

Rental Prices - Exploratory Data Analysis

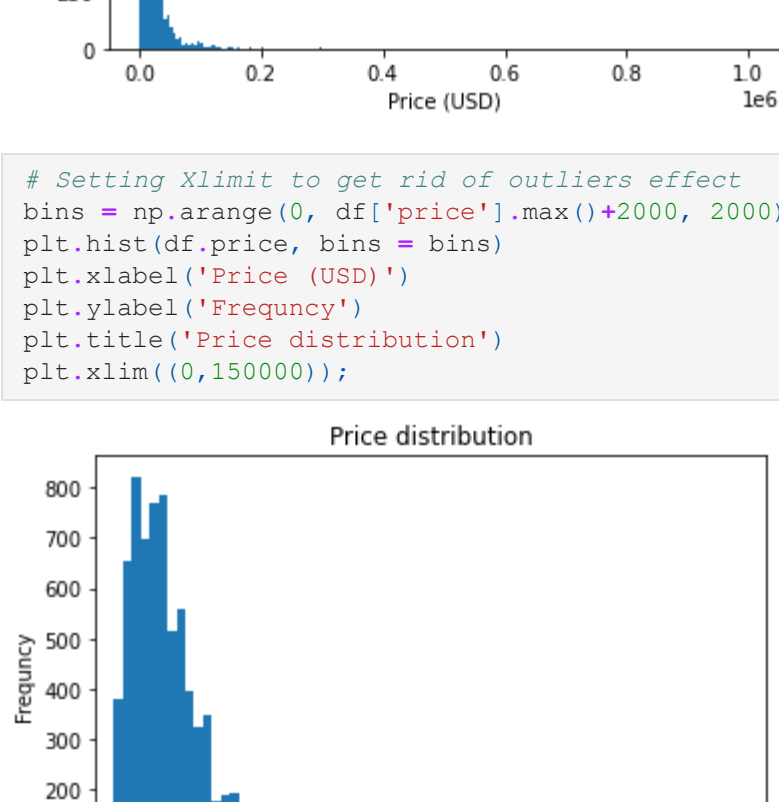
```
In [1]: # Importing packages
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
%matplotlib inline

# Loading data and printing out a few lines.
df = pd.read_csv('rentals_cleaned.csv', index_col=0)
```

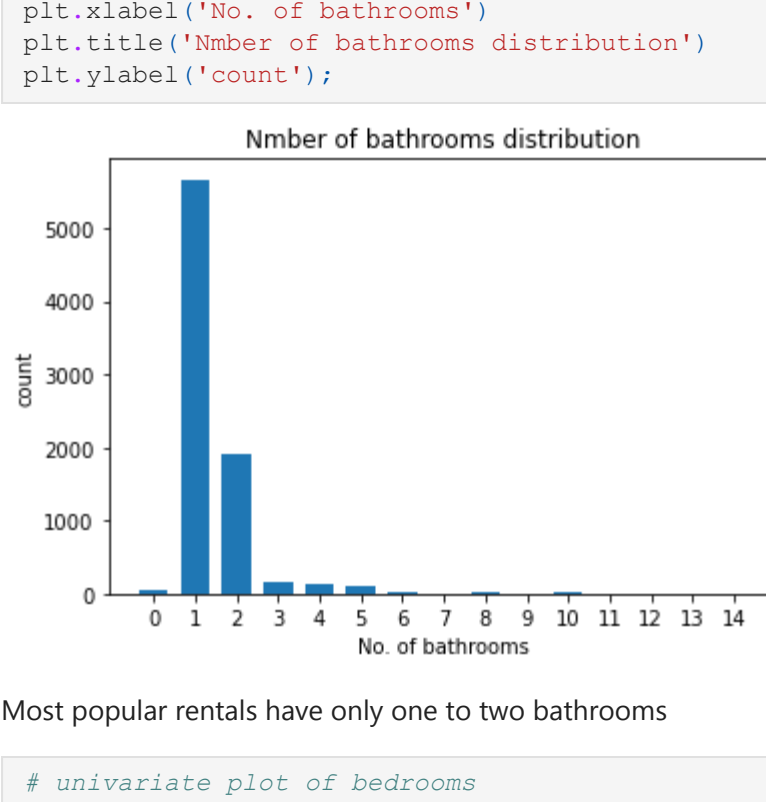
Univariate Exploration

Let's start our exploration by looking at the main variable of interest: price to see its distribution type: skewed or symmetric?

```
In [2]: # univariate plot of price
bins = np.arange(-5, df['price'].max()+5000, 5000)
plt.hist(df.price, bins=bins)
plt.xlabel('Price (USD)')
plt.ylabel('Frequency')
plt.title('Price distribution');
```



