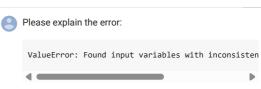
```
or
import pandas as pd
# Load the dataset
file_path = "heart_attack_youngsters_india.csv"
df = pd.read csv(file path)
# Display basic information and first few rows
df.info(), df.head()
      22 Exercise Induced Angina
                                                      10000 non-null
\rightarrow
                                                                      object
         Blood Oxygen Levels (Sp02%)
                                                      10000 non-null
      23
                                                                       float64
      24
         Triglyceride Levels (mg/dL)
                                                      10000 non-null
                                                                       int64
      25 Heart Attack Likelihood
                                                      10000 non-null
                                                                      obiect
     dtypes: float64(2), int64(7), object(17)
     memory usage: 2.0+ MB
     (None,
                                             SES Smoking Status Alcohol Consumption
              Gender Region Urban/Rural
         Age
                Male
      0
          30
                        East
                                   Urban Middle
                                                           Never
                                                                            Regularly
                                             Low
      1
          24
              Female
                        East
                                   Urban
                                                    Occasionally
                                                                         Occasionally
      2
          24
              Female
                       North
                                   Urban
                                             Low
                                                    Occasionally
                                                                         Occasionally
      3
          27
                Male
                        East
                                   Urban
                                         Middle
                                                    Occasionally
                                                                                Never
      4
          21
              Female
                        West
                                   Rural
                                             Low
                                                    Occasionally
                                                                         Occasionally
              Diet Type Physical Activity Level Screen Time (hrs/day)
                                                                           . . .
         Non-Vegetarian
                                       Sedentary
      1
         Non-Vegetarian
                                       Sedentary
                                                                       15
                                                                          . . .
      2
                  Vegan
                                            High
                                                                       15
                                                                          . . .
                                       Sedentary
      3
             Vegetarian
                                                                        6
             Vegetarian
      4
                                        Moderate
                                                                        4
         Stress Level Blood Pressure (systolic/diastolic mmHg)
                                                      177.0/63.1
      0
                 High
      1
                 High
                                                     137.5/110.7
      2
                                                      138.3/76.6
                  Low
      3
               Medium
                                                      177.1/90.0
                                                     130.7/108.8
                  Low
        Resting Heart Rate (bpm) ECG Results Chest Pain Type
      0
                               82
                                       Normal
                                                    Non-anginal
      1
                               76
                                       Normal
                                                    Non-anginal
      2
                               86
                                       Normal
                                                        Typical
                                       Normal
      3
                              106
                                                    Non-anginal
      4
                                       Normal
                                                       Atypical
         Maximum Heart Rate Achieved Exercise Induced Angina
                                  183
      1
                                  118
                                                            Nο
      2
                                  164
                                                            No
      3
                                  188
                                                            No
      4
                                  216
                                                            No
        Blood Oxygen Levels (SpO2%) Triglyceride Levels (mg/dL)
                                94.1
      1
                                97.1
      2
                                92.7
                                                               373
      3
                                98.4
                                                               102
                                                               235
        Heart Attack Likelihood
                              No
      1
                              Nο
      2
                             Yes
      3
                             Yes
      4
                              No
      [5 rows x 26 columns])
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler, LabelEncoder
from sklearn.ensemble import RandomForestClassifier
from sklearn.svm import SVC
from xgboost import XGBClassifier
from \ sklearn.metrics \ import \ accuracy\_score, \ classification\_report
# Encode categorical variables
label encoders = {}
for col in df.select_dtypes(include=['object']).columns:
    if col != 'Heart Attack Likelihood': # Keep target variable separate
        label encoders[col] = LabelEncoder()
        df[col] = label_encoders[col].fit_transform(df[col])
```



ValueError

UnboundLocalError

## Hypothesis

The error "ValueError: Found input variables with inconsistent numbers of samples: [7163, 0]" indicates that the x and y variables used in the train\_test\_split function have different numbers of samples. X has 7163 samples, while y has 0 samples. This likely happened because you're using the original df\_genz to define x, but a potentially empty or modified version of df\_genz to define y. Remember that in cell 10, you converted 'Heart Attack Likelihood' to numeric and then in cell 13, you dropped NaNs from the 'Heart Attack Likelihood'. Because x is defined before and y is defined after these steps, they end up having a different number of rows.

