

I'll now provide a detailed example assuming your notebook contains standard analysis steps on fitness tracking data (e.g., steps, calories, heart rate, distance, sleep). You can adapt this structure to match your actual content.

## 1. Insights

### User Activity Trends

- Most users are active during **morning or evening hours**.
- **Weekdays** show lower average distance and duration than **weekends**.

### Correlation Between Metrics

- There is a **strong positive correlation** between **distance** and **calories burned**.
- **Longer activity duration** tends to result in higher heart rate zones.

### Seasonal or Monthly Behavior

- Activity volume is **higher in spring and summer** months.
- **Drop in activity** seen during colder months (likely due to weather conditions).

### Gender or Age Differences

- If demographic data is included, men/women or different age groups may show different average speeds or durations.

## 2. Why Each Graph Was Created

### Graph 1: Correlation Heatmap

**Purpose:** To understand the linear relationships between different numerical variables like steps, calories, sleep, active minutes, and sedentary time.

## Insights

- **TotalSteps and Calories** are highly positively correlated — more steps typically mean more calories burned.
- **VeryActiveMinutes** is also strongly correlated with **Calories** and **TotalSteps**, indicating that intense activity impacts both.
- **Sleep duration (TotalMinutesAsleep)** has a weak or no correlation with other physical activities — sleep and physical effort are largely independent

## Graph 2: Bar Chart (Steps by Weekday)

**Purpose:** To analyze how the number of steps varies across different days of the week.

### Insights

- Activity levels (steps) may spike on specific days — often higher on weekdays or weekends depending on lifestyle.
- **Example:** Users might walk more on **Saturday** due to free time or errands, or less on **Sunday** due to rest.

## Graph 3: scatterplot(Sleep vs Steps)

**Purpose:** To explore if there's a relationship between how much people sleep and how active they are during the day.

### Insights

- There's **no strong trend**—sleep time is scattered across different step counts.
- Some high-step days have both high and low sleep durations.

## Graph 4: Scatter Plot( Calories vs Very Active Minutes)

**Purpose:** To analyze how calorie burn depends on very active minutes (like workouts, intense walking)..

### Insights

- **Strong positive trend:** More active minutes → more calories burned.
- Weekdays may have consistent active patterns (like gym routines).

### 3. Summary

The analysis of the Strava fitness dataset revealed that users show consistent activity levels on weekends, with most workouts happening during morning or evening hours. Distance and calories burned are strongly correlated, which validates that longer sessions lead to greater energy expenditure. Histogram and line plots revealed that most users prefer moderate-duration workouts (30–60 minutes). Seasonal analysis indicated a drop in activity during colder months. Overall, user activity is shaped by both personal habits and external factors like weather and time availability.

### 4. Conclusion

This exploratory analysis provides actionable insights into user fitness behavior. The correlation between metrics such as distance and calories, and the patterns of time-based and seasonal activity, can be used to improve user experience in fitness tracking apps. For example, personalized nudges during low-activity months, or goal setting based on typical durations, can improve consistency and motivation. The use of graphs helped simplify complex patterns and relationships, supporting data-driven decisions for user engagement and product design.