```
In [197...
          import asyncio
          from datetime import datetime, timedelta
          from playwright.async api import async playwright
          from bs4 import BeautifulSoup
          import re
          from tqdm.asyncio import tqdm asyncio
          import csv
          import pandas as pd
          import numpy as np
          import holidays
          import seaborn as sns
          import matplotlib.pyplot as plt
          from collections import Counter
          from wordcloud import WordCloud
          from sklearn.model_selection import train_test_split, GridSearchCV, cross_val_score
          from sklearn.preprocessing import StandardScaler, OneHotEncoder
          from sklearn.compose import ColumnTransformer
          from sklearn.pipeline import Pipeline
          from sklearn.ensemble import RandomForestRegressor
          from sklearn.metrics import mean absolute error, mean squared error, r2 score
          import joblib
```

Data Scraping

Get links to scrape

```
In [846...
links = [
    'https://www.trivago.com.my/en-MY/odr/hotels-kuala-lumpur-malaysia?search=200-5
    'https://www.trivago.com.my/en-MY/odr/hotels-malacca-malaysia?search=200-55700'
    'https://www.trivago.com.my/en-MY/odr/hotels-kota-kinabalu-malaysia?search=200-
    'https://www.trivago.com.my/en-MY/odr/hotels-genting-highlands-malaysia?search=200-55
    'https://www.trivago.com.my/en-MY/odr/hotels-johor-bahru-malaysia?search=200-55
    'https://www.trivago.com.my/en-MY/odr/hotels-ipoh-malaysia?search=200-55813'
    'https://www.trivago.com.my/en-MY/odr/hotels-kuching-malaysia?search=200-56107'
    'https://www.trivago.com.my/en-MY/odr/hotels-georgetown-malaysia?search=200-558
    'https://www.trivago.com.my/en-MY/odr/hotels-putrajaya-malaysia?search=200-5620
]
In [961... results = []
```

Start scraping

```
In [ ]: # Base link with placeholders
base_link = "https://www.trivago.com.my/en-MY/srl/{city}?search=200-{search};dr-{ch
# List of city data
cities = [
```

```
{"city": "hotels-kuala-lumpur-malaysia", "search": "55488"},
   {"city": "hotels-malacca-malaysia", "search": "55700"},
   {"city": "hotels-kota-kinabalu-malaysia", "search": "55975"},
   {"city": "hotels-genting-highlands-malaysia", "search": "55757"},
   {"city": "hotels-johor-bahru-malaysia", "search": "55507"},
   {"city": "hotels-ipoh-malaysia", "search": "55891"},
   {"city": "hotels-kuantan-malaysia", "search": "55813"},
   {"city": "hotels-kuching-malaysia", "search": "56107"},
   {"city": "hotels-georgetown-malaysia", "search": "55866"},
   {"city": "hotels-putrajaya-malaysia", "search": "56207"},
]
async def scrape_hotels():
   async with async_playwright() as p:
        # for city data in tqdm asyncio(cities[1:]):
       for x in tqdm_asyncio(range(2,10)):
           browser = await p.chromium.launch(headless=False)
           context = await browser.new_context()
           page = await context.new_page()
           cur_city = cities[x]["city"].replace("hotels-", "").replace("-malaysia"
            search_id = cities[x]["search"]
            start_date = datetime(2025, 1, x+1)
           end date = datetime(2025, 12, 31)
            current_date = start_date
           while current date <= end date:</pre>
                checkin_date = current_date.strftime("%Y%m%d")
                checkout_date = (current_date + timedelta(days=1)).strftime("%Y%m%d
                # Loop through room and quest combinations
                for rooms in range(1, 4):
                    for guests in range(rooms, 4):
                        url = base_link.format(
                            city=cities[x]["city"],
                            search=search id,
                            checkin=checkin date,
                            checkout=checkout_date,
                            rooms=rooms,
                            guests=guests,
                        print(f"Fetching data for {cur city} - {url}")
                        try:
                            await page.goto(url)
                            # Wait for the hotel list to load
                            await page.wait_for_selector('ol[data-testid="accommoda
                            hotels = await page.query selector all('li[data-testid=
                            success_count = 0
                            for hotel in hotels[:5]: # Limit to 5 hotels
                                    # Click on the hotel button
```

```
hotel_btn = await hotel.query_selector('button[
    if hotel btn:
        await hotel btn.click()
        await page.wait_for_timeout(1000)
   # Click on the info button
   info_btn = await hotel.query_selector('button[d
   if info btn:
        await info btn.click()
        await page.wait_for_timeout(1000)
   # Extract the page content
    content = await page.content()
   soup = BeautifulSoup(content, 'html.parser')
   hotel_list = soup.find_all("li", {"data-testid"
   for hotel in hotel_list[:5]:
        # Extract hotel details using BeautifulSoup
        hotel_name = hotel.select_one('button[data-
        price = hotel.select_one('span[data-testid=
        accommodation = hotel.select one('button[da
        dis_label = hotel.select_one('span[class="w
        highlight = hotel.select_one('div[class="iD")
        rating = hotel.select one('meta[itemprop="r
        user rating = hotel.select one('meta[itempr
        user_avg = hotel.select_one('span[itemprop=
        user_comment = hotel.select_one('h2[class*=
        amenities = [amenity.text for amenity in so
        # Print or store the extracted details
        result = {
            "name": hotel_name.text if hotel_name e
            "price_per_night": price.text if price
            "accommodation_type": accommodation.tex
            "distance label": dis label.text if dis
            "highlights_section": highlight.text if
            "hotel_rating": rating['content'] if ra
            "user_rating_count": user_rating['conte
            "user_average_rating": user_avg.text if
            "user_comment": user_comment.text if us
            "amenities": amenities,
            "country": "Malaysia",
            "city": cur_city,
            "checkin": checkin_date,
            "checkout": checkout date,
            "rooms": rooms,
            "guests": guests,
        results.append(result)
        success_count += 1
except Exception as e:
```

