

Pre-processing Data

```
#Importing Libraries for pre processing
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.compose import ColumnTransformer
from sklearn.pipeline import Pipeline
from sklearn.impute import SimpleImputer
from sklearn.preprocessing import OneHotEncoder, StandardScaler

#for figures
fWidth = 5
fHeight = 12

df = pd.read_csv("vgsales.csv")
#shape of the dataset
df.shape
df
```

	Rank	Name	Platform	Year	Genre	Publisher	NA_Sales	EU_Sales	JP_Sales	Other_Sales	Global_Sales	
0	1	Wii Sports	Wii	2006.0	Sports	Nintendo	41.49	29.02	3.77	8.46	82.74	
1	2	Super Mario Bros.	NES	1985.0	Platform	Nintendo	29.08	3.58	6.81	0.77	40.24	
2	3	Mario Kart Wii	Wii	2008.0	Racing	Nintendo	15.85	12.88	3.79	3.31	35.82	
3	4	Wii Sports Resort	Wii	2009.0	Sports	Nintendo	15.75	11.01	3.28	2.96	33.00	
4	5	Pokemon Red/Pokemon Blue	GB	1996.0	Role-Playing	Nintendo	11.27	8.89	10.22	1.00	31.37	
...	
16593	16596	Woody Woodpecker in Crazy Castle 5	GBA	2002.0	Platform	Kemco	0.01	0.00	0.00	0.00	0.01	
16594	16597	Men in Black II: Alien Escape	GC	2003.0	Shooter	Infogrames	0.01	0.00	0.00	0.00	0.01	
16595	16598	SCORE International Baja 1000: The Official Game	PS2	2008.0	Racing	Activision	0.00	0.00	0.00	0.00	0.01	
16596	16599	Know How 2	DS	2010.0	Puzzle	7G//AMES	0.00	0.01	0.00	0.00	0.01	
16597	16600	Spirits & Spells	GBA	2003.0	Platform	Wanadoo	0.01	0.00	0.00	0.00	0.01	

16598 rows × 11 columns

Next steps: [Generate code with df](#) [New interactive sheet](#)

About the dataset

```
#first 5 datas of the dataset
df.head()
```

	Rank	Name	Platform	Year	Genre	Publisher	NA_Sales	EU_Sales	JP_Sales	Other_Sales	Global_Sales	
0	1	Wii Sports	Wii	2006.0	Sports	Nintendo	41.49	29.02	3.77	8.46	82.74	
1	2	Super Mario Bros.	NES	1985.0	Platform	Nintendo	29.08	3.58	6.81	0.77	40.24	
2	3	Mario Kart Wii	Wii	2008.0	Racing	Nintendo	15.85	12.88	3.79	3.31	35.82	
3	4	Wii Sports Resort	Wii	2009.0	Sports	Nintendo	15.75	11.01	3.28	2.96	33.00	
4	5	Pokemon Red/Pokemon Blue	GB	1996.0	Role-Playing	Nintendo	11.27	8.89	10.22	1.00	31.37	

Next steps: [Generate code with df](#) [New interactive sheet](#)

```
#Info of the dataset
df.info()

# there are some missing values in year section
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 16598 entries, 0 to 16597
Data columns (total 11 columns):
 #   Column      Non-Null Count  Dtype  
--- 
 0   Rank        16598 non-null   int64  
 1   Name         16598 non-null   object 
 2   Platform     16598 non-null   object 
 3   Year         16327 non-null   float64
 4   Genre        16598 non-null   object 
 5   Publisher    16540 non-null   object 
 6   NA_Sales     16598 non-null   float64
 7   EU_Sales     16598 non-null   float64
 8   JP_Sales     16598 non-null   float64
 9   Other_Sales  16598 non-null   float64
 10  Global_Sales 16598 non-null   float64
dtypes: float64(6), int64(1), object(4)
memory usage: 1.4+ MB
```

```
# full description of the dataset only numerical values
df.describe()
```

