

Part 1: Supply Data Analysis

Important part of our business is a supply/demand balance. We can't control demand but we can shift some supply to necessary hours to cover more demand during peaks.

As part of the task you will have sample **supply** and **demand** data over a few weeks in a single city a few weeks after launch.

We need to understand:

- 1) What is the supply to demand dynamic and whether they match?
- 2) Where are the hours of oversupply? Can we shift some of them to undersupply hours?

Needed output:

1. Show which 36 hours in a week are most undersupplied. Show/describe your decision based on sample data.
2. 24-hour curve of average supply and demand (to illustrate match/mismatch).
3. Visualisation of hours where we lack supply during a weekly period. This one we can send to drivers to show when to online for extra hours.
4. Estimate number of hours needed to ensure we have a high Coverage Ratio during most peak hours.
5. Calculate levels of guaranteed hourly earnings we can offer to drivers during 36 weekly hours with highest demand without losing money + how much extra hours we want to get to capture missed demand.
 - a. Assume that Finished Rides have an average value of €10 (80% goes to driver, 20% is our revenue).
 - b. Assume the same level of demand with increased supply, base it on RPH over 3 hour periods, but with increased supply.
 - c. Assume that with extra hours we will capture "missed coverage" or people attributed to "People saw 0 cars" in demand data.

[Sample Supply data info \(Hourly DriverActivity.csv\)](#)

Shows aggregated driver activity in a city for a given period.

Details:

- Hourly data for 5 full weeks from 2016-11-14 until 2016-12-18
- Real data from a recent launch in a competitive city (2 big apps for years)
- Fields:
 - **Date** – date + hour for which the row of data is presented
 - **Active drivers** – number of active drivers (any level of activity) available during time period
 - **Online (h)** – total supply hours that were available during time period

- **Has booking (h)** – total hours during which drivers had a client booking (any state)
- **Waiting for booking (h)** – total hours which drivers spent waiting for booking
- **Busy (h)** – total hours which drivers were not available to take orders in
- **Hours per active driver** – average number of hours each driver was online during time period
- **Rides per online hour** – aka **RPH** – avg. finished trips per online hour during period
- **Finished Rides** – number of finished trips during period
- Note the data is sorted with more recent data first
- Note that is time period has 0 values in all columns, it would be skipped (no row)

[Sample Demand data info \(Hourly OverviewSearch.csv\)](#)

Shows how many people saw a car in the app when setting the pickup marker on the map. If you saw a car at one point and did not see a car later, you are counted in both columns in that period.

Details:

- The data is from the same period as Supply data above.
- Fields:
 - **Date** – date + hour for which the row of data is presented
 - **People saw 0 cars (unique)** – number of users who didn't not see a car.
 - **People saw +1 cars (unique)** – number of users who saw a car.
 - **Coverage Ratio (unique)** – % of users who saw the car.
- Note the data is sorted with more recent data first
- Note that is time period has 0 values in all columns, it would be skipped (no row)