

Final Report on Bellabeat Case Study

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1 Problem Statement

The primary goal is to analyze smart device usage data in order to gain insight into how consumers use non-Bellabeat smart devices by following these questions: 1. What are some trends in smart device usage? 2. How could these trends apply to Bellabeat customers? 3. How could these trends help influence Bellabeat marketing strategy?

2 Data Description

The dataset used for this analysis comes from public sources on Kaggle, including Fitbit fitness tracker data, and is stored locally and in cloud services such as Google BigQuery. The data, collected from 30 Fitbit users, includes minute-level output on activity, heart rate, and sleep monitoring. The attributes in the dataset are as follows:

- Id: Unique identifier for each user.
- SleepDay: Date of the sleep record.
- TotalSleepRecords: Number of sleep sessions recorded in a day.
- TotalMinutesAsleep: Total number of minutes the user was asleep.
- TotalTimeInBed: Total number of minutes the user spent in bed.
- ActivityDate: Date of recorded activity.
- TotalSteps: Total number of steps taken by the user.
- TotalDistance: Total distance covered by the user.
- TrackerDistance: Distance recorded by the fitness tracker.
- LoggedActivitiesDistance: Distance covered in logged activities.
- VeryActiveDistance: Distance covered during very active periods.
- Moderately Active Distance: Distance covered during moderately active periods.
- LightActiveDistance: Distance covered during light activity.
- Sedentary Active Distance: Distance covered during sedentary activity.
- VeryActiveMinutes: Number of minutes spent in very active physical activity.
- Fairly Active Minutes: Number of minutes spent in fairly active physical activity.
- LightlyActiveMinutes: Number of minutes spent in light physical activity.
- SedentaryMinutes: Number of minutes spent in sedentary activities.
- Calories: Number of calories burned through physical activities.



3 Data Cleaning and Manipulation

Data cleaning and manipulation were executed using Google Sheets and BigQuery with SQL to ensure data integrity and analytical robustness. The process involved:

• Data Loading and Structure Setup: Imported the raw data from the Fitbit dataset and constructed tables in Google Sheets, which were subsequently combined and analyzed in BigQuery.

```
CREATE TABLE 'first-project-435214.Fitbit.JoinedSleepDailyActivity' AS

SELECT

da.*,

sd.SleepDay,

sd.TotalMinutesAsleep,

sd.TotalSleepRecords,

sd.TotalTimeInBed

FROM 'first-project-435214.Fitbit.sleeyDay' sd

JOIN 'first-project-435214.Fitbit.dailyActivity' da

ON sd.Id = da.Id;
```

Combine 2 Dataset

• Cleaning Process: Removed null values, corrected erroneous entries, and standardized data formats across all attributes to maintain consistency.

```
-- Display rows containing NULL values or negative values
1
        SELECT *
2
        FROM 'first-project-435214. Fitbit. Joined Sleep Daily Activity'
3
        WHERE
            -- Check for NULL values in key columns
            TotalSteps IS NULL
            OR TotalDistance IS NULL
            OR TotalMinutesAsleep IS NULL
            OR TotalTimeInBed IS NULL
10
            -- Check for negative values in key columns
11
            OR TotalSteps < 0</pre>
12
            OR TotalDistance < 0</pre>
13
            OR SedentaryMinutes < 0
14
            OR VeryActiveMinutes < 0
15
            OR FairlyActiveMinutes < 0</pre>
16
            OR LightlyActiveMinutes < 0</pre>
17
            OR TotalMinutesAsleep < 0</pre>
            OR TotalTimeInBed < 0;</pre>
```