	Rank	Year	NA_Sales	EU_Sales	JP_Sales	Other_Sales	Global_Sales	
count	16598.000000	16327.000000	16598.000000	16598.000000	16598.000000	16598.000000	16598.000000	
mean	8300.605254	2006.406443	0.264667	0.146652	0.077782	0.048063	0.537441	
std	4791.853933	5.828981	0.816683	0.505351	0.309291	0.188588	1.555028	
min	1.000000	1980.000000	0.000000	0.000000	0.000000	0.000000	0.010000	
25%	4151.250000	2003.000000	0.000000	0.000000	0.000000	0.000000	0.060000	
50%	8300.500000	2007.000000	0.080000	0.020000	0.000000	0.010000	0.170000	
75%	12449.750000	2010.000000	0.240000	0.110000	0.040000	0.040000	0.470000	
max	16600.000000	2020.000000	41.490000	29.020000	10.220000	10.570000	82.740000	

```
# removing "Publisher", "Name", "Rank" column
if "Publisher" in df.columns:
    df = df.drop(columns=["Publisher"])
else:
    df = df.copy()

if "Name" in df.columns:
    df = df.drop(columns=["Name"])
else:
    df = df.copy()

if "Rank" in df.columns:
    df = df.drop(columns=["Rank"])
else:
```

```
df = df.copy()
```

```
df.head()
```

	Platform	Year	Genre	NA_Sales	EU_Sales	JP_Sales	Other_Sales	Global_Sales	
0	Wii	2006.0	Sports	41.49	29.02	3.77	8.46	82.74	
1	NES	1985.0	Platform	29.08	3.58	6.81	0.77	40.24	
2	Wii	2008.0	Racing	15.85	12.88	3.79	3.31	35.82	
3	Wii	2009.0	Sports	15.75	11.01	3.28	2.96	33.00	
4	GB	1996.0	Role-Playing	11.27	8.89	10.22	1.00	31.37	

Next steps: [Generate code with df](#) [New interactive sheet](#)

```
#removing duplicates
df_nodup = df.drop_duplicates()
print("After removing Duplicates:")
print("Before: ",df.shape,"| After: ",df_nodup.shape)
```

After removing Duplicates:
Before: (16598, 8) | After: (15020, 8)

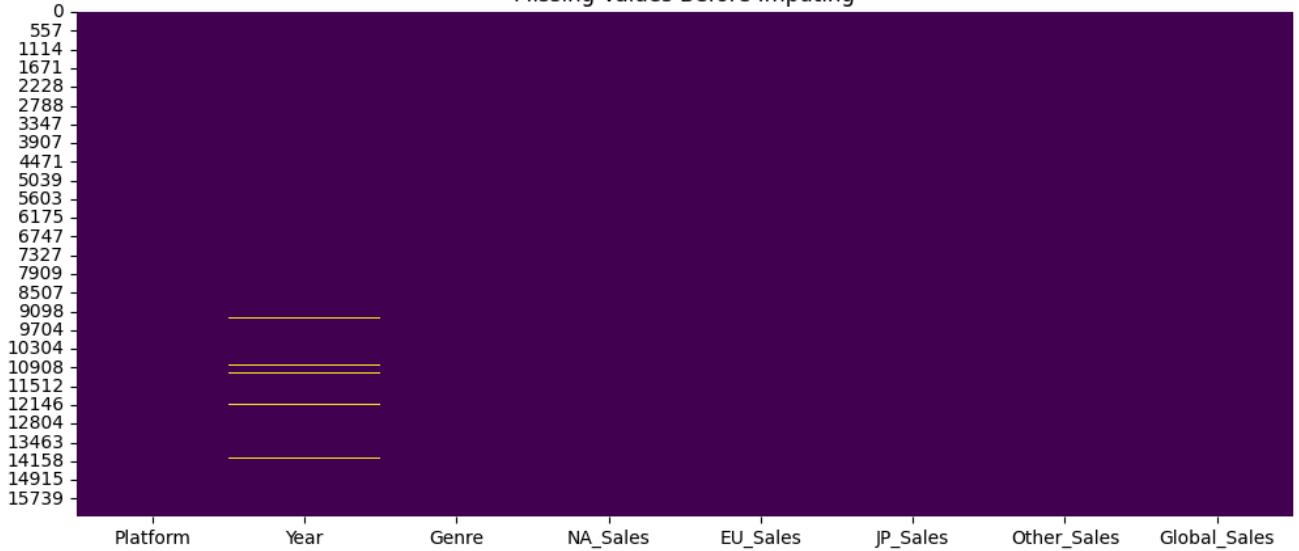
```
# Handling missing values
numeric_cols = df_nodup.select_dtypes(include=["float64", "int64"]).columns.tolist()
categorical_cols = df.select_dtypes(include=["object"]).columns.tolist()

#Visualization Before Imputing
plt.figure(figsize=(12,5))
sns.heatmap(df_nodup.isnull(),cbar=False,cmap="viridis")
plt.title("Missing Values Before Imputing")
plt.show()

#Imputation
df_imputed = df_nodup.copy()
for col in numeric_cols:
    df_imputed[col] = df_imputed[col].fillna(df_imputed[col].median())
for col in categorical_cols:
    df_imputed[col] = df_imputed[col].fillna(df_imputed[col].mode()[0])

#Visualization After Imputing
plt.figure(figsize=(12,5))
sns.heatmap(df_imputed.isnull(),cbar=False,cmap="viridis")
plt.title("Missing Values After Imputing")
plt.show()
```

