



Data Analytics Portfolio

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Project Tools

Games Co.
Video Game Analysis

Tools Utilized:



MSA
Medical Staffing
Analysis

Tools Utilized:



Rockbuster
Online Video Rental
Service Analysis

Tools Utilized:



Instacart
Grocery Delivery
Analysis

Tools Utilized:



Airbnb
Munich Case Study

Tools Utilized:



Games Co. Video Game Analysis



Games Co.Video Game Analysis

▶ Background

Games Co. is a fictional video game company. It aims to enter global competitive video game market. Using data Games co. wants to determine their marketing budget for the year 2017.

Note: Games Co. is a fictional company with real data given

▶ Key Questions

- ▶ Are certain types of games more popular than others?
- ▶ Allocation of the marketing budget for the year 2017 – same budget for every region?
- ▶ How have their sales figured varied between geographic regions over time? – “Sales for various geographic region have stayed th same over time’

▶ Data

- ▶ The dataset contains historical sales of video games over 100.000 copies. Spanning different platforms, genres and publishing studios spanning the EU, NA, Japan and Others.
- ▶ Vgchartz

▶ Tools & Skills



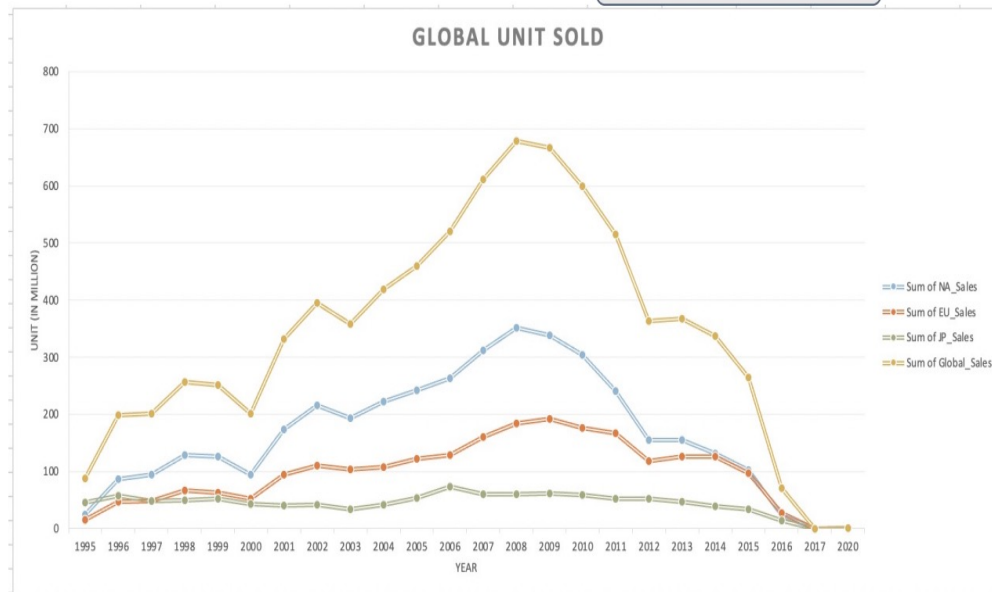
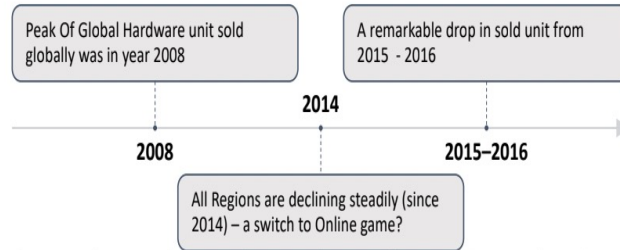
- ▶ Data Cleaning
- ▶ Grouping & Summarizing Data
- ▶ Descriptive Analysis
- ▶ Pivot Tables
- ▶ Visualizations



- ▶ Creating & Presenting Analytical Finding

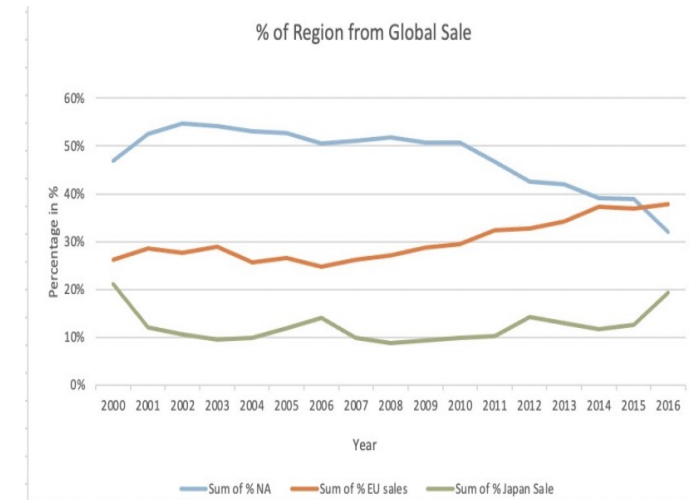
Games Co.Video Game Analysis

Comparison of Global Unit Sold in each Region



Comparison of % of Regional Sale from Global Sale

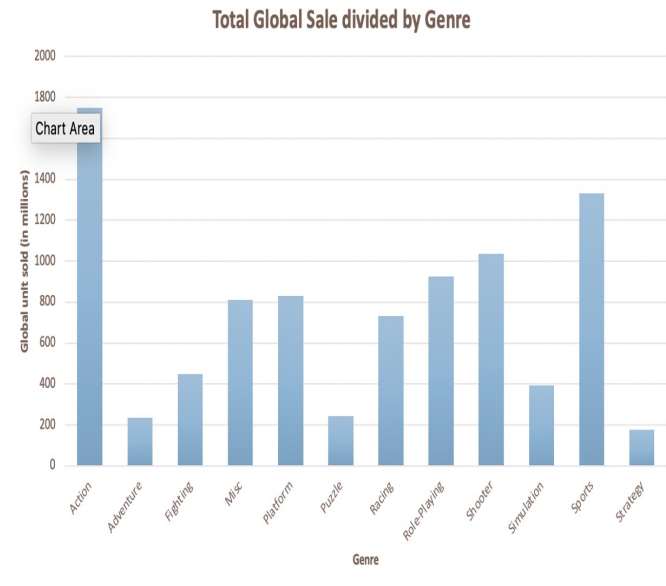
- North America was the market leader until recently being taken over by Europe
- Japan's market share has been steady since 2000
- Europe is on a growth phase on the global market share



