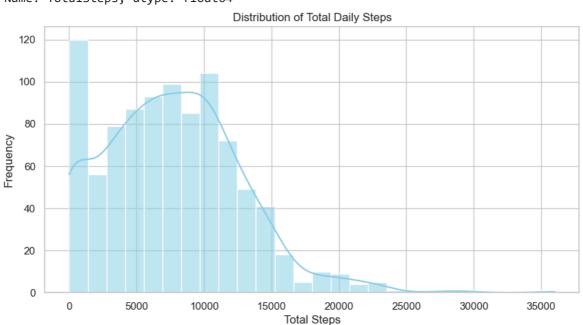
```
In [1]: # ==========
        # 簓 1. Import Libraries
        # =========
        import pandas as pd
        import matplotlib.pyplot as plt
        import seaborn as sns
        # Configure styles
        sns.set(style='whitegrid')
        plt.rcParams['figure.figsize'] = (10, 5)
# 📁 2. Load the 4 Main Datasets
        activity = pd.read_csv('dailyActivity_merged.csv')
        sleep = pd.read_csv('sleepDay_merged.csv')
        weight = pd.read csv('weightLogInfo merged.csv')
        heart = pd.read_csv('heartrate_seconds_merged.csv')
In [3]: # ==========
        # 🥜 3. Data Cleaning
        # ==========
        # Convert date/time columns
        activity['ActivityDate'] = pd.to_datetime(activity['ActivityDate'])
        sleep['SleepDay'] = pd.to_datetime(sleep['SleepDay'])
        weight['Date'] = pd.to_datetime(weight['Date'])
        heart['Time'] = pd.to_datetime(heart['Time'])
        # Drop duplicates
        activity.drop_duplicates(inplace=True)
        sleep.drop_duplicates(inplace=True)
        weight.drop duplicates(inplace=True)
        heart.drop_duplicates(inplace=True)
        # Check for missing values
        print("Missing Values:")
        print(activity.isnull().sum())
        print(sleep.isnull().sum())
        print(weight.isnull().sum())
        print(heart.isnull().sum())
      C:\Users\ASUS\AppData\Local\Temp\ipykernel_3008\2657426346.py:7: UserWarning: Cou
      ld not infer format, so each element will be parsed individually, falling back to
      `dateutil`. To ensure parsing is consistent and as-expected, please specify a for
        sleep['SleepDay'] = pd.to_datetime(sleep['SleepDay'])
      C:\Users\ASUS\AppData\Local\Temp\ipykernel_3008\2657426346.py:8: UserWarning: Cou
      ld not infer format, so each element will be parsed individually, falling back to
      `dateutil`. To ensure parsing is consistent and as-expected, please specify a for
        weight['Date'] = pd.to datetime(weight['Date'])
```

```
Missing Values:
                               0
      ActivityDate
                               0
      TotalSteps
                               0
      TotalDistance
                               0
      TrackerDistance
                               0
      LoggedActivitiesDistance
                               0
      VeryActiveDistance
      ModeratelyActiveDistance
                               0
      LightActiveDistance
      SedentaryActiveDistance
                               0
      VeryActiveMinutes
      FairlyActiveMinutes
                               0
      LightlyActiveMinutes
                               0
      SedentaryMinutes
                               0
      Calories
                               0
      dtype: int64
      Ιd
                          0
      SleepDay
                          0
      TotalSleepRecords
                          0
      TotalMinutesAsleep
      TotalTimeInBed
      dtype: int64
      Ιd
                       0
      Date
                       0
      WeightKg
                       0
      WeightPounds
                      0
      Fat
                      65
      BMI
                       0
      IsManualReport
                       0
      LogId
                       0
      dtype: int64
      Id
              a
      Time
              0
      Value
      dtype: int64
# Ø 4. Merge Sleep + Activity (Optional)
       merged = pd.merge(activity, sleep, left_on=['Id', 'ActivityDate'], right_on=['Id
In [5]:
      # ===========
       # 📊 5. Data Exploration
       # =========
       # --- A. Steps Summary ---
       print("Step Count Summary:")
       print(activity['TotalSteps'].describe())
       sns.histplot(activity['TotalSteps'], kde=True, color='skyblue')
       plt.title("Distribution of Total Daily Steps")
       plt.xlabel("Total Steps")
       plt.ylabel("Frequency")
       plt.show()
```

```
Step Count Summary:
           940.000000
count
          7637.910638
mean
std
          5087.150742
min
             0.000000
25%
          3789.750000
50%
          7405.500000
75%
         10727.000000
         36019.000000
max
Name: TotalSteps, dtype: float64
```