Out[101...

```
In [101... len(results)
```

Save to a csy file

10616

Data saved to hotels_data.csv

Data Preprocessing

```
In [120... pd.set_option('display.max_columns', None)
In [130... df = pd.read_csv('hotels_data.csv')
```

Start cleaning data scraped

```
def clean_amenities(text):
In [142...
              if pd.isna(text):
                   return np.nan
              return ''.join(text.split(', '))
          df.loc[956:2054, 'amenities'] = df.loc[956:2054, 'amenities'].apply(clean_amenities
In [134...
          df.loc[:955, 'amenities'] = df.loc[:955, 'amenities'].apply(clean_amenities)
In [132...
In [135...
          df['price_per_night'] = (
              df['price_per_night']
               .str.replace('RM', '', regex=False)
              .str.replace(',', '', regex=False)
               .str.strip()
               .astype(float)
In [135...
          df['checkin'] = pd.to_datetime(df['checkin'].astype(str), format='%Y%m%d')
          df['checkout'] = pd.to_datetime(df['checkout'].astype(str), format='%Y%m%d')
          df['user comment'] = df['user_comment'].dropna().str.extract(r'[0-9.]+([a-zA-Z ]+)
In [135...

    Adding few new columns extracted from existing data

In [135...
          df['day_of_week'] = df['checkin'].dt.day_name()
           df['day_of_month'] = df['checkin'].dt.day
          df['month'] = df['checkin'].dt.month
          custom_column_names = {
In [147...
               'Free WiFi': 'free_wifi',
               'Pool': 'pool',
               'Parking': 'parking',
               'A/C': 'aircond',
               'Restaurant': 'restaurant',
               'Hotel bar': 'hotel_bar',
               'Balcony / Patio': 'balcony_patio',
               'Spa':'spa',
               'Gym':'gym',
               'Hairdryer': 'hairdryer',
               'Hotel bar': 'hotel_bar',
               'Kitchen / Kitchenette': 'kitchen',
               'Non-smoking rooms':'non_smoking_room',
               'TV / Entertainment':'tv_ent',
               'Washing machine': 'washing_machine',
          }
          df['amenities'] = df['amenities'].fillna('').str.split(', ')
In [137...
In [137...
          unique_amenities = custom_column_names.keys()
```

```
for amenity, column_name in custom_column_names.items():
               df[column_name] = df['amenities'].apply(lambda x: 1 if amenity in x else 0)
           df = df.drop(columns=['amenities'])
In [138...
In [140...
           df['highlights_section'] = df['highlights_section'].notna().astype(int)
In [155...
           holidays_dates = holidays.MY()
           df['holiday_name'] = df['checkin'].apply(lambda x: holidays_dates.get(x) if x in ho
           df['holiday'] = df['holiday_name'].notna().astype(int)
           df.head()
In [148...
Out[148...
                  name price_per_night accommodation_type distance_label highlights_section hotel
                  Hotel
                 Indigo
                  Kuala
                Lumpur
                                  647.0
                                                        Hotel
                                                                Kuala Lumpur
                                                                                              1
                 on the
                Park, an
                 IHG ...
                Dorsett
              Hartamas
                                  161.0
                                                        Hotel
                                                                                              1
                                                                Kuala Lumpur
                  Kuala
                Lumpur
                Crowne
                  Plaza
                  Kuala
                                                 Entire House /
           2
                                  683.0
                                                                Kuala Lumpur
                                                                                              0
                Lumpur
                                                    Apartment
                   City
                 Centre
                   Else
           3
                  Kuala
                                  618.0
                                                        Hotel
                                                                Kuala Lumpur
                                                                                              1
                Lumpur
                  Hotel
                 Indigo
                  Kuala
                                  647.0
                                                                                              1
                Lumpur
                                                        Hotel
                                                                Kuala Lumpur
                 on the
                Park, an
                  IHG ...

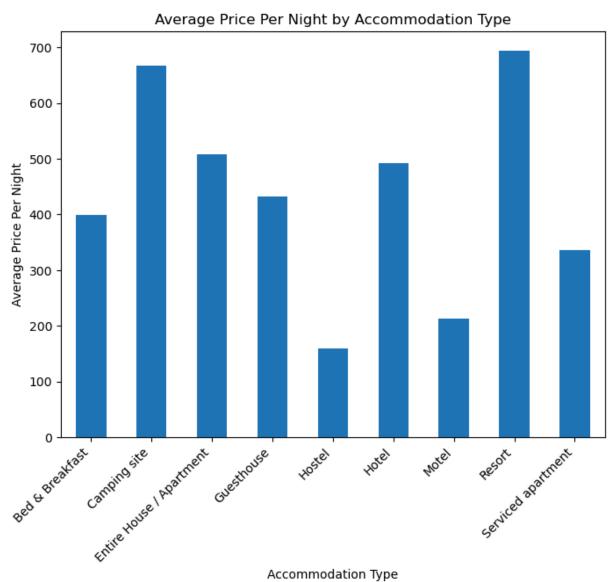
    Save cleaned data to a new csv file
```

```
In [148... # Save the DataFrame to a CSV file
    df.to_csv('cleaned_hotels_data.csv', index=False)
```