Missing Values Before Imputing



Missing Values After Imputing

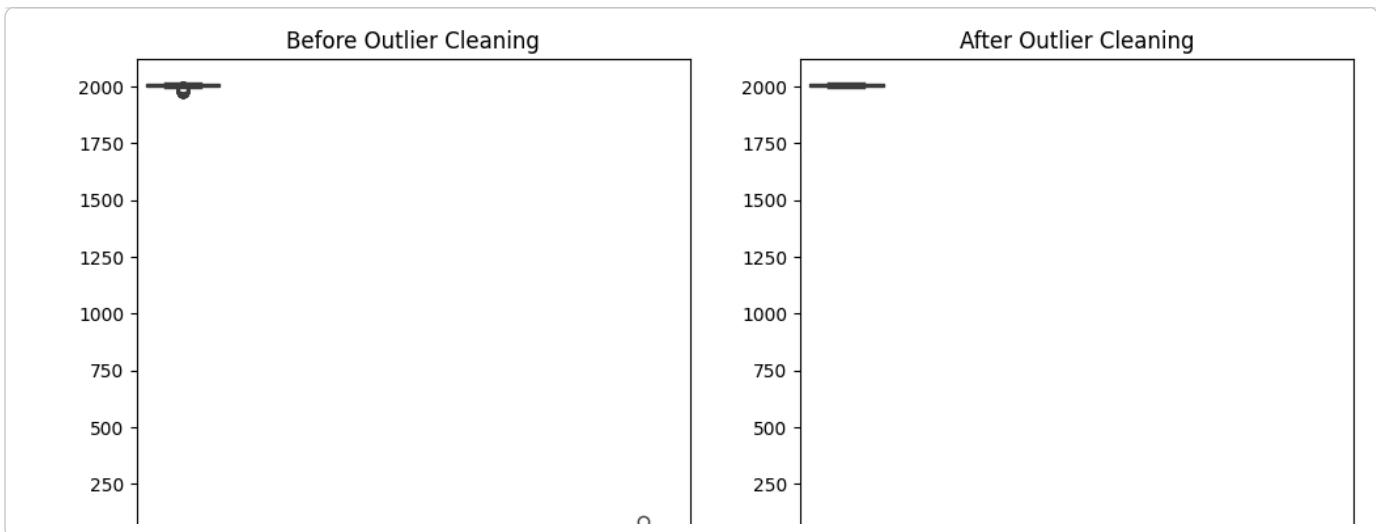


```
# had some problems with year value, the ai helped to clear
#Handling Outliers
df_outlier_cleaned = df_imputed.copy()

for col in numeric_cols:
    q1 = df_outlier_cleaned[col].quantile(0.25)
    q3 = df_outlier_cleaned[col].quantile(0.75)
    iqr = q3 - q1
    lower=q1 -1.5*iqr
    upper=q3+1.5*iqr
    df_outlier_cleaned[col] = df_outlier_cleaned[col].clip(lower, upper)

# Visualization of before and after outlier cleaned
plt.figure(figsize=(12,5))
plt.subplot(1,2,1)
sns.boxplot(data=df_imputed[numeric_cols])
plt.title("Before Outlier Cleaning")

plt.subplot(1,2,2)
sns.boxplot(data=df_outlier_cleaned[numeric_cols])
plt.title("After Outlier Cleaning")
plt.show()
```



```
# Handling categorical value (Encoding+ visualization)
# for col in categorical_cols:
#   plt.figure(figsize=(5,3))
#   df_outlier_cleaned[col].value_counts().plot(kind="bar", color="skyblue", edgecolor="black")
#   plt.title(f"Catagorical Distribution of {col}")
#   plt.show()

#On-hot Encoding
encoded = pd.get_dummies(df_outlier_cleaned[categorical_cols], drop_first=True)
df_encoded = pd.concat([df_outlier_cleaned.drop(categorical_cols, axis=1), encoded], axis=1)
df_encoded.head(5)
```

