.feature_selection import SelectKBest
from sklearn.feature_selection import chi2, f_regression, f_classif, mutual_info_regression, mutual_info_classif
from sklearn.preprocessing import MinMaxScaler, StandardScaler, OneHotEncoder, LabelEncoder
from sklearn.model_selection import train_test_split, GridSearchCV, RandomizedSearchCV
from sklearn.linear_model import LogisticRegression, LinearRegression, Ridge, Lasso, ElasticNet
from sklearn.metrics import accuracy_score, precision_score, recall_score, f1_score, confusion_matrix, classification_report
from sklearn.tree import DecisionTreeClassifier, DecisionTreeRegressor
from sklearn.ensemble import RandomForestClassifier, RandomForestRegressor
from sklearn.naive_bayes import GaussianNB, MultinomialNB, BernoulliNB
from sklearn.neighbors import KNeighborsClassifier, KNeighborsRegressor
from sklearn.cluster import KMeans
from sklearn.decomposition import PCA
from sklearn.pipeline import Pipeline
from sklearn.impute import SimpleImputer
from sklearn.compose import ColumnTransformer
from sklearn.base import BaseEstimator, TransformerMixin
from sklearn.svm import SVC
```

```
# If the nba_api doesn't work properly, run the below commands to update your current version and remove cache.dbf
!pip install nba_api --upgrade
!rm /usr/local/lib/python3.10/dist-packages/nba_api/stats/library/data/cache.dbf
```

```
# If you get a 'ConnectionResetError' run this cell and then restart the runtime (Runtime > Restart runtime)
!pip install nba_api --upgrade
```

Preprocessing

3 Essential Preprocessing Techniques that are:

- Encodings for moving data
- Feature Selection techniques for categorical features
- Normalization or Standardization for scaled features
- Dimensionality Reduction techniques
- Imputation techniques for missing data

3 Model Evaluation Techniques that are:

- Regression Metrics
- Classification Metrics
- Clustering Metrics

Clean and Split Columns into Lists for the MultiLabelBinarizer

```
def split_and_clean_df_columns(df):
    """
    Inputs a df and splits all columns
    Returns a df with the split columns and the original columns removed
    Also returns a list of the original columns that were split
    Also returns a list of the new columns that were created
    Also returns a dictionary, where the keys are the original columns and the values are lists of the new columns created from splitting the original column
    """
    original_columns = df.columns.tolist()
    split_columns = []
    new_columns = []
    new_columns_dict = {}
    for col in df.columns:
        if df[col].dtype == 'object':
            if df[col].str.contains(',').any():
                split_df = df[col].str.split(',', expand=True).add_prefix(col + '_')
                df = pd.concat([df, split_df], axis=1)
                df.drop(col, axis=1, inplace=True)
                split_columns.append(col)
                new_columns.extend(split_df.columns.tolist())
                new_columns_dict[col] = split_df.columns.tolist()
    return df, original_columns, split_columns, new_columns, new_columns_dict

df, og_cols, split_cols, new_cols, new_cols_dict = split_and_clean_df_columns(df)
```

OneHotEncode and MultiLabelBinarize

```
class TrainingOneHotEncoderTransformer(BaseEstimator, TransformerMixin):
    def __init__(self, columns=None):
        self.columns = columns
        self.encoders = {}

    def fit(self, X, y=None):
        if self.columns is None:
            self.columns = X.columns
        for col in self.columns:
            encoder = OneHotEncoder(sparse=False, handle_unknown='ignore')
            encoder.fit(X[[col]])
            self.encoders[col] = encoder
        return self

    def transform(self, X):
        X = X.copy()
        for col, encoder in self.encoders.items():
            encoded = encoder.transform(X[[col]])
            encoded_df = pd.DataFrame(encoded, columns=[f"{col}_{i}" for i in range(encoded.shape[1])])
            X = pd.concat([X.drop(col, axis=1), encoded_df], axis=1)
        return X

class TrainingMultiLabelBinarizerTransformer(BaseEstimator, TransformerMixin):
    def __init__(self, columns=None):
        self.columns = columns
        self.binarizers = {}

    def fit(self, X, y=None):
        if self.columns is None:
            self.columns = X.columns
        for col in self.columns:
            binarizer = MultiLabelBinarizer()
            binarizer.fit(X[col])
            self.binarizers[col] = binarizer
        return self

    def transform(self, X):
        X = X.copy()
        for col, binarizer in self.binarizers.items():
            encoded = binarizer.transform(X[col])
            encoded_df = pd.DataFrame(encoded, columns=[f"{col}_{i}" for i in range(encoded.shape[1])])
            X = pd.concat([X.drop(col, axis=1), encoded_df], axis=1)
        return X
```

Pipeline

