

# Analyzing Mobility Trends: Apple and Google Data Insights During the COVID-19 Pandemic

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**Abstract:** *In this paper, we use publicly accessible Apple and Google mobility data to explore the substantial changes in human movement behavior during COVID-19. It reveals the temporal and spatial evolution of population movement across hundreds of regions in relation to government public health orders. This underscores the importance of using real-time mobility data to inform public health interventions, urban planning and response effort during a global pandemic.*

**Keywords:** COVID-19, mobility trends, Apple mobility data, Google mobility data, public health, pandemic response, urban planning, data analysis, geographic information systems.

## 1. Introduction

The COVID-19 pandemic has induced a mode of external mobility not hitherto noticed, as several governments have executed lockdowns and travel bans to harness the contagion. Apple and Google have released aggregate data depicting the movement of people in different geographies, as they acknowledge that it is important for decision makers to track this change. This study investigates these mobility trends to elucidate the heterogeneity of human mobility patterns under different levels and types of public health directives, followed by an examination on how those movements impacted the control measures concerning COVID-19.

## 2. Problem Statement

The extensive datasets released by Apple and Google offer invaluable insights into global mobility patterns during the pandemic. However, extracting actionable intelligence from

these large datasets is challenging due to their complex nature and the nuanced understanding required to interpret them correctly. This research aims to decode these mobility trends, assess their reliability, and explore the implications of mobility changes for public health policies and economic outcomes.

Below links are the source for analyzing mobility data.



<https://covid19.apple.com/mobility>



<https://www.google.com/covid19/mobility/>

### Solution Implemented:

For this analysis, mobility data from Apple and Google were initially obtained in CSV format, representing a broad spectrum of movement trends across various geographic locations and time frames. This data was first loaded into raw tables within a MySQL database environment, setting up a structured schema to facilitate easy access and manipulation. The loading process was automated using a Python script, which was developed to not only perform the initial data

ingestion but also handle daily incremental updates, ensuring that the dataset remained current with the latest mobility trends.

Both datasets from Apple and Google were normalized to a baseline value of 100 to facilitate a comparative analysis of mobility trends during the pandemic. This normalization allows for the measurement of mobility patterns as percentages relative to the baseline, effectively illustrating

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whether the activity levels were below, at, or above pre-pandemic norms. Such an approach offers clear insights into the extent of changes in movement during different phases of the pandemic, providing a quantitative method to assess deviations from usual mobility behaviors.

Raw Data Sets:

- *Apple Mobility*

```
select geo_type, region, transportation_type, `sub-region`, country, `date`, volume from apple_mobility_v2
```

| geo_type       | region                      | transportation_type | sub-region             | country        | date       | volume |
|----------------|-----------------------------|---------------------|------------------------|----------------|------------|--------|
| country/region | Finland                     | walking             | [NULL]                 | [NULL]         | 2020-04-28 | 70.97  |
| city           | Bridgeport                  | driving             | Connecticut            | United States  | 2020-04-28 | 70.41  |
| sub-region     | Borsod-Abaúj-Zemplén County | driving             | [NULL]                 | Hungary        | 2020-04-28 | 86.01  |
| county         | Adair County                | driving             | Missouri               | United States  | 2020-04-28 | 81.74  |
| county         | Arlington County            | driving             | Virginia               | United States  | 2020-04-28 | 43.66  |
| county         | Benton County               | driving             | Iowa                   | United States  | 2020-04-28 | 115.69 |
| county         | Bonner County               | driving             | Idaho                  | United States  | 2020-04-28 | 136.2  |
| county         | Clark County                | driving             | Wisconsin              | United States  | 2020-04-28 | 89.94  |
| county         | Dickson County              | driving             | Tennessee              | United States  | 2020-04-28 | 97.26  |
| county         | Mifflin County              | driving             | Pennsylvania           | United States  | 2020-04-28 | 88.26  |
| county         | Minidoka County             | driving             | Idaho                  | United States  | 2020-04-28 | 148.33 |
| county         | Nacogdoches County          | driving             | Texas                  | United States  | 2020-04-28 | 73.93  |
| county         | Nottoway County             | driving             | Virginia               | United States  | 2020-04-28 | 91.13  |
| county         | Socorro County              | driving             | New Mexico             | United States  | 2020-04-28 | 54.64  |
| county         | Washington County           | driving             | New York               | United States  | 2020-04-28 | 93.28  |
| county         | York County                 | driving             | Nebraska               | United States  | 2020-04-28 | 95.75  |
| country/region | Norway                      | walking             | [NULL]                 | [NULL]         | 2020-04-28 | 72.91  |
| city           | Aachen                      | driving             | North Rhine-Westphalia | Germany        | 2020-04-28 | 79.04  |
| city           | Cardiff                     | transit             | Wales                  | United Kingdom | 2020-04-28 | 13.59  |
| city           | Boston                      | transit             | Massachusetts          | United States  | 2020-04-28 | 19.41  |