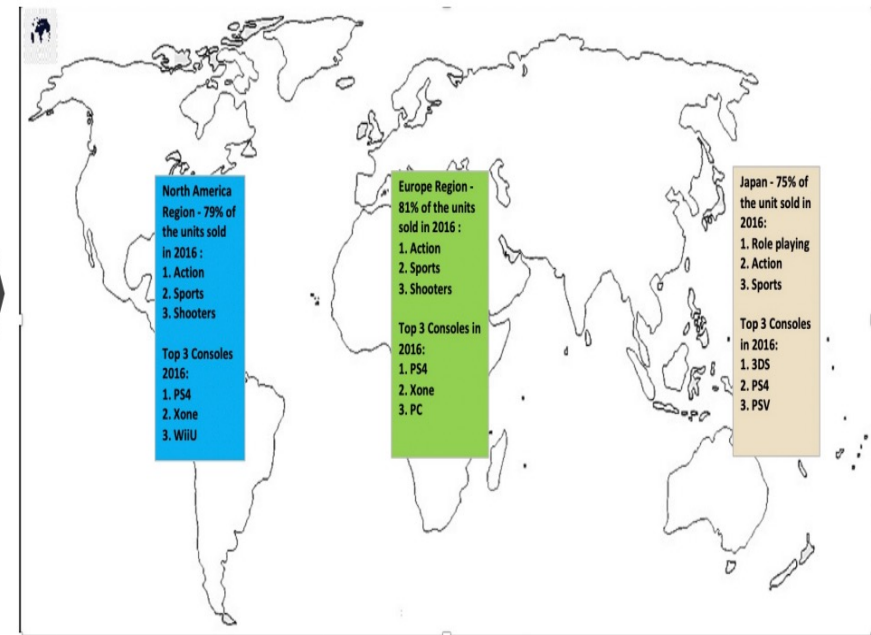
Games Co.Video Game Analysis

Variable: Genres

- 4 Genres are making over 57% of total global unit sold.
- Action, Sports, Shooter and Role – Playing are genre with the most units sold on a global level



Regional
analysis on
Genre &
Console



Games Co. Data Analysis

Recommendation for Marketing Budget Allocation

- All three regions are on different phases of growth. North America is declining, Europe is in a growth phase and Japan is stagnating.
- Each region is unique, so a standardized marketing won't be working on our side (every region has a different preferences in genres & consoles).
- Higher budget should be allocated for the European Region to accelerate growth and gain the momentum, For the North America and Japan we need to allocate budget to maintain our market share.

MSA Medical Staffing Analysis



MSA

Medical Staffing Analysis

► Background

The MSA (Medical Staffing Agency) is a fictional company that needs help in temporarily placing workers for the upcoming 2018 Influenza Season.

► Motivation

The United States has an influenza season where more people than usual suffer from the flu. Some people, particularly those in vulnerable populations, develop serious complications and end up in the hospital. Hospitals and clinics need additional staff to adequately treat these extra patients. The medical staffing agency provides this temporary staff.

► Objective

Determine when to send staff, and how many to each State

► Scope

The agency covers all hospitals in each 50 States in the USA, and the project will plan for the upcoming influenza season

► Data

The following data set covering influenza in the USA were used during the project:

- Influenza deaths by geography, time, age and gender, Source: CDC
- Population data by geography. Source: US Census Bureau
- Count of influenza laboratory test result by state and Influenza visit (survey). Source: CDC (Fluview)
- Survey of flu shot rates in children, Source: CDC

► Tools & Skills

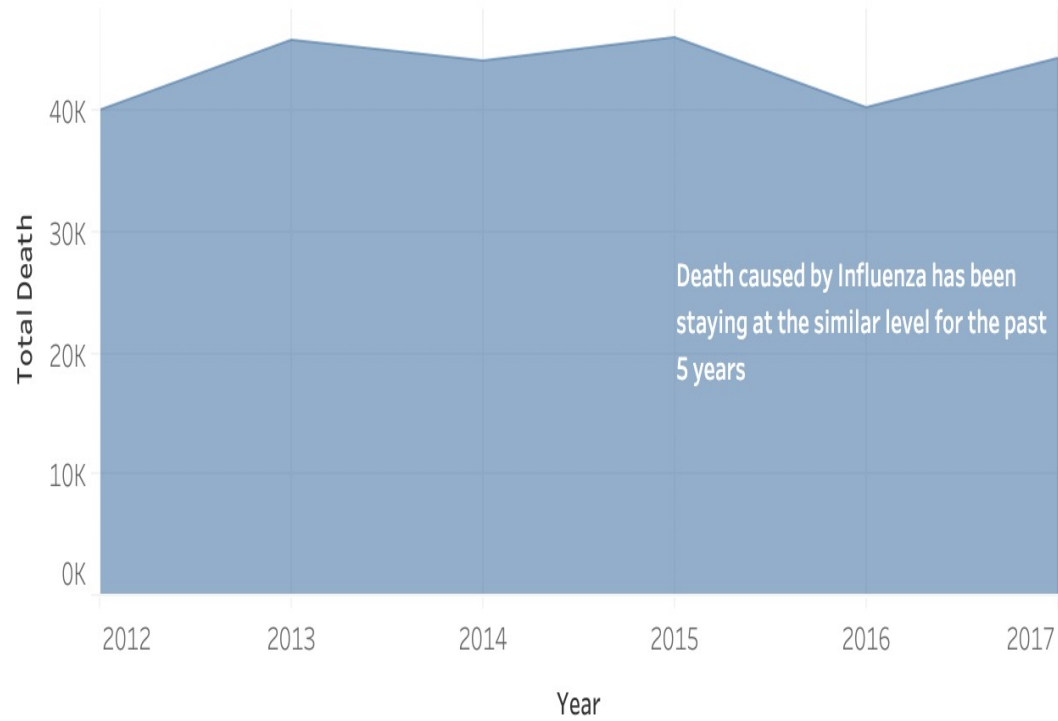


- Data Profiling
- Data Cleaning
- Grouping & Summarizing Data
- Conducting Statistical Analysis
- Hypothesis Testing
- Forecasting
- Visualization & Storyboard Presentation

MSA

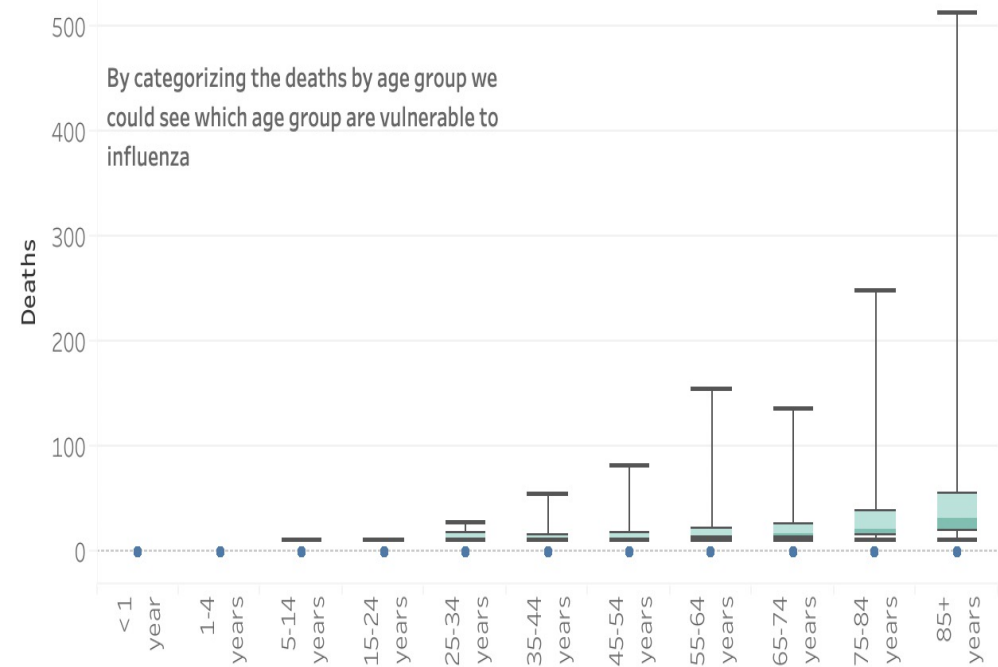
Medical Staffing Analysis

Influenza Death in the United States (2012 - 2017)