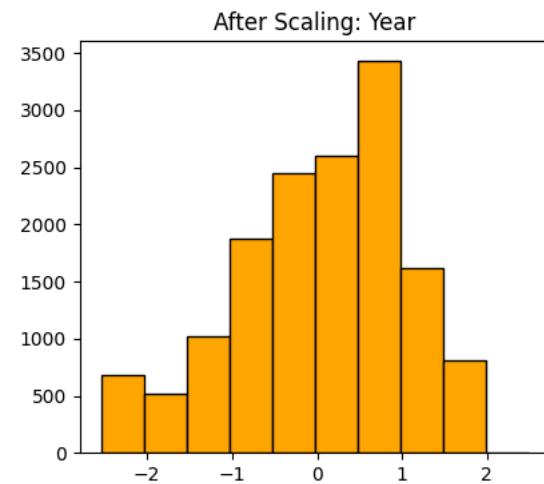
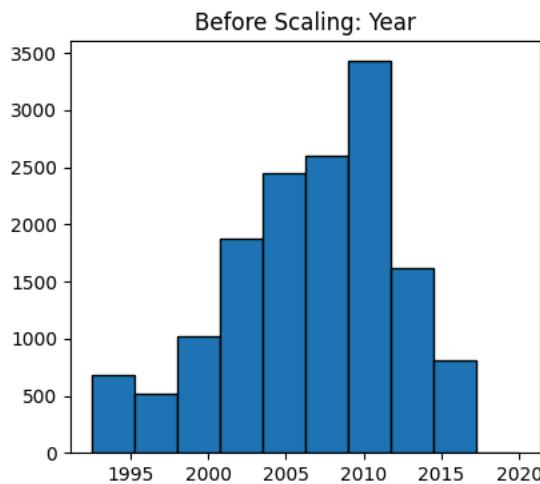
	Year	NA_Sales	EU_Sales	JP_Sales	Other_Sales	Global_Sales	Platform_3DO	Platform_3DS	Platform_DC	Platform_DS	...	Gen
0	2006.0	0.66	0.325	0.1	0.1	1.205	False	False	False	False	...	
1	1992.5	0.66	0.325	0.1	0.1	1.205	False	False	False	False	...	
2	2008.0	0.66	0.325	0.1	0.1	1.205	False	False	False	False	...	
3	2009.0	0.66	0.325	0.1	0.1	1.205	False	False	False	False	...	
4	1996.0	0.66	0.325	0.1	0.1	1.205	False	False	False	False	...	