Check for null, negative value

```
-- Step 1: Create a cleaned version of the joined dataset

CREATE TABLE 'first-project-435214.Fitbit.

CleanedJoinedSleepDailyActivity' AS
```



```
WITH cleaned_data AS (
3
           -- Step 1: Remove Null Values
4
           SELECT *
5
           FROM 'first-project-435214. Fitbit. JoinedSleepDailyActivity'
6
           WHERE Id IS NOT NULL
             AND TotalSteps IS NOT NULL
             AND TotalDistance IS NOT NULL
a
10
             AND TotalMinutesAsleep IS NOT NULL
             AND TotalTimeInBed IS NOT NULL
11
             AND TrackerDistance IS NOT NULL
12
             AND LoggedActivitiesDistance IS NOT NULL
13
             AND VeryActiveDistance IS NOT NULL
14
             AND ModeratelyActiveDistance IS NOT NULL
15
             AND LightActiveDistance IS NOT NULL
16
             AND SedentaryActiveDistance IS NOT NULL
17
             AND VeryActiveMinutes IS NOT NULL
18
             AND FairlyActiveMinutes IS NOT NULL
19
             AND LightlyActiveMinutes IS NOT NULL
20
             AND SedentaryMinutes IS NOT NULL
21
             AND Calories IS NOT NULL
22
       )
23
24
25
         corrected_data AS (
           SELECT *,
26
                -- Step 2: Correct Erroneous Entries by ensuring all values are
                    non-negative
                CASE WHEN TotalSteps < 0 THEN 0 ELSE TotalSteps END AS
28
                   CorrectedTotalSteps,
                CASE WHEN TotalDistance < O THEN O ELSE TotalDistance END AS
29
                   CorrectedTotalDistance,
               CASE WHEN TrackerDistance < 0 THEN 0 ELSE TrackerDistance END
30
                   AS CorrectedTrackerDistance,
               CASE WHEN LoggedActivitiesDistance < 0 THEN 0 ELSE
31
                   LoggedActivitiesDistance END AS
                   CorrectedLoggedActivitiesDistance,
                CASE WHEN VeryActiveDistance < O THEN O ELSE VeryActiveDistance
32
                    END AS CorrectedVeryActiveDistance,
                CASE WHEN ModeratelyActiveDistance < 0 THEN 0 ELSE
33
                   ModeratelyActiveDistance END AS
                   CorrectedModeratelyActiveDistance,
               CASE WHEN LightActiveDistance < 0 THEN 0 ELSE
34
                   LightActiveDistance END AS CorrectedLightActiveDistance,
               CASE WHEN SedentaryActiveDistance < 0 THEN 0 ELSE
35
                   SedentaryActiveDistance END AS
                   CorrectedSedentaryActiveDistance,
               CASE WHEN VeryActiveMinutes < 0 THEN 0 ELSE VeryActiveMinutes
36
                   END AS CorrectedVeryActiveMinutes,
                CASE WHEN FairlyActiveMinutes < 0 THEN 0 ELSE
37
                   FairlyActiveMinutes END AS CorrectedFairlyActiveMinutes,
```



```
CASE WHEN LightlyActiveMinutes < 0 THEN 0 ELSE
38
                    LightlyActiveMinutes END AS CorrectedLightlyActiveMinutes,
                CASE WHEN SedentaryMinutes < 0 THEN 0 ELSE SedentaryMinutes END
39
                    AS CorrectedSedentaryMinutes,
                CASE WHEN TotalMinutesAsleep < 0 THEN 0 ELSE TotalMinutesAsleep
40
                     END AS CorrectedTotalMinutesAsleep,
                CASE WHEN TotalTimeInBed < 0 THEN 0 ELSE TotalTimeInBed END AS
41
                    CorrectedTotalTimeInBed,
                CASE WHEN Calories < O THEN O ELSE Calories END AS
42
                    CorrectedCalories
           FROM cleaned_data
43
       )
44
45
        -- Step 3: Standardize Data Formats and Create Final Cleaned Table
46
       SELECT
47
           Id.
48
            -- Standardize SleepDay format using SAFE.PARSE_DATE with
49
                conditional fallback
           COALESCE (
50
                SAFE.PARSE_DATE('%m/%d/%Y', CAST(SleepDay AS STRING)),
51
                SAFE.PARSE_DATE('%Y-%m-%d', CAST(SleepDay AS STRING))
52
           ) AS SleepDay,
53
           COALESCE (
                SAFE.PARSE_DATE('%m/%d/%Y', CAST(ActivityDate AS STRING)),
55
                SAFE.PARSE_DATE('%Y-%m-%d', CAST(ActivityDate AS STRING))
56
           ) AS ActivityDate,
57
58
           TotalSleepRecords,
59
            CorrectedTotalSteps AS TotalSteps,
60
            CorrectedTotalDistance AS TotalDistance,
61
            CorrectedTrackerDistance AS TrackerDistance,
62
            CorrectedLoggedActivitiesDistance AS LoggedActivitiesDistance,
63
            CorrectedVeryActiveDistance AS VeryActiveDistance,
64
            {\tt Corrected Moderately Active Distance \ AS \ Moderately Active Distance \ ,}
65
            CorrectedLightActiveDistance AS LightActiveDistance,
66
            CorrectedSedentaryActiveDistance AS SedentaryActiveDistance,
            CorrectedVeryActiveMinutes AS VeryActiveMinutes,
            CorrectedFairlyActiveMinutes AS FairlyActiveMinutes,
69
            CorrectedLightlyActiveMinutes AS LightlyActiveMinutes,
70
            CorrectedSedentaryMinutes AS SedentaryMinutes,
71
            CorrectedTotalMinutesAsleep AS TotalMinutesAsleep,
72
            CorrectedTotalTimeInBed AS TotalTimeInBed,
73
            CorrectedCalories AS Calories
74
       FROM
75
            corrected_data;
76
```