First I analyzed the total death in the United States on year to year basis

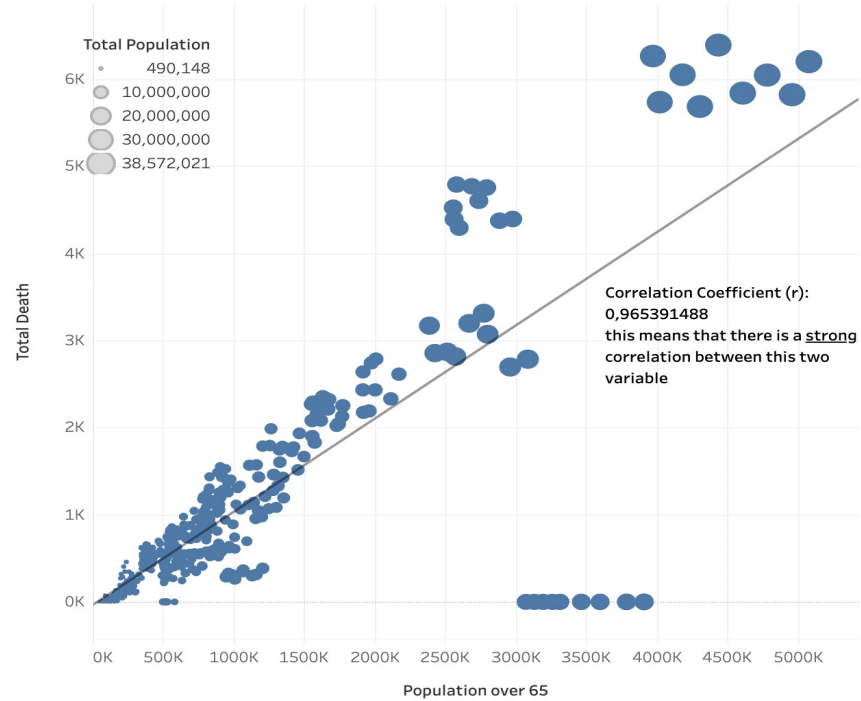
Influenza Deaths by Age (USA, 2009 - 2017)



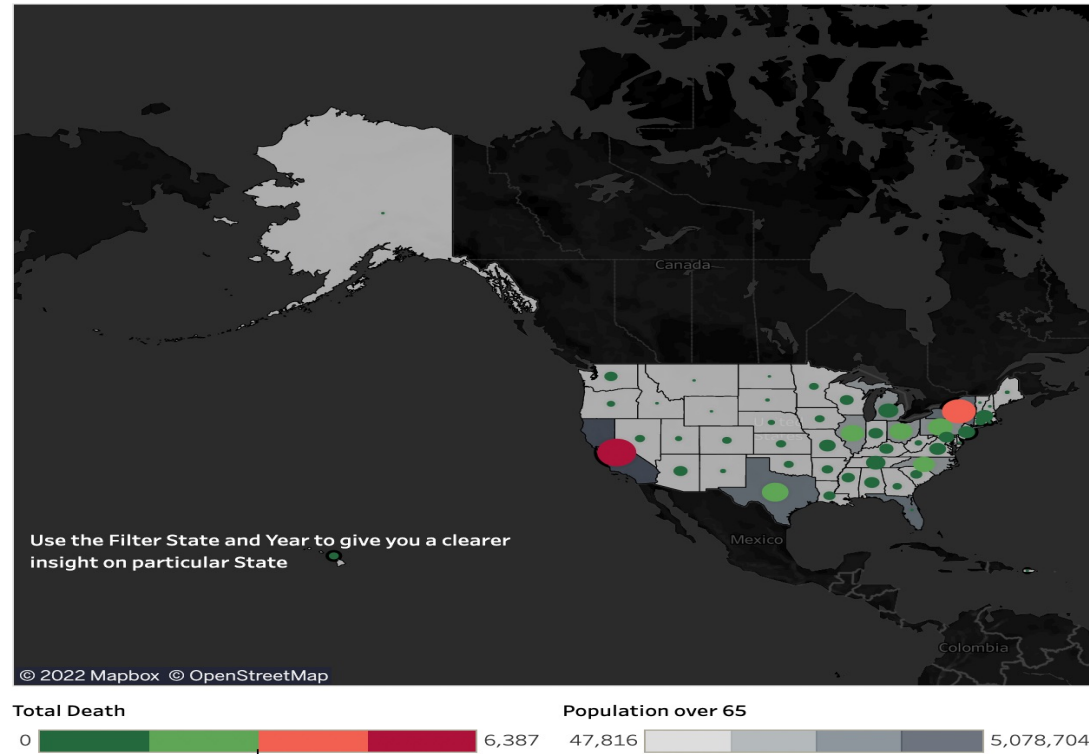
MSA

Medical Staffing Analysis

Correlation between the number of population over 65 and total death in the United States



Influenza Death Rate and Over 65 Population in the United States (2009 - 2017)



To reconfirm my hypothesis, the correlation was tested

Spatial analysis was done with Tableau on each States

MSA

Medical Staffing Analysis – Recommendation

Year
2016 — 2017

Rank of State with the highest number of vulnerable population

1	California	10.037.722
2	Florida	7.694.685
3	Texas	6.045.740
4	New York	5.862.476
5	Pennsylvania	4.285.382
6	Illinois	3.515.609
7	Ohio	3.498.126
8	Michigan	2.953.409
9	North Carolina	2.811.153
10	New Jersey	2.655.693
11	Georgia	2.364.092
12	Virginia	2.151.399
13	Arizona	2.098.986
14	Massachusetts	2.062.686
15	Washington	1.962.560

Rank of Total Influenza Death

1	California	12.015
2	New York	8.764
3	Texas	5.478
4	Pennsylvania	4.941
5	Illinois	4.292
6	Ohio	4.149
7	North Carolina	3.716
8	Michigan	3.181
9	Tennessee	2.953
10	Massachusetts	2.453
11	New Jersey	2.277
12	Virginia	2.174
13	Missouri	2.144
14	Alabama	1.919
15	Indiana	1.751

73%

- Of the total influenza death were counted in the Top 15 States

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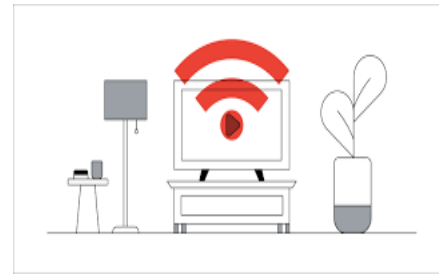
- States with a higher number of vulnerable population have a higher influenza death.

MSA

Medical Staffing Analysis – Recommendation

- Based on our prediction the influenza death rate will remain stable for the next influenza season.
- Vulnerable population has been defined as people who are older than 65 years old – this particular age group has a significantly higher mortality rate
- More Staff and Personal will be needed for the Top 15 States with the highest number of vulnerable population.
- Boost preventive measures (Flu Shots) in the summer months especially for the vulnerable population age group.
- 1:4 Nurse – Patient ratio should be implemented through the influenza season, states with higher ration should be supported from another state with lower ratio.

Rockbuster
Online Video
Rental Service
Analysis



Rockbuster

Online Video Rental Service Analysis

► Background

Rockbuster is a movie rental company that used to have stores around the world. Facing a stiff competition from streaming services, Rockbuster management team is planning to use its existing movie licenses to launch an online video rental services in order to stay competitive.

Note: Rockbuster is a fictional company with generated data given.