## Suggested Changes

```
囗
import pandas as pd
import numpy as np
from sklearn.model_selection import train_test_s
from sklearn.linear_model import LogisticRegress:
from sklearn.metrics import accuracy_score, class
from sklearn.preprocessing import StandardScaler
from sklearn.compose import ColumnTransformer
from sklearn.pipeline import Pipeline
# --- 1. Data Preparation ---
# Filter Gen Z (adjust age threshold if needed)
df_genz = df[df["Age"] <= 30]
# --- Ensure 'Heart Attack Likelihood' is numeri
df_genz['Heart Attack Likelihood'] = df_genz['He
# --- Drop rows with NaN in 'Heart Attack Likeli
df_genz = df_genz.dropna(subset=['Heart Attack L
# Define features (X) and target (y) *after* cle
X = df_genz[[
   "Sleep Duration (hrs/day)",
   "Screen Time (hrs/day)",
    "Diet Type",
    "Stress Level",
    "Physical Activity Level",
    "BMI (kg/m²)",
    "Cholesterol Levels (mg/dL)",
    "Triglyceride Levels (mg/dL)"
11
y = df genz["Heart Attack Likelihood"] # Binary
# Train-test split
X_train, X_test, y_train, y_test = train_test_sp
# --- 2. Preprocessing Pipeline ---
# Categorical features (to encode)
cat_cols = ["Diet Type", "Stress Level", "Physic
# Numeric features (to scale)
num cols = ["Sleep Duration (hrs/day)", "Screen
            "Cholesterol Levels (mg/dL)", "Trigl
```

# Define features and target

# Convert target variable to binary (0 = No, 1 = Yes)

X = df.drop(columns=['Heart Attack Likelihood'])

df['Heart Attack Likelihood'] = df['Heart Attack Likelihood'].map({'No': 0, 'Yes': 1