• *Google Mobility*

```
select country_region_code, country_region, sub_region_1, date, retail_and_recreation_percent_change_from_baseline retail_and_recreation_percent_change, grocery_and_pharmacy_percent_change_from_baseline retail_and_recreation_percent_change, parks_percent_change_from_baseline parks_percentage_change from google_mobility
```

| country_region_code | country_region       | sub_region_1      | date       | retail_and_recreation_percent_change | retail_and_recreation_percent_change | parks_percentage_change |
|---------------------|----------------------|-------------------|------------|--------------------------------------|--------------------------------------|-------------------------|
| AG                  | Antigua and Barbuda  | [NULL]            | 2020-02-17 | 0                                    | -3                                   | 3                       |
| AE                  | United Arab Emirates | Fujairah          | 2020-02-17 | -7                                   | -2                                   | -4                      |
| AE                  | United Arab Emirates | Dubai             | 2020-02-17 | 0                                    | 1                                    | 7                       |
| AR                  | Argentina            | Buenos Aires      | 2020-02-17 | -12                                  | -12                                  | -29                     |
| AE                  | United Arab Emirates | Ajman             | 2020-02-17 | -1                                   | -1                                   | 3                       |
| AF                  | Afghanistan          | [NULL]            | 2020-02-17 | 1                                    | 11                                   | -3                      |
| AR                  | Argentina            | Buenos Aires      | 2020-02-17 | -12                                  | -15                                  | -42                     |
| AR                  | Argentina            | Buenos Aires      | 2020-02-17 | -9                                   | -14                                  | -27                     |
| AF                  | Afghanistan          | [NULL]            | 2020-02-17 | 6                                    | 11                                   | 2                       |
| AR                  | Argentina            | Buenos Aires      | 2020-02-17 | -11                                  | -13                                  | -29                     |
| AR                  | Argentina            | Buenos Aires      | 2020-02-17 | -7                                   | 3                                    | -14                     |
| AR                  | Argentina            | Buenos Aires      | 2020-02-17 | -6                                   | -15                                  | -8                      |
| AE                  | United Arab Emirates | Umm Al Quawain    | 2020-02-17 | 0                                    | 3                                    | [NULL]                  |
| AO                  | Angola               | Huambo Province   | 2020-02-17 | [NULL]                               | [NULL]                               | [NULL]                  |
| AG                  | Antigua and Barbuda  | Saint George      | 2020-02-17 | [NULL]                               | [NULL]                               | [NULL]                  |
| AO                  | Angola               | Benguela Province | 2020-02-17 | [NULL]                               | [NULL]                               | [NULL]                  |
| AO                  | Angola               | Huila Province    | 2020-02-17 | [NULL]                               | [NULL]                               | [NULL]                  |
| AG                  | Antigua and Barbuda  | Saint John        | 2020-02-17 | -1                                   | -1                                   | -1                      |
| AE                  | United Arab Emirates | [NULL]            | 2020-02-17 | -1                                   | 1                                    | 5                       |
| AE                  | United Arab Emirates | Abu Dhabi         | 2020-02-17 | -3                                   | 2                                    | 4                       |

To make the data actionable and easily accessible, an application programming interface (API) was designed and implemented. This API facilitated the extraction and aggregation of data across different streams, allowing users

within the app to view and interact with the mobility data in a user-friendly manner. Through the app, data could be visualized in various formats, enabling stakeholders to track

mobility trends effectively and make informed decisions based on real-time data.

- Driving
- Walking
- Transit

Apple Mobility data was tracked across three different mobility patterns or streams i. e.



Google Mobility data was tracked across six different mobility patterns or streams i. e.

- Transit
- Retail
- Grocery
- Parks
- Workplaces
- Residential



For the oil and gas industry, understanding mobility trends during the COVID-19 pandemic was crucial, particularly in forecasting the demand for jet fuel, which saw a dramatic decrease due to global travel restrictions. The industry needed to predict when and how quickly demand might return to pre-pandemic levels as mobility restrictions lifted and economic activities resumed. By analyzing mobility data, especially

trends related to driving and transit use, the industry could gain insights into the pace of economic recovery and the potential rebound in travel activity. This was vital for planning production, refining operations, and logistic strategies to align with anticipated increases in fuel demand. Effective use of this mobility data allowed the industry to optimize resource allocation, reduce operational risks, and

strategically plan for market re-entry as conditions normalized.

#### **Potential Extended Use Cases:**

The techniques created and the observations drawn from analysis of pandemic-related mobility data can be applied more broadly in other parts. An example of this is the adaptation to urban planning, where traffic can be optimized, or a better transit system implemented from public data analysis. Moreover, economic analysts can use mobility trends to forecast the degree of economic activity in many sectors, which is crucial for policymaking as well as investment decisions.

### **3. Impact**

It demonstrates a unique direct link between mobility reductions and decreased virus transmission, confirming that when lockdown measures are strictly implemented, they enable effective epidemic management. It also shows how movement in economic sectors has been affected differently with retail and recreation activity down substantially but residential visits stable or increased. The study emphasizes the value of real-time mobility data for use in public health planning and its function as part of a trade-off between healthcare risk circulated with economic cost.

### **4. Scope**

This analysis uses data from Apple and Google, which provide the most extensive tracking of mobility patterns but may not represent all forms of movement or demographic groups. Data collected includes transportation trends—those walking, driving or taking transit in different countries and cities—but does not provide detailed insight on individual adherence to directives nor situational social-socioeconomic factors affecting mobility.

### **5. Conclusion**

The conclusion synthesizes the insights that mobility data can provide regarding health of a population as well as how this information would have been actionable for public health and emergency preparedness. Highly recommended for all governments, campaigners, and social scientists aiming to ride the curve of digital traces in governance.

### **References**

- [1] Apple COVID-19 Mobility Trends Reports, Available at <https://covid19.apple.com/mobility>
- [2] Google COVID-19 Community Mobility Reports, Available at <https://www.google.com/covid19/mobility/>
- [3] How to load CSV data using python, Available at <https://docs.python.org/3/library/csv.html>