5 rows × 47 columns

```
scaler = StandardScaler()
df_scaled = df_encoded.copy()
df_scaled[numeric_cols] = scaler.fit_transform(df_scaled[numeric_cols])

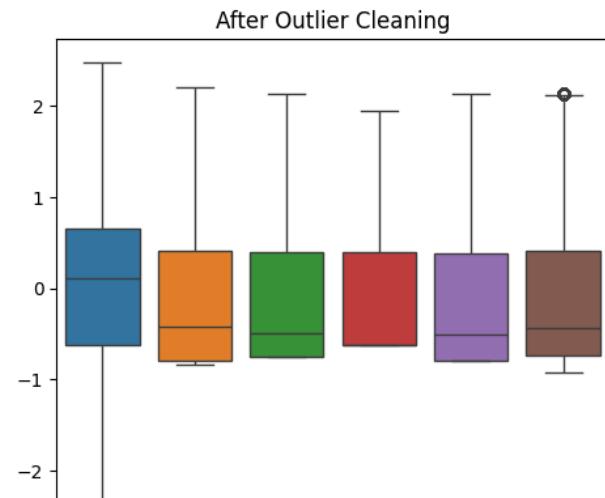
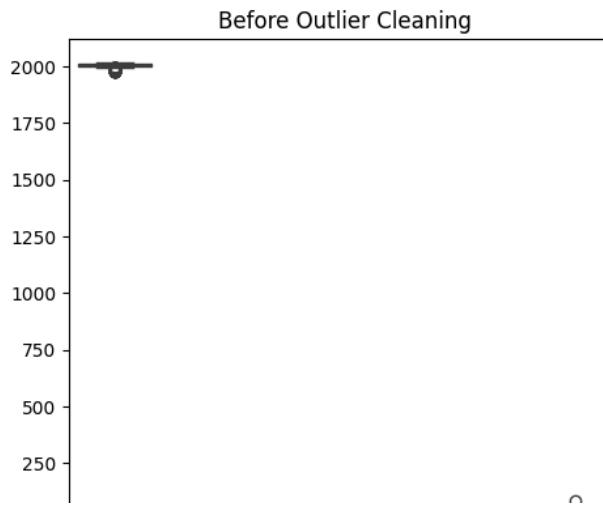
# Visualization: Histogram before vs after scaling (Age example)
plt.figure(figsize=(10,4))
plt.subplot(1,2,1)
plt.hist(df_encoded["Year"], bins=10, edgecolor="black")
plt.title("Before Scaling: Year")

plt.subplot(1,2,2)
plt.hist(df_scaled["Year"], bins=10, edgecolor="black", color="orange")
plt.title("After Scaling: Year")
plt.show()
```



```
# Visualization of before and after outlier cleaned
plt.figure(figsize=(12,5))
plt.subplot(1,2,1)
sns.boxplot(data=df_imputed[numeric_cols])
plt.title("Before Outlier Cleaning")

plt.subplot(1,2,2)
sns.boxplot(data=df_scaled[numeric_cols])
plt.title("After Outlier Cleaning")
plt.show()
```



```
# Example: Suppose "EmployeeID" column is irrelevant
if "Rank" in df_scaled.columns:
    df_final = df_scaled.drop(columns=["Rank"])
else:
    df_final = df_scaled.copy()

print("◆ After removing irrelevant features")
print(df_final.head())
```

◆ After removing irrelevant features

Year	NA_Sales	EU_Sales	JP_Sales	Other_Sales	Global_Sales
0	-0.074254	2.214921	2.138876	1.9512	2.139563
1	-2.543353	2.214921	2.138876	1.9512	2.139563
2	0.291538	2.214921	2.138876	1.9512	2.139563
3	0.474434	2.214921	2.138876	1.9512	2.139563
4	-1.903216	2.214921	2.138876	1.9512	2.139563

Platform_3DO	Platform_3DS	Platform_DC	Platform_DS	...	Genre_Fighting
0	False	False	False	False	False
1	False	False	False	False	False
2	False	False	False	False	False
3	False	False	False	False	False
4	False	False	False	False	False

```

Genre_Misc  Genre_Platform  Genre_Puzzle  Genre_Racing  Genre_Role-Playing \
0      False        False     False     False       False
1      False       True     False     False       False
2      False      False     False      True       False
3      False      False     False     False       False
4      False      False     False     False       True

Genre_Shooter  Genre_Simulation  Genre_Sports  Genre_Strategy
0      False        False      True     False
1      False        False     False     False
2      False        False     False     False
3      False        False      True     False
4      False        False     False     False

[5 rows x 47 columns]