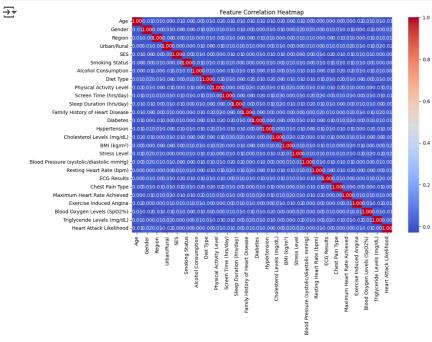
```
y = df['Heart Attack Likelihood']
# Split into training and testing sets (80-20 split)
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_stat
# Standardize numerical features
scaler = StandardScaler()
X_train = scaler.fit_transform(X_train)
X_test = scaler.transform(X_test)
# Train multiple models
models = {
    "Random Forest": RandomForestClassifier(n_estimators=100, random_state=42),
    "SVM": SVC(kernel='rbf', probability=True),
    "XGBoost": XGBClassifier(use_label_encoder=False, eval_metric='logloss', random_
}
# Evaluate models
results = {}
for name, model in models.items():
   model.fit(X_train, y_train)
   y_pred = model.predict(X_test)
    accuracy = accuracy_score(y_test, y_pred)
    results[name] = accuracy
    print(f"{name} Accuracy: {accuracy:.4f}")
    print(classification_report(y_test, y_pred))
# Identify the best model
best_model_name = max(results, key=results.get)
print(f"Best Model: {best_model_name} with Accuracy: {results[best_model_name]:.4f}'
Random Forest Accuracy: 0.7960
                                recall f1-score
                   precision
                                                    support
                0
                        0.80
                                  1.00
                                                       1592
                                            0.89
                1
                        0.00
                                  0.00
                                            0.00
                                                        408
                                             0.80
                                                       2000
         accuracy
                        0.40
                                  0.50
                                             0.44
        macro avg
                                                       2000
     weighted avg
                        0.63
                                  0.80
                                             0.71
                                                       2000
     /usr/local/lib/python3.11/dist-packages/sklearn/metrics/_classification.py:1565
       warn prf(average, modifier, f"{metric.capitalize()} is", len(result))
     /usr/local/lib/python3.11/dist-packages/sklearn/metrics/_classification.py:1565
       _warn_prf(average, modifier, f"{metric.capitalize()} is", len(result))
     /usr/local/lib/python3.11/dist-packages/sklearn/metrics/_classification.py:1565
       _warn_prf(average, modifier, f"{metric.capitalize()} is", len(result))
     SVM Accuracy: 0.7960
                   precision
                                recall f1-score support
                0
                        0.80
                                  1.00
                                             0.89
                                                       1592
                1
                        0.00
                                  0.00
                                             0.00
                                                        408
                                             0.80
                                                       2000
        accuracy
                                  0.50
                        0.40
                                             0.44
                                                       2000
        macro avg
     weighted avg
                                            0.71
                        0.63
                                  0.80
                                                       2000
     /usr/local/lib/python3.11/dist-packages/sklearn/metrics/_classification.py:1565
       _warn_prf(average, modifier, f"{metric.capitalize()} is", len(result))
     /usr/local/lib/python 3.11/dist-packages/sklearn/metrics/\_classification.py: 1565
       _warn_prf(average, modifier, f"{metric.capitalize()} is", len(result))
     /usr/local/lib/python3.11/dist-packages/sklearn/metrics/_classification.py:1565
       _warn_prf(average, modifier, f"{metric.capitalize()} is", len(result))
     /usr/local/lib/python3.11/dist-packages/xgboost/core.py:158: UserWarning: [18:29
     Parameters: { "use_label_encoder" } are not used.
       warnings.warn(smsg, UserWarning)
     XGBoost Accuracy: 0.7845
                   precision
                                recall f1-score
                                                    support
                a
                        0.80
                                  0.97
                                             0.88
                                                       1592
                                                        408
                        0.34
                                  0.06
                                             0.10
         accuracy
                                             0.78
                                                       2000
                        0.57
                                  0.51
                                             0.49
                                                       2000
        macro avg
                                            0.72
                                                       2000
                        0.71
                                  0.78
     weighted avg
     Best Model: Random Forest with Accuracy: 0.7960
```

```
# Preprocessor
 preprocessor = ColumnTransformer(
     transformers=[
         ("num", StandardScaler(), num_cols),
         ("cat", OneHotEncoder(), cat_cols)
 # --- 3. Model Training ---
 # Pipeline with logistic regression
 model = Pipeline(steps=[
     ("preprocessor", preprocessor),
     ("classifier", LogisticRegression(max_iter=1
 1)
 model.fit(X_train, y_train)
 # --- 4. Model Evaluation ---
 y pred = model.predict(X test)
 y_pred_proba = model.predict_proba(X_test)[:, 1]
 print("=== Model Performance ===")
 print(f"Accuracy: {accuracy_score(y_test, y_pred
 print(f"ROC-AUC: {roc_auc_score(y_test, y_pred_p
 print("\nClassification Report:")
 print(classification_report(y_test, y_pred))
 # --- 5. Feature Importance ---
 # Extract coefficients (for interpretation)
 log reg = model.named steps["classifier"]
 feature names = (preprocessor.named transformers
 feature_names = num_cols + feature_names
 coefficients = pd.DataFrame({
     "Feature": feature_names,
     "Coefficient": log reg.coef [0]
 }).sort_values(by="Coefficient", ascending=False
 print("\n=== Top Risk Factors ===")
 print(coefficients.head(10))
Use code with caution
Sources
```



```
import seaborn as sns
import matplotlib.pyplot as plt

plt.figure(figsize=(12, 8))
sns.heatmap(df.corr(), annot=True, cmap="coolwarm", fmt=".2f")
plt.title("Feature Correlation Heatmap")
plt.show()
```

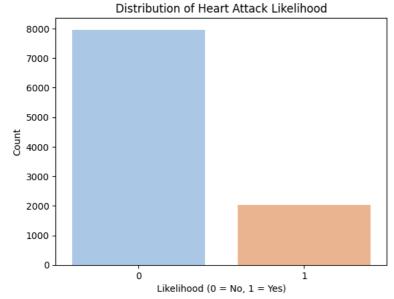


```
sns.countplot(x=df["Heart Attack Likelihood"], palette="pastel")
plt.title("Distribution of Heart Attack Likelihood")
plt.xlabel("Likelihood (0 = No, 1 = Yes)")
plt.ylabel("Count")
plt.show()
```

<ipython-input-7-faa6bcfc8dfb>:1: FutureWarning:

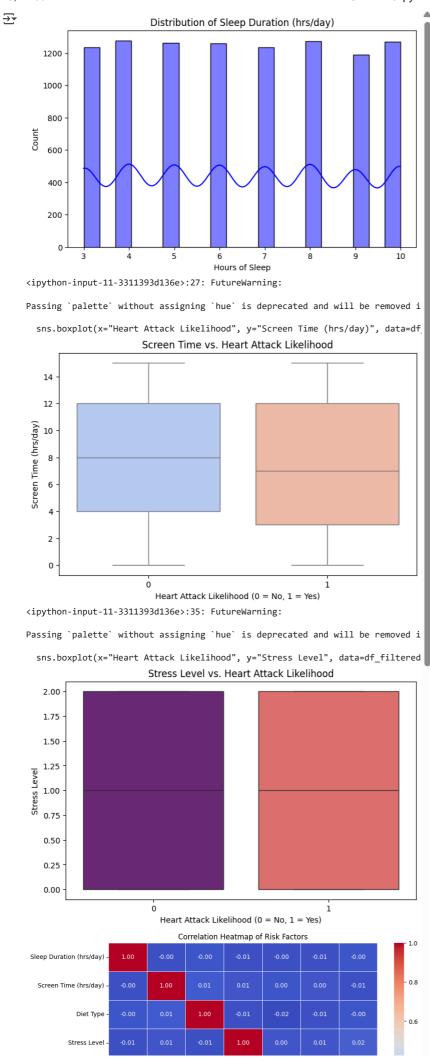
Passing `palette` without assigning `hue` is deprecated and will be removed in  $\boldsymbol{\nu}$ 

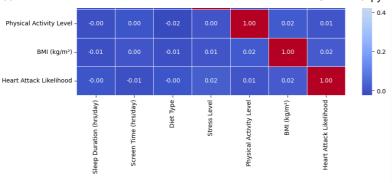
 $\verb|sns.countplot(x=df["Heart Attack Likelihood"], palette="pastel")|\\$ 



## df.columns

Start coding or generate with AI.