Exploratory Data Analysis (EDA)

```
In [157...
          df.accommodation_type.unique()
Out[157...
          array(['Hotel', 'Entire House / Apartment', 'Serviced apartment',
                  'Guesthouse', 'Motel', 'Hostel', 'Resort', 'Bed & Breakfast',
                 'Camping site'], dtype=object)
In [148...
          df.info()
        <class 'pandas.core.frame.DataFrame'>
        RangeIndex: 10616 entries, 0 to 10615
        Data columns (total 34 columns):
             Column
                                  Non-Null Count Dtype
             -----
                                  -----
         0
             name
                                  10616 non-null object
          1
             price_per_night
                                  10616 non-null float64
             accommodation_type
                                  10616 non-null object
                                  10616 non-null object
             distance label
          4
             highlights section
                                  10616 non-null int32
          5
             hotel_rating
                                  10609 non-null float64
                                  10609 non-null float64
             user_rating_count
          7
             user_average_rating 10609 non-null float64
             user_comment
                                  5412 non-null
                                                 object
          9
             country
                                  10616 non-null object
          10 city
                                  10616 non-null object
         11 checkin
                                  10616 non-null datetime64[ns]
          12 checkout
                                  10616 non-null datetime64[ns]
         13 rooms
                                  10616 non-null int64
          14
             guests
                                  10616 non-null int64
          15 day_of_week
                                  10616 non-null object
                                  10616 non-null int32
          16 day_of_month
         17 month
                                  10616 non-null int32
          18 free_wifi
                                  10616 non-null int64
          19
                                  10616 non-null int64
             pool
          20 parking
                                  10616 non-null int64
          21 aircond
                                  10616 non-null int64
          22 restaurant
                                  10616 non-null int64
          23 hotel bar
                                  10616 non-null int64
          24 balcony_patio
                                  10616 non-null int64
          25 spa
                                  10616 non-null int64
                                  10616 non-null int64
          26
             gym
          27 hairdryer
                                  10616 non-null int64
          28 kitchen
                                  10616 non-null int64
          29 non_smoking_room
                                  10616 non-null int64
          30 tv_ent
                                  10616 non-null int64
          31 washing_machine
                                  10616 non-null int64
          32 holiday
                                  10616 non-null int32
          33 holiday_name
                                  469 non-null
                                                  object
        dtypes: datetime64[ns](2), float64(4), int32(4), int64(16), object(8)
        memory usage: 2.6+ MB
In [149...
          numeric_columns = df.select_dtypes(include=['float64', 'int64', 'int32'])
          numeric columns.describe()
```