```
cols_binary = ['Home_Game',
               'Target_Game_Win',
               'Target_Game_Loss',
               'Target_Game_Tie']

# Make sure all data types are encoded a format that can be string
df[cols_binary] = df[cols_binary].astype('str')

cols_binary = ['Category_0', 'Category_1', 'Category_2']
```

```
# Numerical Features
cols_numerical = ['Additional_Assists_Count', 'Points']
```

```
# Multicollinearity Check
corr_matrix = df[cols_numerical].corr()
corr_matrix
```

Explore Data

- **Full Statistics Data Analysis (EA etc)**
  - Expected Value (probability statistics) or statistical tests like chi2
  - Correlation Matrix and Plots for Variability
- **Distribution summary of:**
  - Plot for discrete variables
  - Target variable / Feature Input
  - Numerical Features (Mean, Median, Standard Deviation)

```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1048575 entries, 0 to 1048574
Data columns (total 19 columns):
 #   Column                  Non-Null Count    Dtype
---  ------                  --------------    -----
 0   Player_ID               1048575 non-null  object
 1   Player_Name             1048575 non-null  object
 2   Game_ID                 1048575 non-null  object
 3   Game_Date               1048575 non-null  object
 4   Points                  1048575 non-null  object
 5   Assists_Count           1048575 non-null  object
 6   Additional_Assists_Count  1048575 non-null  object
 7   Rebounds_Count          1048575 non-null  object
 8   Home_Game               1048575 non-null  object
 9   Target_Game_Win         1048575 non-null  object
 10  Target_Game_Loss        1048575 non-null  object
 11  Target_Game_Tie         1048575 non-null  object
 12  Category_0              1048575 non-null  object
 13  Category_1              1048575 non-null  object
 14  Category_2              1048575 non-null  object
 15  Category_3              1048575 non-null  object
 16  Category_4              1048575 non-null  object
 17  Category_5              1048575 non-null  object
 18  Category_6              1048575 non-null  object
dtypes: object(19)
memory usage: 152.0+ MB
```

```
df.describe(include='all').T
```

	count	unique	top	freq
Category	1048575	7	Lower Tower	181
Player_Name	1048575	1048575	Jaylen Brown	1
Universal_ID	0	0	Universal Exception	0
Total_Data_Points	19	19456	20241003	54
Player_Name_Count	19	19456	20241003	54
NFPA_Technique_Count	19456	523	2719302	1048
Identifying_Attribute_Count	19456	19456	Player_Identity_Attribute	1
Player_Name	19456	19456	Player_Identity_Attribute	1
Universal_Data_Point	19456	19456	Universal_Data_Point	1

```
# Print the number of unique values in each column
for column in df.columns:
    print(f"{column}: {df[column].nunique()} unique values")
```

```
Player_ID: 1048575 unique values
Player_Name: 1048575 unique values
Game_ID: 1048575 unique values
Game_Date: 1048575 unique values
Points: 1048575 unique values
Assists_Count: 1048575 unique values
Additional_Assists_Count: 1048575 unique values
Rebounds_Count: 1048575 unique values
Home_Game: 1048575 unique values
Target_Game_Win: 1048575 unique values
Target_Game_Loss: 1048575 unique values
Target_Game_Tie: 1048575 unique values
Category_0: 1048575 unique values
Category_1: 1048575 unique values
Category_2: 1048575 unique values
Category_3: 1048575 unique values
Category_4: 1048575 unique values
Category_5: 1048575 unique values
Category_6: 1048575 unique values
```