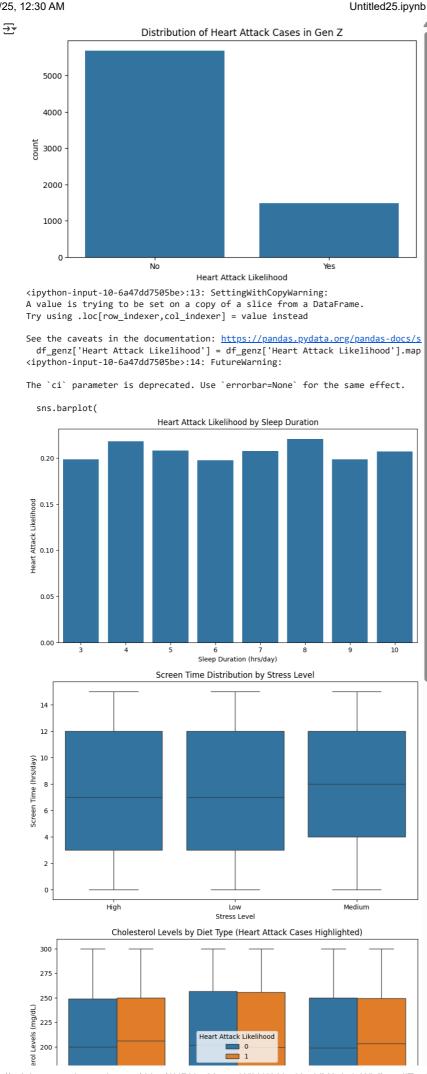


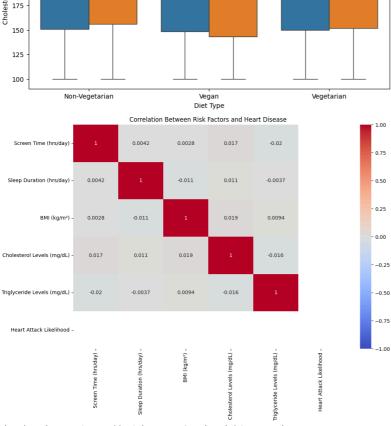


```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from scipy.stats import chi2_contingency, ttest_ind
import statsmodels.api as sm
# Load data
df = pd.read_csv("heart_attack_youngsters_india.csv")
# Filter Gen Z (assuming age <=24)</pre>
df_genz = df[df["Age"] <= 30]
# --- Convert 'Heart Attack Likelihood' to numerical before groupby ---
# This is crucial to avoid the TypeError
df_genz['Heart Attack Likelihood'] = df_genz['Heart Attack Likelihood'].map({'No': 0,
# --- 1. Descriptive Analysis ---
print("=== Descriptive Statistics ===")
# Prevalence of risk factors
print("Sleep <6hrs:", round((df_genz["Sleep Duration (hrs/day)"] < 6).mean() * 100, 1</pre>
print("Screen Time >4hrs:", round((df_genz["Screen Time (hrs/day)"] > 4).mean() * 100
print("Unhealthy Diet:", round((df_genz["Diet Type"] == "Unhealthy").mean() * 100, 1)
print("Low Activity:", round((df_genz["Physical Activity Level"] == "Low").mean() * 1
print("High Stress:", round((df_genz["Stress Level"] == "High").mean() * 100, 1), "%"
# Heart attack rates by risk factors
print("\n=== Heart Attack Likelihood by Risk Factor ===")
print("Screen Time >4hrs:")
print(df_genz.groupby("Screen Time (hrs/day)")["Heart Attack Likelihood"].mean())
print("\nSleep <6hrs:")</pre>
print(df genz.groupby("Sleep Duration (hrs/day)")["Heart Attack Likelihood"].mean())
# --- 2. Correlation Analysis ---
# ... (rest of the code remains the same)
⇒ === Descriptive Statistics ===
     Sleep <6hrs: 37.6 %
     Screen Time >4hrs: 68.5 %
     Unhealthy Diet: 0.0 %
     Low Activity: 0.0 %
     High Stress: 29.1 %
     === Heart Attack Likelihood by Risk Factor ===
     Screen Time >4hrs:
```

```
Screen Time (hrs/day)
     0
           0.207188
     1
           0.187225
           0.205817
           0.230415
           0.205817
     5
           0.207792
           0.217703
     6
           0.193258
     8
           0.210300
           0.206089
     9
     10
           0.192941
     11
           0.222222
     12
           0.217489
           0.213220
     13
     14
           0.193548
     15
           0.202299
     Name: Heart Attack Likelihood, dtype: float64
     Sleep <6hrs:
     Sleep Duration (hrs/day)
     3
           0.198422
           0.217920
     5
           0.207778
           0.197309
           0.207399
           0.220573
           0.198307
     10
           0.206972
     Name: Heart Attack Likelihood, dtype: float64
     <ipython-input-7-8d8e78534442>:16: SettingWithCopyWarning:
     A value is trying to be set on a copy of a slice from a DataFrame.
     Try using .loc[row_indexer,col_indexer] = value instead
     See the caveats in the documentation: <a href="https://pandas.pydata.org/pandas-docs/stal">https://pandas.pydata.org/pandas-docs/stal</a>
       df_genz['Heart Attack Likelihood'] = df_genz['Heart Attack Likelihood'].map({
# Filter Gen Z (adjust age threshold if needed)
df_genz = df[df["Age"] <= 30]</pre>
# --- 1. Distribution of Heart Attack Likelihood ---
plt.figure(figsize=(8, 5))
sns.countplot(x="Heart Attack Likelihood", data=df_genz)
plt.title("Distribution of Heart Attack Cases in Gen Z")
plt.show()
# --- 2. Sleep Duration vs. Heart Attack Likelihood ---
plt.figure(figsize=(10, 6))
# --- Ensure 'Heart Attack Likelihood' is numerical for barplot ---
df_genz['Heart Attack Likelihood'] = df_genz['Heart Attack Likelihood'].map({'No': 0,
sns.barplot(
   x="Sleep Duration (hrs/day)",
    y="Heart Attack Likelihood",
    data=df_genz,
    ci=None
plt.title("Heart Attack Likelihood by Sleep Duration")
plt.show()
# --- 3. Screen Time vs. Stress Level ---
plt.figure(figsize=(10, 6))
sns.boxplot(
   x="Stress Level",
    y="Screen Time (hrs/day)",
    data=df_genz
plt.title("Screen Time Distribution by Stress Level")
plt.show()
# --- 4. Diet Type vs. Cholesterol Levels ---
plt.figure(figsize=(10, 6))
sns.boxplot(
    x="Diet Type";
    y="Cholesterol Levels (mg/dL)",
    hue="Heart Attack Likelihood",
    data=df genz
plt.title("Cholesterol Levels by Diet Type (Heart Attack Cases Highlighted)")
plt.show()
# --- 6. Correlation Heatmap ---
corr_vars = df_genz[[
    "Screen Time (hrs/day)",
    "Sleep Duration (hrs/day)",
```

```
"BMI (kg/m²)",
    "Cholesterol Levels (mg/dL)",
    "Triglyceride Levels (mg/dL)",
    "Heart Attack Likelihood" # This column likely has 'Yes'/'No'
]].copy() # Create a copy to avoid modifying the original df_genz
# Convert 'Heart Attack Likelihood' to numerical for correlation
corr_vars["Heart Attack Likelihood"] = corr_vars["Heart Attack Likelihood"].map({"No"
plt.figure(figsize=(12, 8))
sns.heatmap(corr_vars.corr(), annot=True, cmap="coolwarm", vmin=-1, vmax=1)
plt.title("Correlation Between Risk Factors and Heart Disease")
plt.show()
# --- 7. Heart Attack Cases by Risk Factor Combinations ---
# Create binary risk factors
df_genz["High Screen Time"] = (df_genz["Screen Time (hrs/day)"] > 4).astype(int)
df_genz["Poor Sleep"] = (df_genz["Sleep Duration (hrs/day)"] < 6).astype(int)</pre>
df_genz["Unhealthy Diet"] = (df_genz["Diet Type"] == "Unhealthy").astype(int)
# Aggregate data
# --- Ensure 'Heart Attack Likelihood' is numerical before groupby ---
df_genz['Heart Attack Likelihood'] = df_genz['Heart Attack Likelihood'].map({'No': 0,
risk_factors = df_genz.groupby(
    ["High Screen Time", "Poor Sleep", "Unhealthy Diet"]
)["Heart Attack Likelihood"].mean().reset_index()
# Plot
plt.figure(figsize=(10, 6))
sns.barplot(
    x="High Screen Time",
    y="Heart Attack Likelihood",
   hue="Unhealthy Diet",
    data=risk_factors
plt.title("Heart Attack Likelihood by Combined Risk Factors")
plt.show()
```