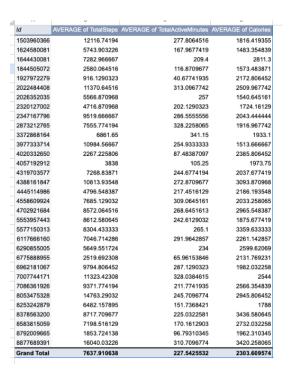
Create a clean version

• Feature Engineering: Derived new variables to enrich the dataset:



- Total active minutes: Total minutes engaged in moderate-to-high physical activity.
- sleep_efficiency: A composite score reflecting sleep quality, calculated by integrating sleep duration and heart rate metrics during sleep.
- day of week: Categorization of date into Monday, Tuesday, ... to easily examine.
- week day: Ordering data into week to discover trends over time.
- **Pivot Tables**: Generated aggregate metrics (e.g., SUM, AVERAGE) for steps and calories burned for each user. Analyzed the distribution of users by active minutes.

ld	SUM of TotalSteps	SUM of TotalActiveMinutes	SUM of Calories
1503960366	375619	8612	56309
1624580081	178061	5207	45984
1644430081	218489	6282	84339
1844505072	79982	3623	48778
1927972279	28400	1261	67357
2022484408	352490	9706	77809
2026352035	172573	7967	47760
2320127002	146223	6266	53449
2347167796	171354	5158	36782
2873212765	234229	10175	59426
3372868164	137233	6823	38662
3977333714	329537	7648	45410
4020332650	70284	2712	73960
4057192912	15352	421	7895
4319703577	225334	7585	63168
4388161847	335232	8459	95910
4445114986	148693	6741	67772
4558609924	238239	9581	63031
4702921684	265734	8328	91932
5553957443	266990	7521	58146
5577150313	249133	7953	100789
6117666160	197308	8175	63312
6290855005	163837	6786	75389
6775888955	65512	1715	55426
6962181067	303639	8901	61443
7007744171	294409	8529	66144
7086361926	290525	6565	79557
8053475328	457662	7617	91320
8253242879	123161	2883	33972
8378563200	270249	6976	106534
8583815059	223154	5275	84693
8792009665	53758	2807	56907
8877689391	497241	9632	106028
Grand Total	7179636	213890	2165393