Out[149	р	rice_per_night	highlights_section	hotel_rating	user_rating_count	user_average_rat
	count	10616.000000	10616.000000	10609.000000	10609.000000	10609.000
	mean	490.300584	0.894970	4.789820	9371.038458	8.082
	std	292.658158	0.306607	1.815419	11822.947641	0.787
	min	28.000000	0.000000	1.000000	2.000000	3.000
	25%	270.000000	1.000000	4.000000	566.000000	7.600
	50%	448.000000	1.000000	4.000000	5561.000000	8.200
	75%	640.000000	1.000000	5.000000	15134.000000	8.700
	max	3275.000000	1.000000	10.000000	116723.000000	10.000
	4					
						•
In [151	<pre>correl_map = numeric_columns.corr() plt.figure(figsize=(16,12)) sns.heatmap(correl_map, annot=True, fmt=".2f") plt.show()</pre>					
	highlights_section hotel_rating user_rating_count user_average_rating rooms guests	- 0.04	31 0.47 0.32 0.01 -0.00 -0.05 -0.05 01 -0.03 -0.06 -0.02 0.11 0.01 0.02 19 -0.03 0.01 0.00 -0.06 0.01 0.02 00 -0.09 -0.08 0.01 0.11 -0.01 -0.00 00 -0.11 -0.02 0.01 -0.08 -0.02 -0.01 11 1.00 0.50 -0.00 -0.00 -0.02 -0.06 02 0.50 1.00 0.00 0.01 -0.02 -0.03 01 -0.00 0.00 1.00 0.00 -0.01 -0.02 -0.03 01 -0.00 0.00 1.00 0.00 -0.01 -0.02 -0.03 01 -0.00 0.00 1.00 0.00 -0.01 -0.02 -0.03 01 -0.00 0.00 1.00 0.00 -0.01 -0.02 -0.03 01 -0.00 0.00 1.00 0.00 -0.01 -0.02 -0.03 01 -0.00 0.00 1.00 0.00 -0.01 -0.02 -0.03 01 -0.00 0.00 1.00 0.00 -0.01 -0.02 -0.03 01 -0.00 0.00 1.00 0.00 -0.01 -0.02 -0.03 01 -0.00 0.00 1.00 0.00 -0.01 -0.02 -0.03 01 -0.00 0.00 1.00 0.00 -0.01 -0.02 -0.03 01 -0.00 0.00 1.00 0.00 -0.01 -0.02 -0.03 01 -0.00 0.00 1.00 0.00 -0.01 -0.02 -0.03 01 -0.00 0.00 1.00 0.00 -0.01 -0.02 -0.03 01 -0.00 0.00 1.00 0.00 -0.01 -0.02 -0.03 01 -0.00 0.00 0.00 -0.01 -0.02 -0.03 01 -0.00 0.00 0.00 -0.01 -0.02 -0.03 01 -0.00 0.00 -0.00 -0.01 -0.02 -0.03 01 -0.00 0.00 0.00 -0.00 -0.01 -0.02 -0.03 01 -0.00 0.00 0.00 -0.00 -0.01 -0.02 -0.03 01 -0.00 0.00 0.00 -0.00 -0.01 -0.02 -0.03 01 -0.00 0.00 0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 0 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.00 -0.	