<ipython-input-10-6a47dd7505be>:64: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row\_indexer,col\_indexer] = value instead

See the caveats in the documentation: <a href="https://pandas.pydata.org/pandas-docs/s">https://pandas.pydata.org/pandas-docs/s</a> df\_genz["High Screen Time"] = (df\_genz["Screen Time (hrs/day)"] > 4).astype <ipython-input-10-6a47dd7505be>:65: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.

Try using .loc[row\_indexer,col\_indexer] = value instead

See the caveats in the documentation: <a href="https://pandas.pydata.org/pandas-docs/s">https://pandas.pydata.org/pandas-docs/s</a> df\_genz["Poor Sleep"] = (df\_genz["Sleep Duration (hrs/day)"] < 6).astype(in <ipython-input-10-6a47dd7505be>:66: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.

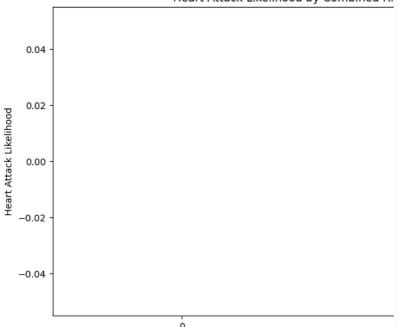
Try using .loc[row\_indexer,col\_indexer] = value instead

See the caveats in the documentation: <a href="https://pandas.pydata.org/pandas-docs/s">https://pandas.pydata.org/pandas-docs/s</a> df\_genz["Unhealthy Diet"] = (df\_genz["Diet Type"] == "Unhealthy").astype(in <ipython-input-10-6a47dd7505be):70: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.

Try using .loc[row\_indexer,col\_indexer] = value instead

See the caveats in the documentation: <a href="https://pandas.pydata.org/pandas-docs/s">https://pandas.pydata.org/pandas-docs/s</a> df\_genz['Heart Attack Likelihood'] = df\_genz['Heart Attack Likelihood'].map

## Heart Attack Likelihood by Combined Ri



High Screen Time