► Motivation

Provide Rockbuster Management's Board with data driven answer regarding any ad-hoc business questions from various departments, providing the analyst in order for the management to prepare in transforming the business into an online video rental service.

► Data

The data is a relational database containing information about Rockbuster's film inventory, customers, and payment among other things.

[Rockbusters data set](#) , [Data Dictionary](#)

[PostgreSQL Database](#)

► Tools & Skills



- Relational Database
- SQL & Database query
- Filtering
- Cleaning & Summarizing
- Joins
- CTE



- Visualization



- Creating & Presenting Analytical Findings

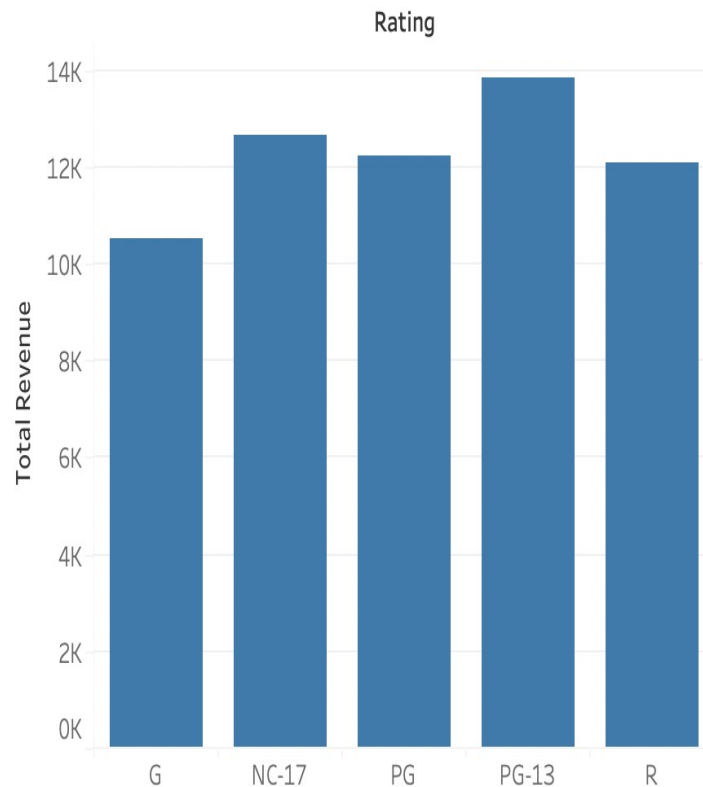
Rockbuster

Online Video Rental Service Analysis

► Using various SQL Queries we managed to determine Rockbuster's several key facts:

- **1.000** film licensed are currently active
- We offer a diversified selection of various genres ,rating and film length in our assortment.
- Our shortest film duration is 46 mins , the longest one is 185 mins - on average our film is 155 mins length.
- All of our films are in **one language** = english
- On average our films are rented for **4,9 days**.
- Rental price varies for films, our **rental rate** starts at **\$0,99 up to \$4,99** (average rental rate: \$2,98)
- **Average Replacement Cost** for our film is **\$ 19,98**
- We have **customers present in 108 countries** around the world.

Revenue Generated by Rating



Rating PG-13 and NC-17 contributed the most revenue for Rockbuster

\$ 61.313 has been generated from our films

Rockbuster Online Video Rental Service Analysis

Revenue generated by Genre



Genre revenue (Genre r..
48  4,892

Sports, Sci- Fi and Animation and drama are the top 4 genre that contribute most revenue

Thriller, Music and Travel contributed the least revenue for Rockbuster

Top 10 Movies

Title	Types of Genre	Revenue
Telegraph Voyage	Music	215,75
Zorro Ark	Comedy	199,72
Wife Turn	Documentary	198,73
Innocent Usual	Foreign	191,74
Hustler Party	Comedy	190,78
Saturday Lambs	Sports	190,74
Titans Jerk	Sci-Fi	186,73
Harry Idaho	Drama	177,73
Torque Bound	Drama	169,76
Dogma Family	Animation	168,72

'Telegraph Voyage' is a film that bring the most revenue for Rockbuster.

2,77 % of the total revenue from Rockbuster are generated from the Top 10 movies

Bottom 10 Movies

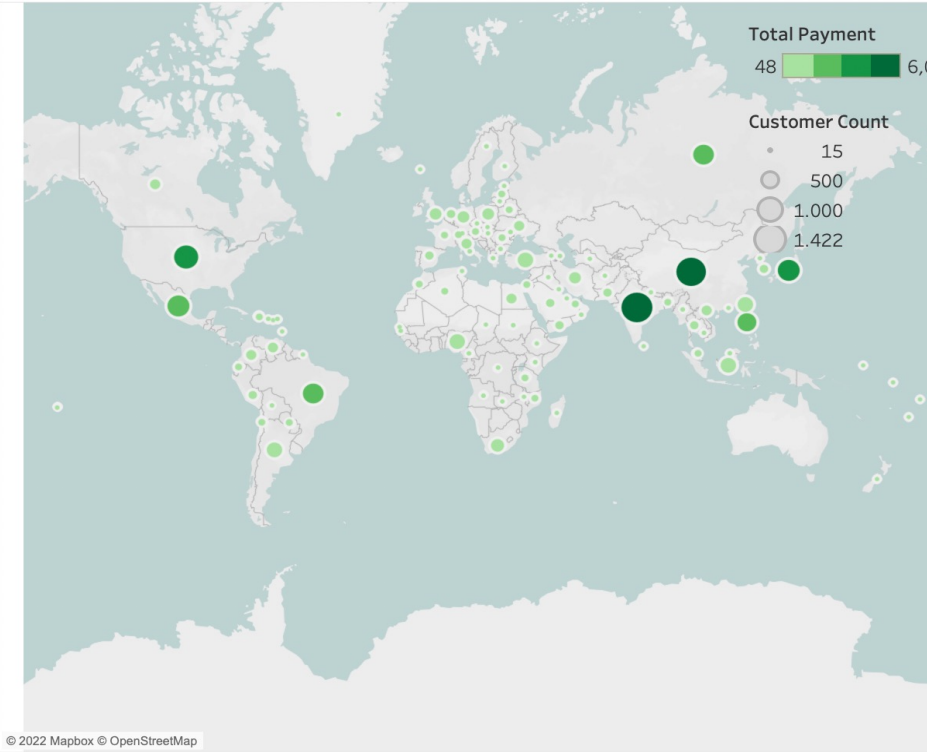
Title	Types of Genre	Revenue
Cruelty Unforgiven	Classics	6,94
Duffel Apocalypse	Documentary	5,94
Freedom Cleopatra	Comedy	5,95
Japanese Run	Horror	7,94
Lights Deer	Classics	7,93
Oklahoma Jumanji	New	5,94
Rebel Airport	Music	6,93
Texas Watch	Horror	5,94
Treatment Jekyll	Drama	6,94
Young Language	Documentary	6,93