0.01 0.01 0.01 0.03 0.01 0.00 0.00 0.01 0.03 0.10 0.01 0.01 0.00 0.04 0.02 0.03 0.03 0.02 0.01 0.12 0.03 0.03 0.05 0.17 0.02 0.03 0.02 0.02 0.09 0.03	0.02 0.04 0.02 0.00 0.04 0.03 0 0.01 0.03 0.00 0.07 0.00 0.01 0 0.02 0.01 0.01 0.04 0.02 0.04 0 0.00 0.05 0.10 0.06 0.02 0.08 0 0.07 0.00 0.04 0.04 0.02 0.00 0 0.05 0.00 0.00 0.04 0.02 0.01 0.04 0	0.02 0.03 0.04 0.01 0.02 -0.01 - 0.8 0.03 0.01 0.01 -0.00 - 0.6
	free_wifi pool parking aircond restaurant hotel_bar	-0.05 0.01 0.01 -0.01 -0 -0.05 0.02 0.02 -0.00 -0 -0.06 0.01 0.00 -0.01 -0 -0.06 0.01 0.00 -0.01 -0 -0.04 0.01 0.01 0.00 -0 -0.07 0.03 -0.03 0.04 0	08 -0.00 0.01 0.00 1.00 0.01 -0.02 0.20 0.20 -0.02 -0.01 -0.01 1.00 0.83 0.1 -0.06 -0.03 -0.02 -0.01 0.22 0.83 0.3 -0.03 -0.02 -0.02 0.31 0.20 0.83 0.3 -0.03 -0.02 -0.03 -0.02 0.33 0.86 0.2 -0.05 -0.02 -0.02 -0.00 0.83 0.94 0.1 -0.17 -0.09 0.01 0.06 0.48 0.58	0.92 0.93 0.83 0.48 0.08 0.88 0.86 0.94 0.52 0.09 0.87 0.52 0.08 0.87 0.08 0.87 0.08 0.87 0.08 0.87 0.08 0.87 0.88 0.87 0.88 0.87 0.88 0.87 0.88 0.89 0.89 0.89 0.89 0.89 0.89 0.89	0.38 0.28 0.08 0.06 0.10 0.10 0 0.44 0.31 0.07 0.06 0.10 0.09 0 0.41 0.29 0.08 0.06 0.10 0.09 0 0.40 0.28 0.07 0.06 0.10 0.09 0 0.42 0.32 0.07 0.06 0.09 0.09 0 0.26 0.17 0.01 0.05 0.02 0.03 0	0.03 0.01 0.03 0.02 0.03 0.01 0.03 0.01 0.03 0.02 - 0.2
	spa gym hairdryer kitchen	-0.03 0.02 0.01 0.02 0 -0.07 0.04 0.03 -0.01 -0 -0.04 0.02 0.00 -0.01 -0 -0.02 0.00 0.07 -0.04 -0	12 -0.02 0.03 -0.00 -0.07 0.08 0.09 0.0 -0.07 -0.05 0.01 0.07 0.38 0.44 0.5 0.00 0.00 0.01 0.28 0.31 0.00 0.04 0.05 -0.00 0.14 0.28 0.07 0.6 0.04 0.05 -0.00 0.24 0.06 0.06 0.02 -0.02 0.01 0.00 0.13 0.10 0.10	0.41 0.40 0.42 0.26 0.07 0.29 0.28 0.32 0.17 0.20 0.08 0.07 0.07 0.01 0.40 0.06 0.06 0.06 0.05 0.45	1.00 0.25 0.09 0.02 0.07 0.10 0 0.25 1.00 0.00 -0.02 0.12 0.07 -0 0.09 0.00 1.00 0.34 0.48 0.82 0 0.02 -0.02 0.34 1.00 0.30 0.46 0	0.02 -0.00 0.05 0.03 - 0.0 0.09 -0.01
	washing_machine	- 0.06 -0.02 0.04 -0.02 0 - 0.01 0.03 0.01 -0.01 0	08 -0.00 0.04 -0.01 -0.16 0.10 0.09 03 -0.01 -0.03 -0.00 -0.14 0.03 0.03 01 -0.00 -0.00 0.02 -0.05 0.01 0.02 -1	0.03 0.03 0.03 0.04 0.21	0.02 -0.05 0.09 0.53 0.08 0.05 1	00 0.05