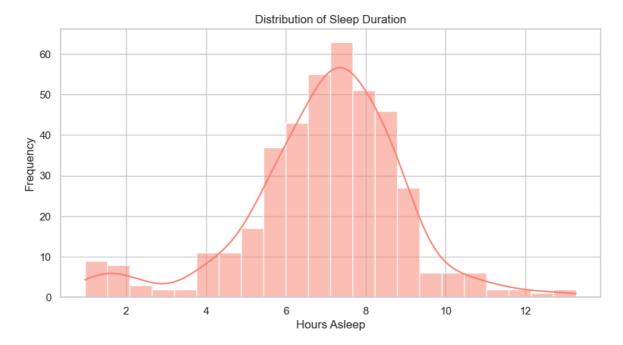


```
In [6]: # --- B. Sleep Duration ---
print("Sleep Summary (in hours):")
print((sleep['TotalMinutesAsleep'] / 60).describe())

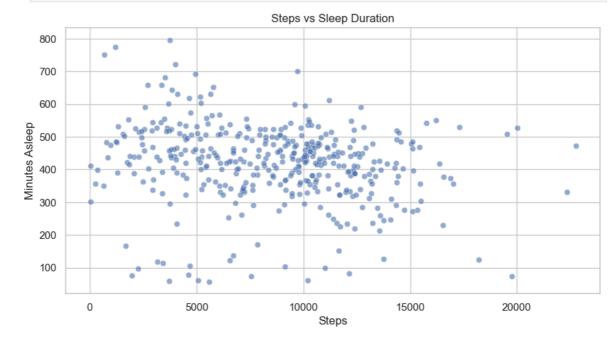
sns.histplot(sleep['TotalMinutesAsleep'] / 60, kde=True, color='salmon')
plt.title("Distribution of Sleep Duration")
plt.xlabel("Hours Asleep")
plt.ylabel("Frequency")
plt.show()
```

```
Sleep Summary (in hours):
         410.000000
count
mean
           6.986220
std
           1.977265
           0.966667
min
25%
           6.016667
50%
           7.208333
75%
           8.166667
          13.266667
max
```

Name: TotalMinutesAsleep, dtype: float64

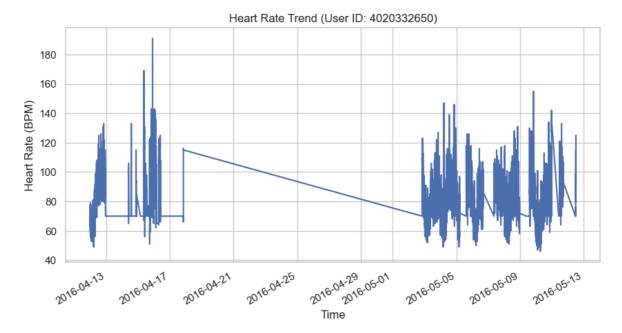


```
In [7]: # --- C. Steps vs Sleep Scatter ---
sns.scatterplot(data=merged, x='TotalSteps', y='TotalMinutesAsleep', alpha=0.6)
plt.title("Steps vs Sleep Duration")
plt.xlabel("Steps")
plt.ylabel("Minutes Asleep")
plt.show()
```