\$ 5, 94 is the lowest revenue generated by a film

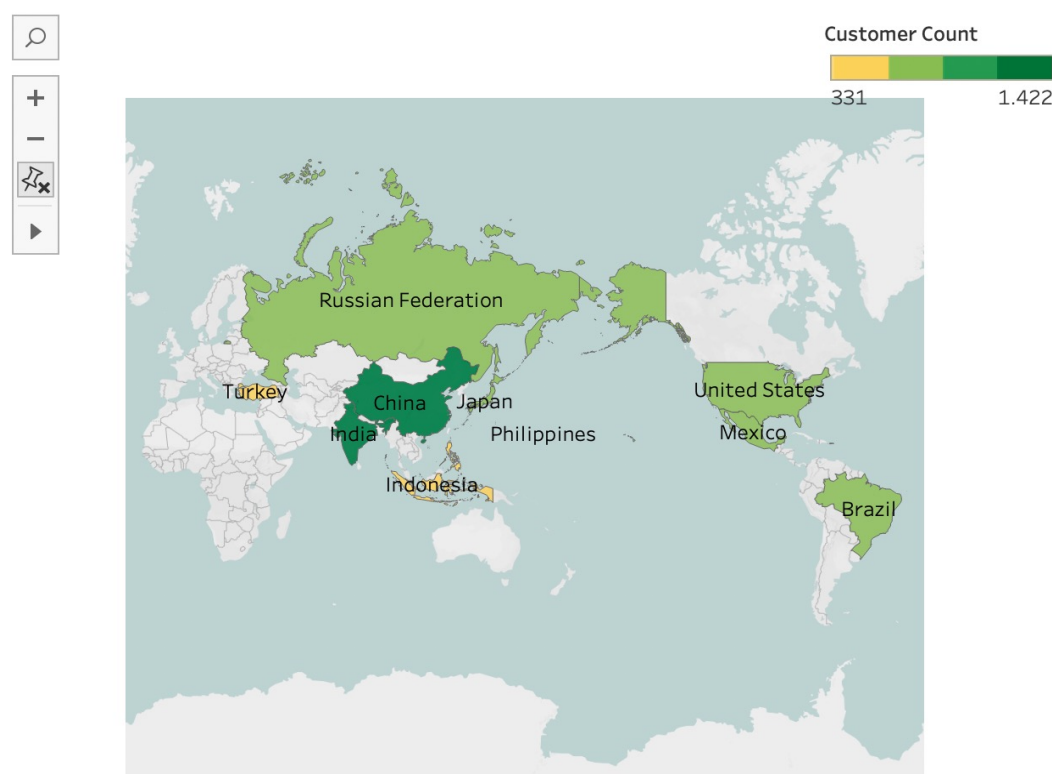
3 Films :Duffel Apocalypse, Oklahoma Jumanji and Texas watch is our least favourite movie (based on the revenue generated)

Rockbuster Online Video Rental Service Analysis

Total Customer and Revenue by Country



Top 10 Countries in terms of revenue & customer count



With customer based in 108 countries around the world. Rockbuster generated **52%** revenue from the Top 10 Countries on right table.

Rockbuster

Online Video Rental Service Analysis - Reccomendation

- Sports, Sci-fi, Animation and Drama are the genre that brings the most revenue.
- Film Rated: PG-13 & NC-17 are our most rented rating
- Rockbuster have over 1.000 active movie licensed and customers based in 108 countries.
- Our customers with high loyalty are scattered across the world. A reward program should be implemented for our loyal customers.
- Asia Pacific Region is the region with the highest concentration of customer based for Rockbuster – this region should be the priorities in our launching phase

Instacart Grocery Delivery Analysis



Instacart

Grocery Delivery Analysis

► Background

Instacart is an online grocery store that operates through an app. Instacart wants to uncover more information about their sales pattern in order to derive insights and suggest strategies for better marketing.

Note: Instacart is a real company that has their data available online. However, the contents of this project have been fabricated for the purpose of this learning exercise

► Key Questions

- The sales team needs to know what the busiest days of the week and hours of the day are in order to schedule ads at times when there are fewer orders.
- They also want to know whether there are certain times of the day when people spend the most money, as this might inform the type of products they advertise at these times.
- Are there certain types of products that are more popular than others?
- The marketing and sales teams are particularly interested in the different types of customers in their system and how their ordering behaviours differ (regarding loyalty status, region, age, family status, and income)

► Data

The data is a combination of multiple open-source data sets from Instacart. A customer data set was created and included for the purpose of this project and does not reflect the actual Instacart company.

[Data Dictionary](#), [Instacart data set](#)

► Tools & Skills



- Data wrangling
- Data merging
- Deriving Variables
- Grouping Data
- Aggregating Data
- Visualizations