```

```

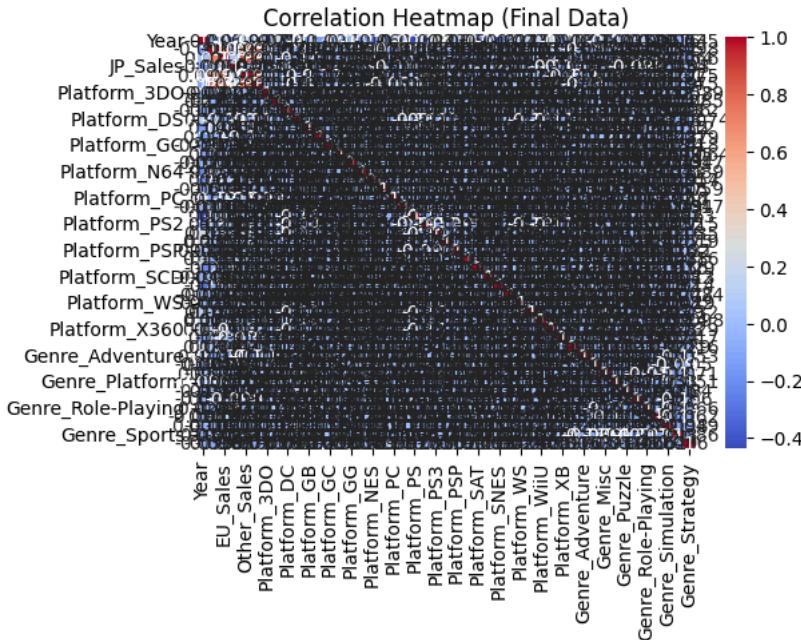
# save and finalize
# with pd.ExcelWriter("processed_employees.xlsx") as writer:
#     df.to_excel(writer, sheet_name="Raw Data", index=False)
#     df_nodup.to_excel(writer, sheet_name="No Duplicates", index=False)
#     df_imputed.to_excel(writer, sheet_name="After Imputation", index=False)
#     df_outlier_cleaned.to_excel(writer, sheet_name="After Outlier Clipping", index=False)
#     df_encoded.to_excel(writer, sheet_name="After Encoding", index=False)
#     df_scaled.to_excel(writer, sheet_name="After Scaling", index=False)
#     df_final.to_excel(writer, sheet_name="Final Processed", index=False)

print("✅ All datasets exported to 'processed_employees.xlsx'")

#Final visualization: Heatmap of correlations
plt.figure(figsize=(6,4))
sns.heatmap(df_final.corr(), annot=True, cmap="coolwarm")
plt.title("Correlation Heatmap (Final Data)")
plt.show()

```

All datasets exported to 'processed_employees.xlsx'



Using ML Algorithms

```

# Importing libraries for ML
from sklearn.model_selection import train_test_split
from sklearn.metrics import accuracy_score
from sklearn.tree import DecisionTreeClassifier
from sklearn.neighbors import KNeighborsClassifier
from sklearn.ensemble import RandomForestClassifier

df_ml=df_outlier_cleaned.copy()
df_ml

```

	Platform	Year	Genre	NA_Sales	EU_Sales	JP_Sales	Other_Sales	Global_Sales	
0	Wii	2006.0	Sports	0.66	0.325	0.1	0.1	1.205	
1	NES	1992.5	Platform	0.66	0.325	0.1	0.1	1.205	
2	Wii	2008.0	Racing	0.66	0.325	0.1	0.1	1.205	
3	Wii	2009.0	Sports	0.66	0.325	0.1	0.1	1.205	
4	GB	1996.0	Role-Playing	0.66	0.325	0.1	0.1	1.205	
...	
16592	DS	2008.0	Simulation	0.01	0.000	0.0	0.0	0.010	
16593	GBA	2002.0	Platform	0.01	0.000	0.0	0.0	0.010	
16595	PS2	2008.0	Racing	0.00	0.000	0.0	0.0	0.010	
16596	DS	2010.0	Puzzle	0.00	0.010	0.0	0.0	0.010	
16597	GBA	2003.0	Platform	0.01	0.000	0.0	0.0	0.010	

Next steps: [Generate code with df_ml](#) [New interactive sheet](#)

```
# DecisionTreeClassifier
#Spliting data
X = df_ml.drop(columns=["Global_Sales","Platform","Genre"])
Y = df_ml["Global_Sales"]

X_train, X_test, Y_train, Y_test = train_test_split(X, Y, test_size=0.2)

model = DecisionTreeClassifier()
model.fit(X_train, Y_train)
predictions = model.predict(X_test)
score = accuracy_score(Y_test, predictions)
print("Accuracy Score: ",score)

# problem: DT can not convert to string to number
```

```
-----  
KeyError                                     Traceback (most recent call last)  
/tmp/ipython-input-3070237132.py in <cell line: 0>()  
      1 # DecisionTreeClassifier  
      2 #Spliting data  
----> 3 X = df_ml.drop(columns=["Global_Sales","Platform","Genre"])  
      4 Y = df_ml["Global_Sales"]  
      5  
-----  
          ▲ 3 frames  
/usr/local/lib/python3.12/dist-packages/pandas/core/indexes/base.py in drop(self, labels, errors)  
 7068         if mask.any():
```