The top 3 accommodation type with the highest average price per night is Resort, Camping Site, and Entire House / Apartment

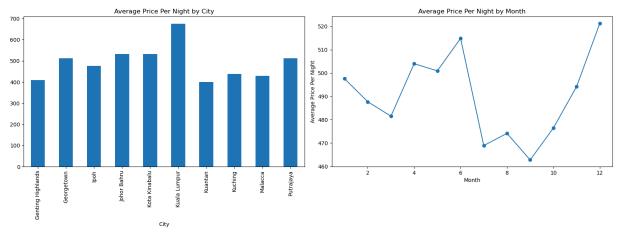
```
fig, axes = plt.subplots(1, 2, figsize=(16, 6))

# Average price per night by month
average_price_by_month = df.groupby('month')['price_per_night'].mean()
average_price_by_month.plot(kind='line', marker='o', ax=axes[1], title='Average Pri
```

```
axes[1].set_xlabel('Month')
axes[1].set_ylabel('Average Price Per Night')

# Average price per night by city
average_price_by_city = df.groupby('city')['price_per_night'].mean()
average_price_by_city.plot(kind='bar', ax=axes[0], title='Average Price Per Night b'
axes[0].set_xlabel('City')

plt.tight_layout()
plt.show()
```



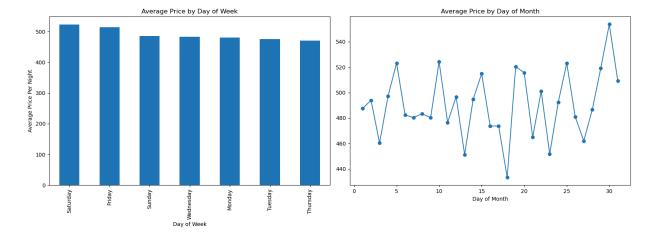
The top 3 city with the highest average price per night is Kuala Lumpur, Johor Bahru and Kota Kinabalu. While the month with the highest average price per night is December.

```
In [159... fig, axes = plt.subplots(1, 2, figsize=(16, 6))

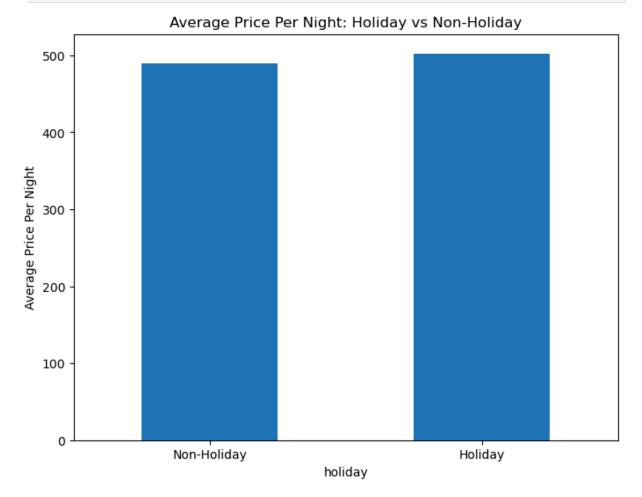
# Average price by day of week
average_price_by_day_of_week = df.groupby('day_of_week')['price_per_night'].mean().
average_price_by_day_of_week.plot(kind='bar', ax=axes[0], title='Average Price by D
axes[0].set_xlabel('Day of Week')
axes[0].set_ylabel('Average Price Per Night')

# Average price by day of month
average_price_by_day_of_month = df.groupby('day_of_month')['price_per_night'].mean(
average_price_by_day_of_month.plot(kind='line', marker='o', ax=axes[1], title='Averaxes[1].set_xlabel('Day of Month')

plt.tight_layout()
plt.show()
```



Friday and weekends have higher average price per night compared to other days of the week. While the day of month with the highest average price per night is 30th.



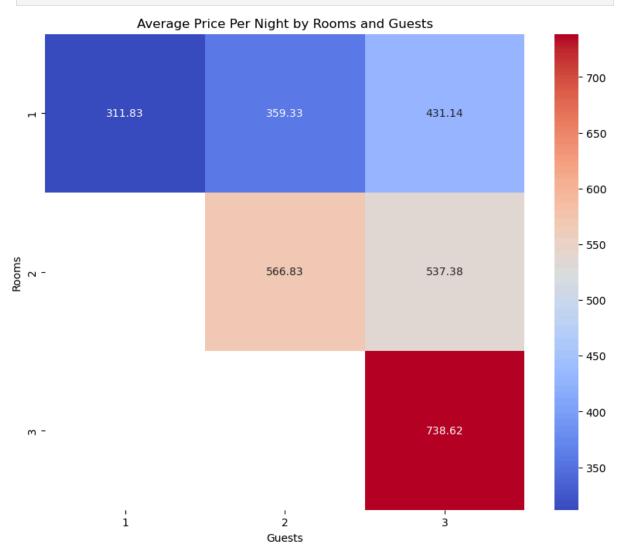
Holidays have slightly higher average price per night compared to non-holidays.

```
# Group by room and guests
average_price_by_combination = df.groupby(['rooms', 'guests'])['price_per_night'].m

# Pivot for heatmap-like visualization
pivot_table = average_price_by_combination.pivot(index='rooms', columns='guests', v

# Plot as a heatmap
import seaborn as sns

plt.figure(figsize=(10, 8))
sns.heatmap(pivot_table, annot=True, fmt=".2f", cmap="coolwarm")
plt.title('Average Price Per Night by Rooms and Guests')
plt.xlabel('Guests')
plt.ylabel('Rooms')
plt.show()
```



Number of guests and number of rooms are directly proportional to the average price per night.

```
In [170... # Combine all text from 'user_comment' column
all_comments = " ".join(df['user_comment'].dropna().astype(str).str.replace(r'\s+',
```

```
# Word frequency
word_freq = Counter(all_comments.split())

# Word cloud visualization
wordcloud = WordCloud(width=800, height=400, background_color='white').generate(all
plt.figure(figsize=(10, 5))
plt.imshow(wordcloud, interpolation='bilinear')
plt.axis('off')
plt.title('Most Common Words in User Comments')
plt.show()
```