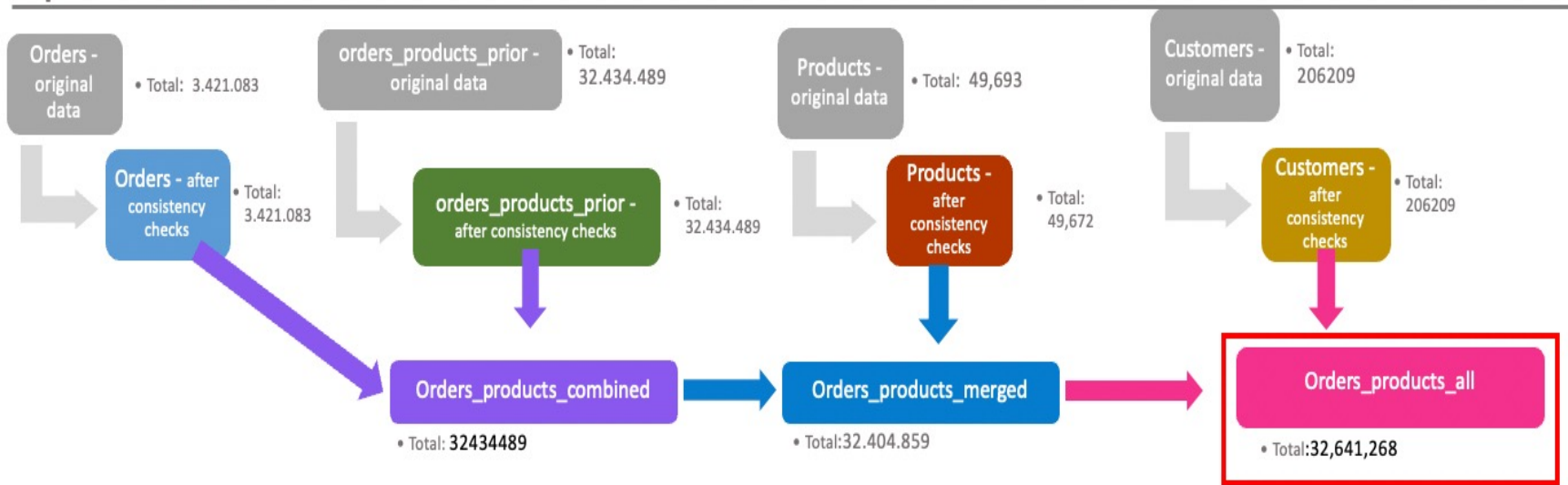


- Consistency Check Reporting

Instacart Grocery Delivery Analysis

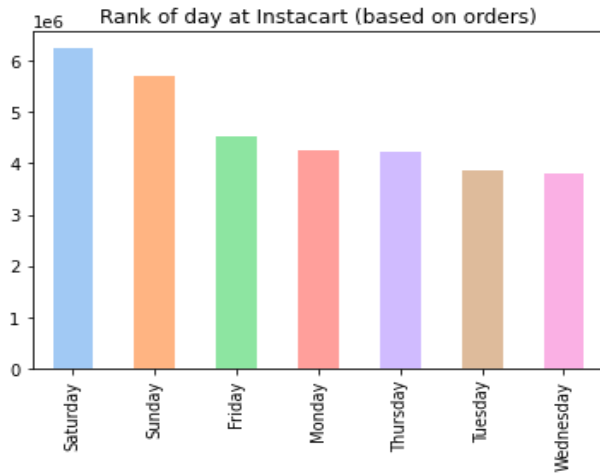


Population flow

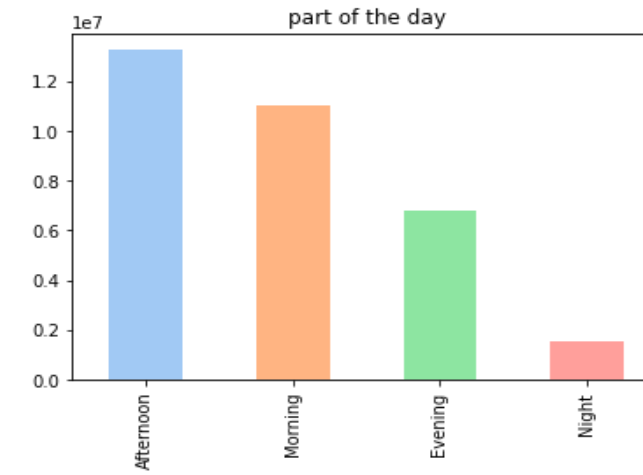


This flow chart represents the original data and progresses to include all added and subtracted data from the Jupyter notebook. The grey boxes in the first row of the population flow represent the original data sets as they were when it was downloaded. The second row of boxes (colored), represents the data sets after it was manipulated, for example, whenever I removed missing values and duplicates. Lastly, the third row represents the merges I performed between the datasets. 23

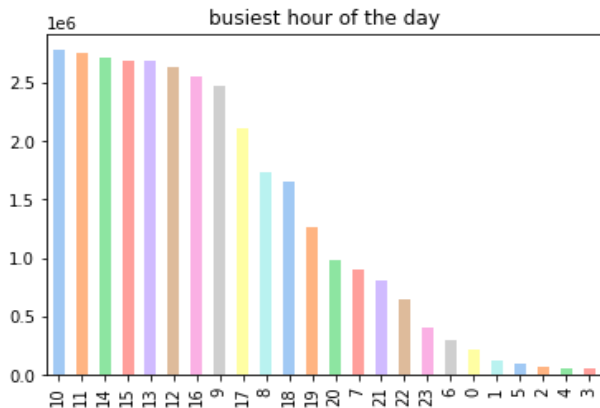
Instacart Analysis - Performance



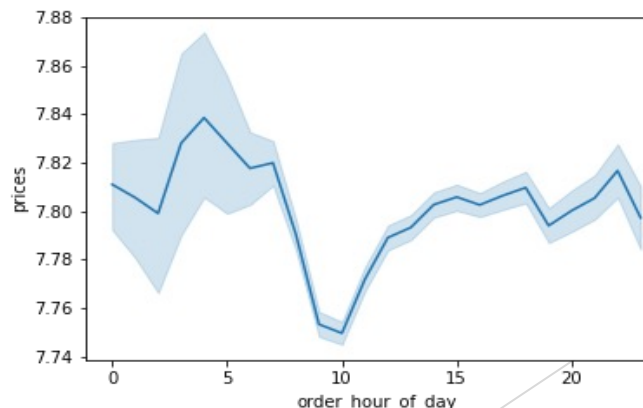
The busiest day at Instacart is on weekend (Saturday, Sunday and Friday).



By grouping the order hour to be able to determine our busiest period of the day, Afternoon and morning are our busiest time of the day.

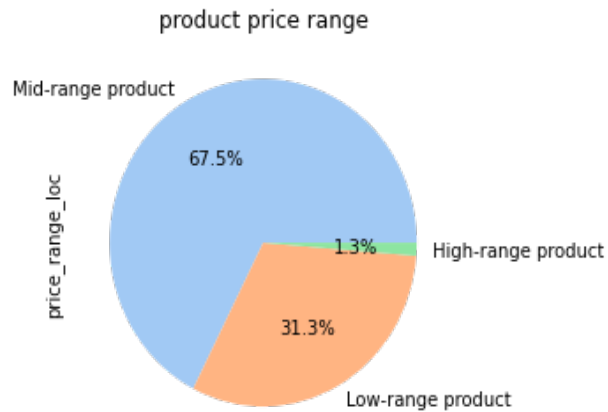


To find better time for advertisement and a smoother staff planning, I analyzed the busiest hour of the day

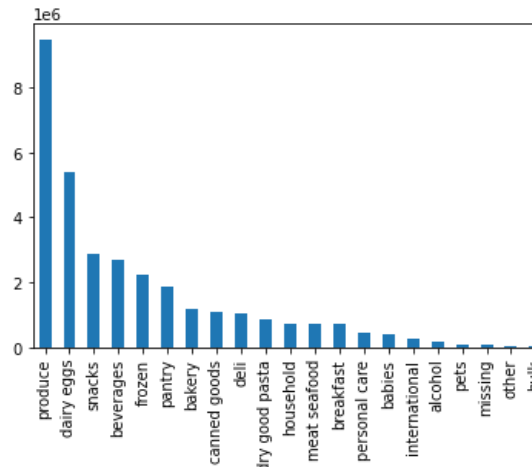


To identify purchasing habits, I looked at the prices of goods at the hour they were ordered. Our chart showed that customers purchase more expensive items at night.

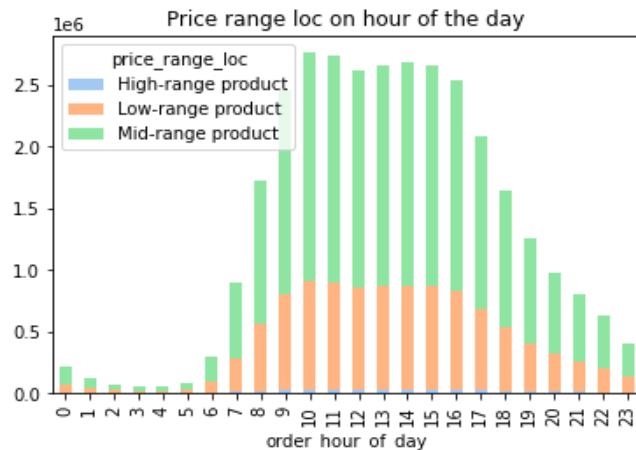
Instacart Grocery Delivery Analysis – Product



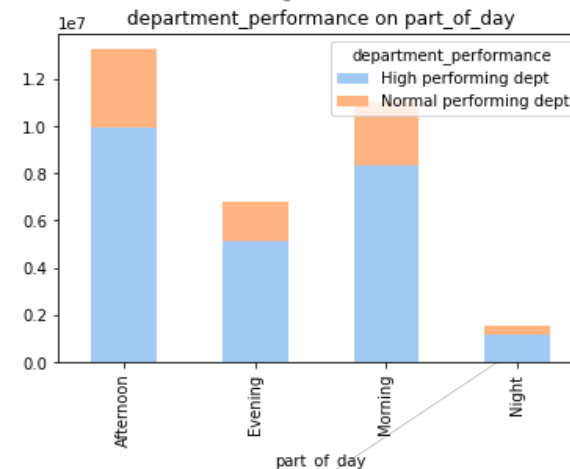
For our pricing department we have found out that **67,5%** of the product offered at Instacart are Mid-range product (\$5 – \$15).



76% of our revenue are generated from the top 6 departments (high performing department).