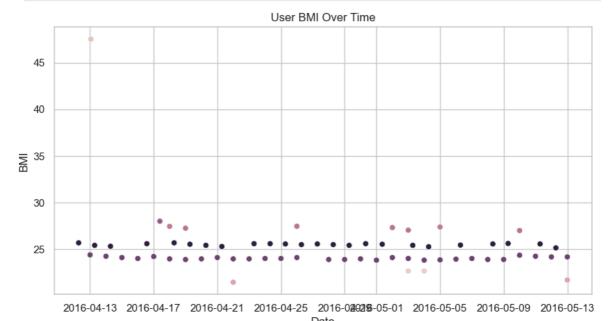
```
In [8]: # --- D. Heart Rate Trend (One User) ---
# Pick the user with the most heart rate records
top_user = heart['Id'].value_counts().idxmax()
user_heart = heart[heart['Id'] == top_user]

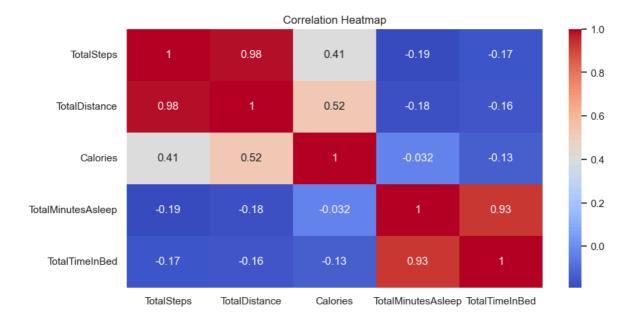
user_heart.set_index('Time')['Value'].plot()
plt.title(f"Heart Rate Trend (User ID: {top_user})")
plt.ylabel("Heart Rate (BPM)")
plt.xlabel("Time")
plt.show()
```

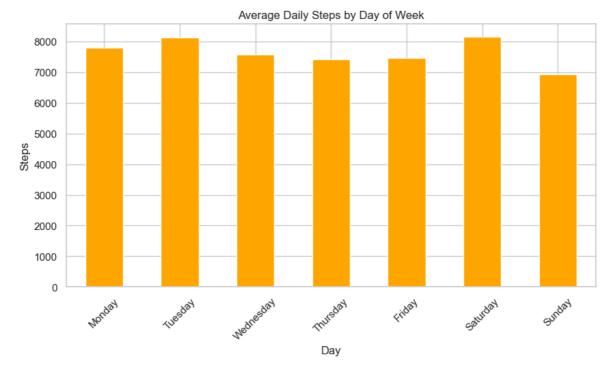


```
In [9]: # --- E. Weight Trends ---
# Only include weight Logs with BMI
weight_clean = weight.dropna(subset=['WeightKg', 'BMI'])

sns.scatterplot(data=weight_clean, x='Date', y='BMI', hue='Id', legend=False)
plt.title("User BMI Over Time")
plt.ylabel("BMI")
plt.xlabel("Date")
plt.show()
```







In []:

Bellabeat STRAVA Fitness Case Study



Analyze Fitbit/Strava user data to understand behavior patterns (steps, sleep, heart rate, weight) and recommend improvements for Bellabeat's women's wellness products.

Key Insights

- Users with higher daily steps burn more calories (positive correlation)
- Most users average between 6–8 hours of sleep
- Daily step count drops significantly on weekends
- Heart rate and BMI show user-level variation opportunity for personalized health insights

Recommendations for Bellabeat

- 1. Promote weekend activity campaigns
- 2. Encourage consistent sleep routines
- 3. Use personalized feedback (BMI, HR trends) for user engagement
- 4. Target low-activity users with nudges/reminders

Deliverables

- Python .ipynb file (this notebook)
- Exported .pdf of the notebook (via browser)
- Visuals: steps, sleep, HR, BMI, correlation, weekday patterns