# KAGGLE – Airbnb prices in European cities

Project number: H8

# Setting up

Our repository is hosted in GitHub and can be found at this link: https://github.com/ElisVingisar/IDS-project

### Business understanding

Identifying the business goals

#### Background:

In the hospitality industry, Airbnb has become a prominent player, offering a diverse range of accommodations across European cities. For the hosts or customers themselves, it's crucial to understand the factors that influence Airbnb prices. The 20 datasets that we have been provided with, representing 10 European cities with weekday and weekend pricing details, provide a great source of information for comprehensive analysis.

#### Business goals:

- 1. Price analysis across cities:
  - Our first goal is to understand and analyze how the prices between the 10 given European cities vary. We aim to identify the overall prices in the 10 cities under observation, which could be useful for future travelers in deciding where to go depending on their budget.
- 2. Weekday vs. weekend price disparities:
  - The second goal is to examine the variations in Airbnb prices between weekdays and weekends for each of the cities. This will help us identify the city with the biggest price difference, depending on whether it is a weekday or weekend, of the 10 European cities.
- 3. Price prediction model:
  - The final goal of the project is to predict one of the cities accommodation's price based on it's attributes. A reliable pricing prediction can benefit hosts by setting competitive rates and even assisting travelers in budgeting for their trips.

Business success criteria

The project is considered successful when

- we have identified the cities with highest and lowest housing prices,
- we have gained insight on which of the cities housing is most affected by the time when making a housing reservation through Airbnb,

#### Assessing our situation

#### Inventory of Resources

```
1. Data:
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amsterdam\_weekdays.csv amsterdam\_weekends.csv athens\_weekdays.csv

athens\_weekends.csv

barcelona\_weekdays.csv

barcelona\_weekends.csv

berlin\_weekdays.csv

berlin\_weekends.csv

budapest\_weekdays.csv

budapest\_weekends.csv

lisbon\_weekdays.csv

lisbon\_weekends.csv

london\_weekdays.csv

london\_weekends.csv

paris\_weekdays.csv

paris\_weekends.csv

rome\_weekdays.csv

rome\_weekends.csv

vienna\_weekdays.csv

vienna\_weekends.csv

2. Other: Our main contact in case of any questions or problems would be our lab supervisor Carel Kuusk.

#### Requirements, assumptions and constraints

- 1. Requirements: Most importantly we need access to Airbnb pricing data and computational resources for model development. Planning related requirement is to submit the project before the deadline 11th December 2023.
- 2. Assumptions: The dataset is representative of Airbnb listings in the specified 10 cities. Prices are influenced by a combination of factors as outlined in the dataset.
- 3. Constraints: Limited to no ability to factor in external events affecting prices.

#### Risks and Contingenceis:

- 1. Risk: Incomplete or inaccurate data in the dataset and model overfitting.
- 2. Contingency: Implementing rigorous data cleaning processes.

#### Terminology

Superhost – In Airbnb, a host is someone who is providing the housing. Superhosts on the other hand tend to have a higher occupancy rate (and more potential earnings) on average because they've met the Airbnb criteria of becoming a Superhost.

#### Costs and benefits

- 1. Cost: Time invested into completing the project
- 2. Benefit: Applying oneself, wider knowledge, grade based on the project outcome. Results of the project can be useful for future travelers choosing housing through Airbnb or for the hosts to estimate the total price per night of the housing.

Other than that, we do not need to assess any more costs and benefits.

Defining data-mining goals

Data-mining goals

Our goals are the following:

- to develop a predictive model to estimate accommodation prices (mainly focused on the location),
- to deliver pricing differences on various graphs and therefore find out more patterns.

Data-mining success criteria

The data mining is considered successful when

- we can make accurate predictions based on the accommodations factors,
- we can report on the cities overall accommodation's pricings.

### Data understanding

#### **Gathering Data**

Data requirements outline

The data we need must contain information, whether the accommodation's price is set for a weekday or weekend. Also, we need to know the main specifics about the accommodations to make different and accurate predictions – total cost, number of rooms, if the room is private or not, person capacity, distance from nearest public transportation possibility, etc.

Data availability verification

The data we have acquired is from Kaggle and has been verified.

Selection criteria

The data we are using is in CSV file format. We have 20 CSV files which contain information about 10 cities accommodations pricings. For each city there is two datasets – weekday prices and weekend prices. For example, 1 of the 10 European cities is Amsterdam. The according datasets are:

amsterdam\_weekdays.csv

amsterdam\_weekends.csv

All of the datasets contain the same information:

Column name	Description			
realSum	The total price of the Airbnb listing.			
room_type	The type of room being offered (e.g. private, shared, etc.).			
room_shared	Whether the room is shared or not.			
Toom_sharea	Whether the room is shared or not.			
room_private	Whether the room is private or not.			
person_capacity	The maximum number of people that can stay in the room.			
host_is_superhost	Whether the host is a superhost or not.			
multi	Whether the listing is for multiple rooms or not.			
biz	Whether the listing is for business purposes or not.			
cleanliness_rating	The cleanliness rating of the listing.			
guest_satisfaction_overall	The overall guest satisfaction rating of the listing.			

bedrooms	The number of bedrooms in the listing.		
dist	The distance from the city centre.		
metro_dist	The distance from the nearest metro station.		
Ing	The longitude of the listing.		
lat	The latitude of the listing.		

# Describing data

The data is from Kaggle and it seems to have been gathered since the beginning of 2023 (no exact date range stated in Kaggle).

Column name	Format		
realSum	Numeric		
room_type	Categorical – Entrie home/apt, Private room, Other		
room_shared	Boolean – true, false		
room_private	Boolean – true, false		
person_capacity	Numeric		
host_is_superhost	Boolean – true, false		
multi	Numeric – 1, 0		
biz	Numeric – 1, 0		
cleanliness_rating	Numeric (max 10.0)		
guest_satisfaction_overall	Numeric (max 100.0)		
bedrooms	Numeric		
dist	Numeric		
metro_dist	Numeric		
Ing	Numeric		
lat	Numeric		

### Data quality verification

The data we have accuired should be suitable for our project and we haven't found any data errors yet.

# Project plan

Task	Tools	Elis	Rahel
Homework 10	Google Docs	1h	4h
Research about	Google, Kaggle,	3h	2h
previous work done in	Microsoft Excel		
that area and			
exploring more ideas			
for presenting the			
outcome			
Decide on final	Google, Microsoft	2h	0h
approaches	Excel		
Data exploration or	Jupyter Notebook,	4h	3h
cleaning	Microsoft Excel		
Data visualisation	Jupyter Notebook	6h	6h
Data training for	Jupyter Notebook	8h or more	8h or more
predictions			
Results analysis	Microsoft Excel,	4h or more	4h or more
	Jupyter Notebook		
Poster	Canva, Google'i	3h or more	3h or mroe
	joonised		