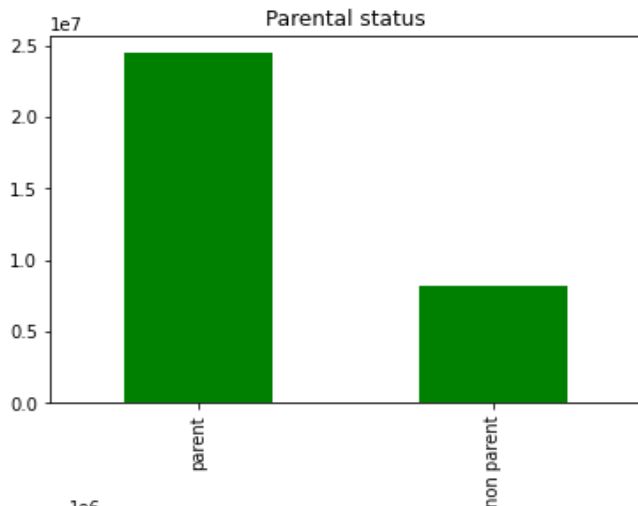


The number of money spent on our product correspondent with our busiest hour

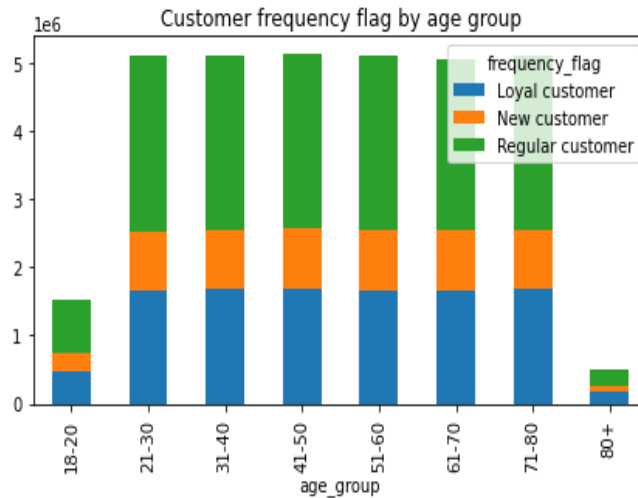


Our Top 6 performing department seem to correlate when there is a high amount of order, there will be more product from our top 6 departments sold

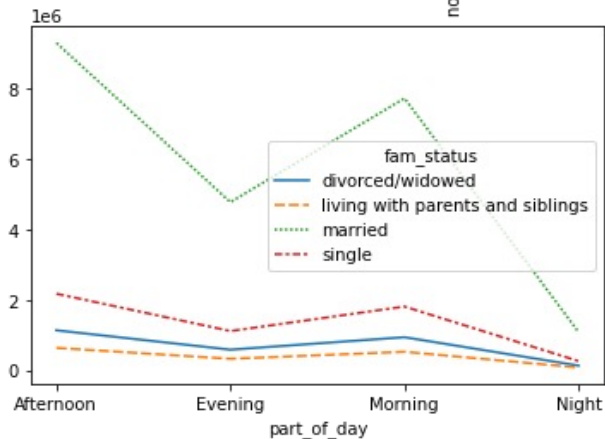
Instacart Analysis – Customer Demographic



75% of Instacart customers are parent



Age-wise we could see that there is a similar balance in terms of customers loyalty flag



There is no significant difference of family status in terms of ordering habits.

frequency_flag	days_since_prior_order		
	mean	min	max
Loyal customer	6.113281	0.0	30.0
New customer	18.250000	0.0	30.0
Regular customer	12.453125	0.0	30.0

For marketing activities we could reward our loyal customers with a voucher with an expiration date under 6 days.

Instacart

Grocery Delivery Analysis - Recommendation

- Our busiest period of the week are weekend, and most orders placed are from 10- 4pm – This should always be considered for the staff planning and products replenishment should happened before this time period.
- 76% of products sold were generated from the Top 6 Departments and 67% of our products consist of mid range products. In order to push sales our other departments we need to advertise on other department group and put a better placement on the low performing department.
- 75% of our customers are parents. Big chunk of our customers are regular customer with an average reordering every 12 days for this type of customer.
- To reward our loyal customer as well as to improve their reordering time we could create a special discount activities with an expiration date of less than 6 days
- Special advertisement should be done in the weakest days (e.g. Wednesday, Thursday) and in the evening/night hour.

Airbnb Case Study: Munich



Airbnb

Case Study: Munich

► Background

Analysed the Airbnb listing data for the city of Munich, which was released in December 2021. Since entering the German market in 2011, Airbnb has been accused of contributing to the overheating rental market in various cities in Germany. Since 2017 Munich has banned short-term rental for residential units (there is some exception). However, I will concentrate more on analysing Airbnb's business model for this analysis.

Key Questions

- What is the average price of an Airbnb's listing in Munich?
- What type of accommodation is available?
- What factors influence the price of the listing?
- Which Munich Neighborhoods are the most popular?

► Data

Open source data from Insideairbnb about Airbnb listing in Munich, scraped in March 2022. Data includes price, availability, review and neighbourhood.

Data Source

► Tools & Skills



- Sourcing Data, data wrangling, consistency check.
- Supervised machine learning: linear regression
- Unsupervised machine learning: clustering
- Spatial analysis
- Time series analysis
- Statistical visualization



- Visualization of analytical journey
- Advanced dashboards

Exploratory Analysis: Munich's Airbnb KPIs

Exploratory Analysis

In the year 2017, The city of Munich banned using the residential unit to reduce the overheating residential market. However, a special license could be obtained and some exemption to this law.

Munich has 4.995 accommodations listed on Airbnb. 68% of the listing are rented from a private host.

Host Type	Count
Private host	3.424
Commercial host	1.571

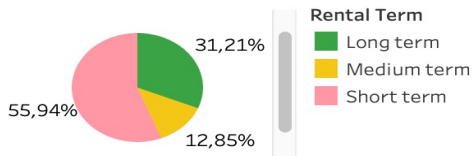
Private host = a host that has one listing on Airbnb
Commercial host = a host that has more than one listing on Airbnb

Average Price Comparison

Room Type	Host Type	
	Commercial host	Private host
Entire home/apt	153,73 €	126,23 €
Hotel room	166,34 €	111,14 €
Private room	106,33 €	86,10 €
Shared room	69,48 €	102,00 €

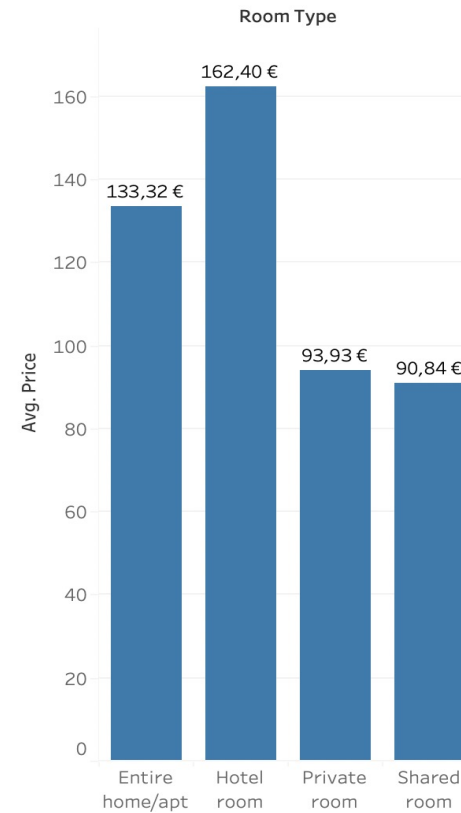
There is 4 different type of accommodations listed on Airbnb. 60% of the accommodations offered are entire home/apt, followed by private room with 37%