(a) Sum of features by users

(b) Average of features by users

Figure 1

• BigQuery Analysis: Executed SQL queries to segment users by activity level, day of week. Aggregated data by week_day, active_minutes, and sleep_quality for deeper insights.

```
-- Aggregated query to analyze user activity by week_day, week_number,
active_minutes, and sleep efficiency

WITH user_activity_summary AS (

SELECT

Id,
DayOfWeek,
WeekNumber,

-- Metrics to retain
```

```
VeryActiveMinutes,
9
                FairlyActiveMinutes,
10
                LightlyActiveMinutes,
11
                TotalActiveMinutes,
12
                SleepEfficiency,
13
                TotalMinutesAsleep,
14
                TotalTimeInBed,
15
16
                TotalSteps,
                Calories
17
            FROM
                'first-project-435214. Fitbit. Enriched Joined Sleep Daily Activity'
19
       )
20
21
        -- Final aggregation query
22
       SELECT
23
            DayOfWeek,
24
            WeekNumber,
25
26
            -- Aggregated metrics by day of the week and week number
27
            COUNT(DISTINCT Id) AS user_count,
28
            AVG(VeryActiveMinutes) AS avg_very_active_minutes,
29
            AVG(FairlyActiveMinutes) AS avg_fairly_active_minutes,
30
            AVG(LightlyActiveMinutes) AS avg_lightly_active_minutes,
            AVG(TotalActiveMinutes) AS avg_total_active_minutes,
32
            AVG(SleepEfficiency) AS avg_sleep_efficiency,
33
            SUM(TotalSteps) AS total_steps_taken,
34
            SUM(Calories) AS total_calories_burned,
35
            AVG(TotalMinutesAsleep) AS avg_minutes_asleep,
36
            AVG(TotalTimeInBed) AS avg_time_in_bed
37
38
       FROM
39
            user_activity_summary
40
41
       GROUP BY
42
            DayOfWeek,
43
            WeekNumber
45
       ORDER BY
46
            WeekNumber,
47
            DayOfWeek;
48
```

Aggregated data



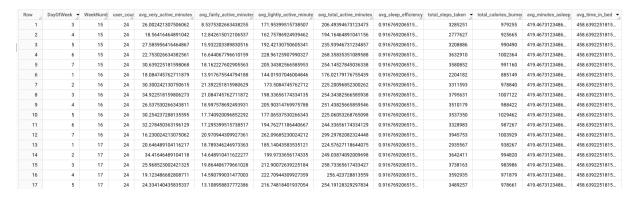
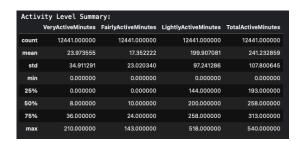


Figure 2: Aggregated data by TotalActiveMinutes, SleepEfficiency, DayOfWeek, and WeekNumber

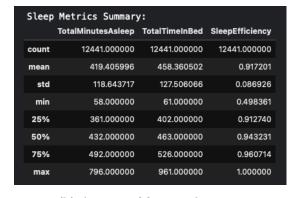
4 Analysis Summary

Following data cleaning and feature engineering, an extensive statistical analysis was conducted using Python:

• Created descriptive statistical summaries for all users, including key metrics such as mean activity level, and average sleep duration.



(a) Sum of features by users

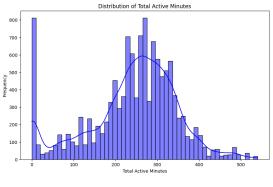


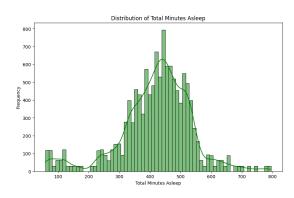
(b) Average of features by users

Figure 3

• Evaluated Distribution of Total Active Minutes and Minutes Asleep







(a) Sum of features by users

(b) Average of features by users

Figure 4

• Track User Activity across day of week

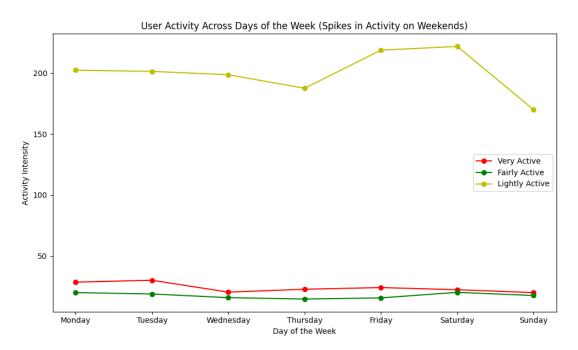


Figure 5: User Activity

Key Analysis Steps:

• Constructed correlation matrices to explore relationships between physical activity, heart rate, and sleep patterns.