Most Common Words in User Comments



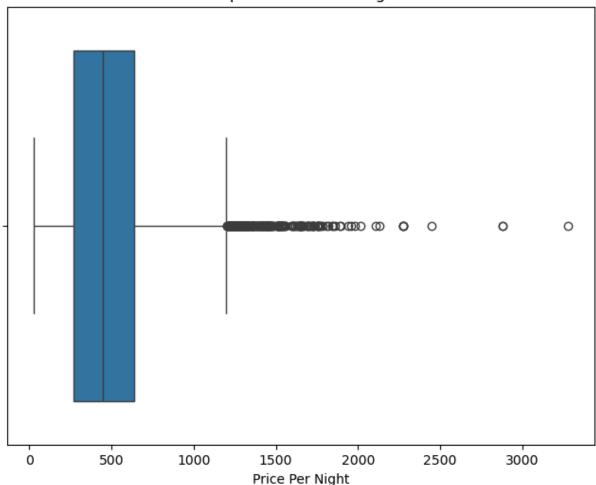
Excellent Excellent

Excellent Verygood

The most common user comment is 'Excellent' and 'Very good'

```
In [165... # Boxplot for price_per_night
   plt.figure(figsize=(8, 6))
   sns.boxplot(x=df['price_per_night'])
   plt.title('Boxplot of Price Per Night')
   plt.xlabel('Price Per Night')
   plt.show()
```

Boxplot of Price Per Night



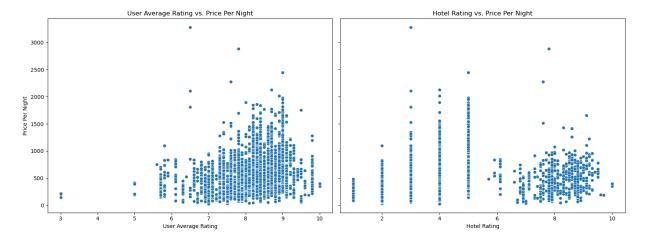
Most hotel prices per night are concentrated below 500, with a few extreme outliers exceeding 1000, indicating a positively skewed distribution.

```
In [166... fig, axes = plt.subplots(1, 2, figsize=(16, 6), sharey=True)

sns.scatterplot(data=df, x='user_average_rating', y='price_per_night', ax=axes[0])
axes[0].set_title('User Average Rating vs. Price Per Night')
axes[0].set_xlabel('User Average Rating')
axes[0].set_ylabel('Price Per Night')

sns.scatterplot(data=df, x='hotel_rating', y='price_per_night', ax=axes[1])
axes[1].set_title('Hotel Rating vs. Price Per Night')
axes[1].set_xlabel('Hotel Rating')
axes[1].set_ylabel('Price Per Night')

plt.tight_layout()
plt.show()
```



The scatterplots show a weak positive correlation between price per night and both user average ratings and hotel ratings, with higher-rated hotels generally having more expensive prices, but significant variability and outliers are present in both cases.

Model Making

```
In [189... df = pd.read_csv('cleaned_hotels_data.csv')
```

• Dropping some columns due to irrelevancy

```
In [189... df = df.drop(columns=['holiday_name', 'month', 'day_of_month', 'day_of_week', 'user
```

• Handling missing values

```
In [190... df['user_comment'] = df['user_comment'].fillna(0)
In [190... df.info()
```