Rental Term



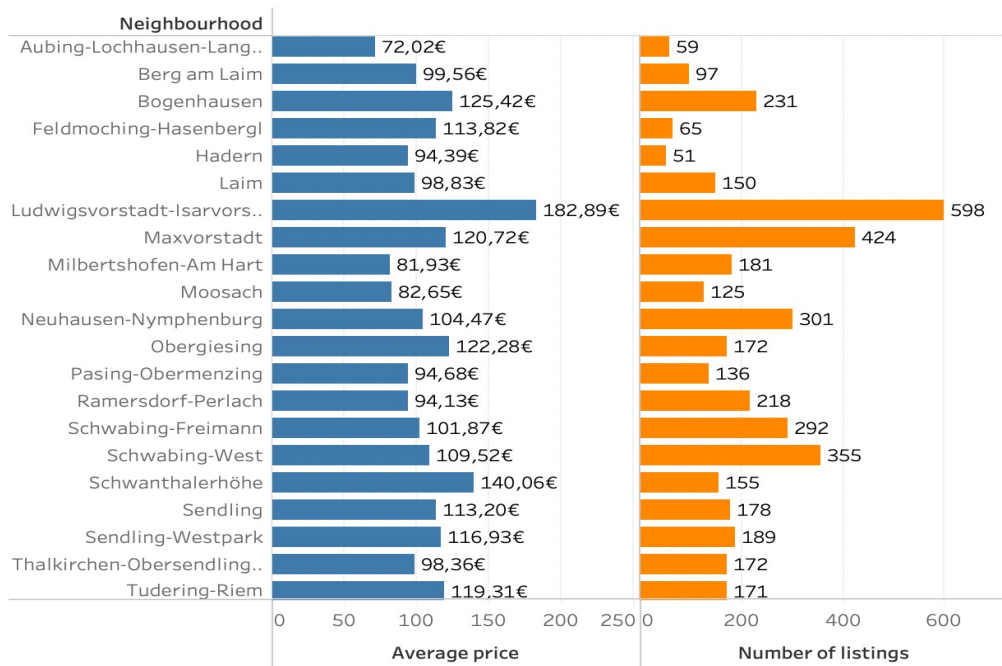
Three different rental terms were categorized based on the minimum nights: short-term is for accommodation available to rent for less than 90 nights, and medium-term is 90 to 180 nights. Long-term is for accommodation that can be rented for over 180 nights.

Average Price for accomodation's type



Exploratory Analysis: Munich 25s neighbourhoods

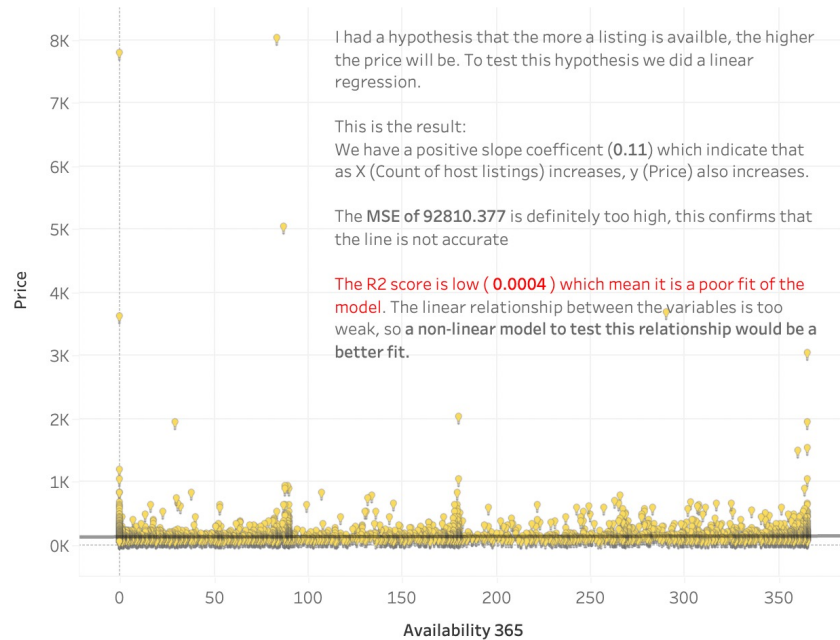
Average prices and number of accommodations do varies in each district in Munich



Price does vary between neighborhoods; we could see that the highest price and highest numbers of listings are mainly in the area in the city center.

Airbnb's Munich: Linear Regression & Clustering

Linear Regression

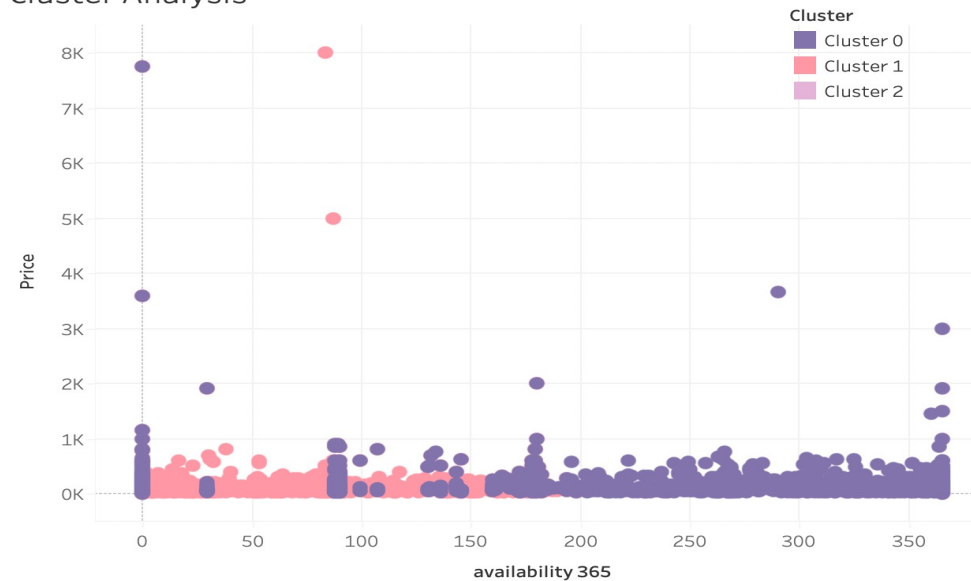


Linear Regression did not prove my hypothesis. A cluster analysis (non-linear approach) was done to further analyze the data.

Cluster analysis searches for similarities between variables in the data set and groups them. With the clustering, we can then analyze the summary statistics of these groups to gain a better insight.

3 distinct groups were returned, which could be seen on the plot below

Cluster Analysis



To test my hypothesis, I tried to find a correlation between variables. However, I failed in using the linear regression; after using a non-linear model (clustering), 3 distinct groups were created.

Recommendation & Key Findings

- 4995 Listings are listed on Airbnb in the city of Munich, this is approximately 1,63% of the total housing unit available in the Munich.
- **Commercial hosts** typically rent their listing **20%** more expensive than a private host.
- **55%** of the listings available are for a short-term period
- The average price for a room in Munich is **118, 48€** and the average **minimum night is 9**.
- 50 % of the property listed are located in the 8 neighborhoods located in the city center of Munich. Ludwigsvorstadt - Isarvorstadt is the neighborhood with the highest concentration of recordings and the highest average price of 182,9€/night (166% more expensive than the average listing price in the least favorite area)...