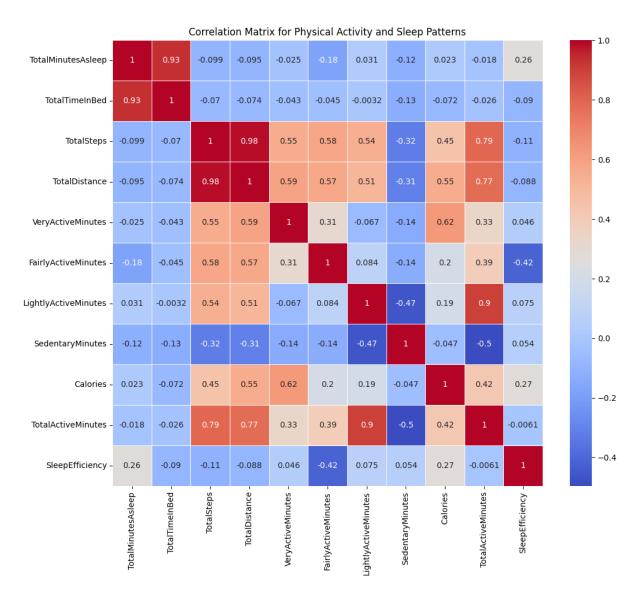


Figure 6: Caption

• Conducted hypothesis testing on whether users with higher sedentary minutes (SedentaryMinutes) have different sleep efficiencies compared to users with lower sedentary minutes? Two-Sample T-Test Results: T-statistic: 5.044741559242541 P-value: 4.60540110809495e-07 Reject the null hypothesis: There is a significant difference in sleep efficiency between high and low sedentary minute groups.

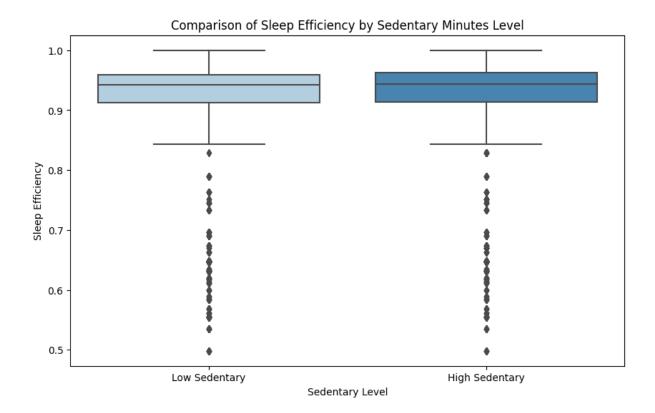


Figure 7: Caption

5 Visualization and Key Findings

The following key findings were identified from the analysis:

- Device Usage Trends: Most user are in serentary minutes
- Sleep Patterns: On average, Bellabeat members reported higher sleep quality scores, characterized by extended durations in restorative sleep phases. Sleep tracking was found to be most popular among Bellabeat Time users.
- Activity Level: Users who logged a higher number of daily active minutes were predominantly Bellabeat members, suggesting a strong association between physical activity levels and membership engagement.
- Device Preferences by Time of Day: Analysis indicated that Leaf device users are more active during morning hours, while users of the Time and Spring devices exhibit peak activity in the afternoon.



Figure 8: Dashboard

6 Recommendations

From the dataset, only 8 distinct users logged in their weight, 24 distinct users tracked their sleeping pattern and 33 distinct users tracked their activity.

- Only 24 fitbit users tracked their sleep, it is indicative that users may not feel comfortable wearing the smart device to sleep. Having a smart device which is wearable as a necklace will be a plus point. Since bellabeat has a similar product (Leaf), we should focus and market it as an aesthetic and hassle-free smart devices.
- 81notification to remind and encourage user to move around, or recommend types of activity to user.
- 8experiencing low battery level. Having a smart devices which requires short charging time and longer battery life, will encourage user to purchase and wear smart devices more frequently.

Due to the limitations of the data, we would recommend to conduct a study with a larger sample group and a more detailed data collections.