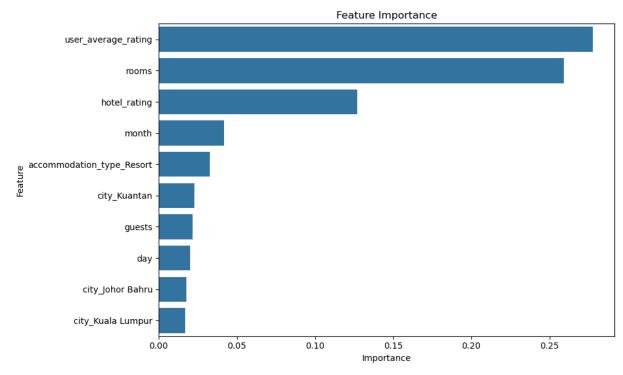
```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 10616 entries, 0 to 10615
Data columns (total 24 columns):
# Column
                       Non-Null Count Dtype
--- -----
                       _____
0
    price_per_night
                       10616 non-null float64
1
    accommodation_type 10616 non-null object
 2
    hotel_rating
                       10609 non-null float64
 3
    user average rating 10609 non-null float64
4
                       10616 non-null object
    user_comment
 5
    city
                       10616 non-null object
 6
    checkin
                      10616 non-null object
 7
    rooms
                       10616 non-null int64
   guests
                      10616 non-null int64
 9
    free_wifi
                       10616 non-null int64
10 pool
                      10616 non-null int64
11 parking
                      10616 non-null int64
                      10616 non-null int64
 12 aircond
13 restaurant
                      10616 non-null int64
 14 hotel bar
                       10616 non-null int64
15 balcony_patio
                       10616 non-null int64
                       10616 non-null int64
16 spa
17 gym
                       10616 non-null int64
18 hairdryer
                      10616 non-null int64
19 kitchen
                       10616 non-null int64
 20 non_smoking_room 10616 non-null int64
21 tv_ent
                       10616 non-null int64
22 washing_machine
                       10616 non-null int64
23 holiday
                       10616 non-null int64
dtypes: float64(3), int64(17), object(4)
memory usage: 1.9+ MB
```

```
In [192...
          # Convert datetime to numeric features
          if 'checkin' in df.columns:
              df['checkin'] = pd.to_datetime(df['checkin'])
              df['month'] = df['checkin'].dt.month
              df['day'] = df['checkin'].dt.day
              df['day_of_week'] = df['checkin'].dt.dayofweek
              df.drop(columns=['checkin'], inplace=True)
          # Encode categorical features
          categorical_columns = df.select_dtypes(include=['object']).columns
          numerical_columns = df.select_dtypes(include=['float64', 'int64', 'int32']).columns
          # Define the target column
          target_column = 'price_per_night'
          X = df.drop(columns=[target_column])
          y = df[target_column]
          # One-hot encoding for categorical variables
          preprocessor = ColumnTransformer(
              transformers=[
                  ('num', StandardScaler(), numerical_columns),
                  ('cat', OneHotEncoder(handle_unknown='ignore'), categorical_columns)
              ])
```

```
# Split data into training and testing sets
In [192...
          X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_sta
          # Convert to string
In [192...
          for col in categorical columns:
              X_train[col] = X_train[col].astype(str)
              X_test[col] = X_test[col].astype(str)
In [193...
          # Create a pipeline with preprocessing and a model
          pipeline = Pipeline(steps=[
              ('preprocessor', preprocessor),
              ('model', RandomForestRegressor(random_state=42))
          1)
          # Train the model
          pipeline.fit(X_train, y_train)
Out[193...
                                  Pipeline
                       preprocessor: ColumnTransformer
                          num
                                                    cat
                  StandardScaler
                                             OneHotEncoder
                          RandomForestRegressor
          # Hyperparameter tuning for RandomForest
In [193...
          param_grid = {
              'model__n_estimators': [50, 100, 200],
              'model__max_depth': [None, 10, 20, 30],
               'model__min_samples_split': [2, 5, 10]
          }
          grid_search = GridSearchCV(pipeline, param_grid, cv=3, scoring='r2', n_jobs=-1)
          grid_search.fit(X_train, y_train)
          # Best parameters and score
          print("Best Parameters:", grid_search.best_params_)
          print("Best Score:", grid_search.best_score_)
         Best Parameters: {'model__max_depth': 30, 'model__min_samples_split': 5, 'model__n_e
         stimators': 200}
         Best Score: 0.8346677723140319
          # Extract feature importance
In [194...
          final_model = grid_search.best_estimator_.named_steps['model']
          feature_names = numerical_columns.tolist() + list(grid_search.best_estimator_.named
          importance = final_model.feature_importances_
          # Plot feature importance
```

```
importance_df = pd.DataFrame({'Feature': feature_names, 'Importance': importance})
importance_df = importance_df.sort_values(by='Importance', ascending=False)

plt.figure(figsize=(10, 6))
sns.barplot(x='Importance', y='Feature', data=importance_df.head(10))
plt.title('Feature Importance')
plt.tight_layout()
plt.show()
```



```
In [193...
          # Predictions and evaluation
          y_pred = grid_search.predict(X_test)
           print("Mean Absolute Error:", mean_absolute_error(y_test, y_pred))
           print("Mean Squared Error:", mean_squared_error(y_test, y_pred))
          print("Root Mean Squared Error:", np.sqrt(mean_squared_error(y_test, y_pred)))
           print("R2 Score:", r2_score(y_test, y_pred))
         Mean Absolute Error: 51.06890466867687
         Mean Squared Error: 14927.473348319634
         Root Mean Squared Error: 122.17803955015661
         R<sup>2</sup> Score: 0.8296858573216669
          # Save the model
In [193...
          joblib.dump(grid_search.best_estimator_, 'hotel_price_prediction_model.pkl')
Out[193...
           ['hotel_price_prediction_model.pkl']
```