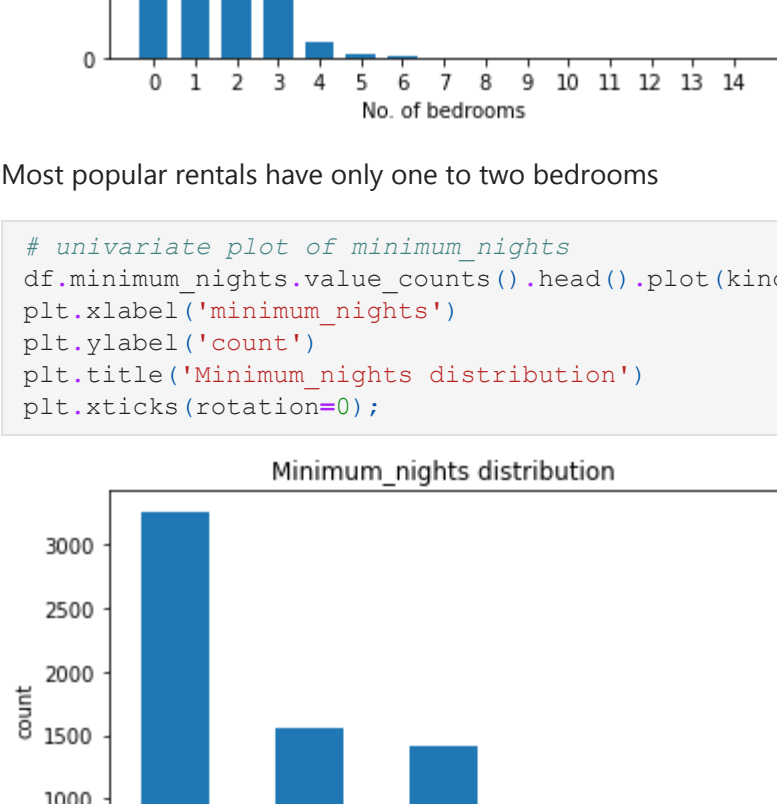
```
In [3]: # Setting Xlimit to get rid of outliers effect
bins = np.arange(0, df['price'].max()+2000, 2000)
plt.hist(df.price, bins=bins)
plt.xlabel('Price (USD)')
plt.ylabel('Frequency')
plt.title('Price distribution')
plt.xlim((0,150000));
```



Price distribution is skewed, most popular prices are under 20000\$

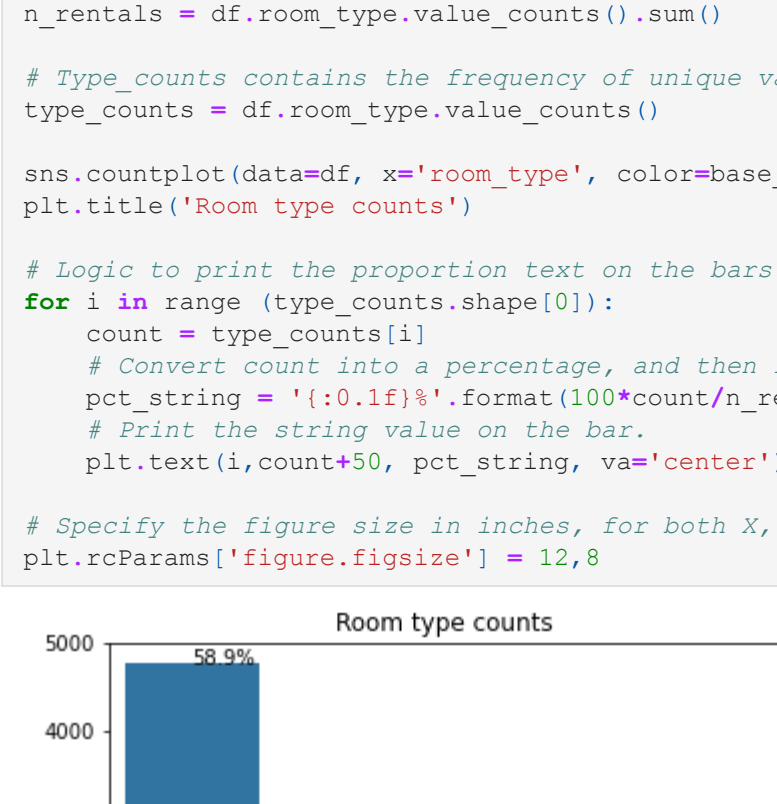
Next, create a plot of numerical factors affecting price: bathrooms, bedrooms, and find if there is an interesting pattern in these values?

```
In [4]: # univariate plot of bathrooms
bins = np.arange(-5, df['bathrooms'].max()+1, 1)
plt.hist(data=df, x='bathrooms', bins=bins, rwidth=0.7)
plt.xticks(np.arange(0, 14+1, 1))
plt.xlabel('No. of bathrooms')
plt.title('Nber of bathrooms distribution')
plt.ylabel('count');
```



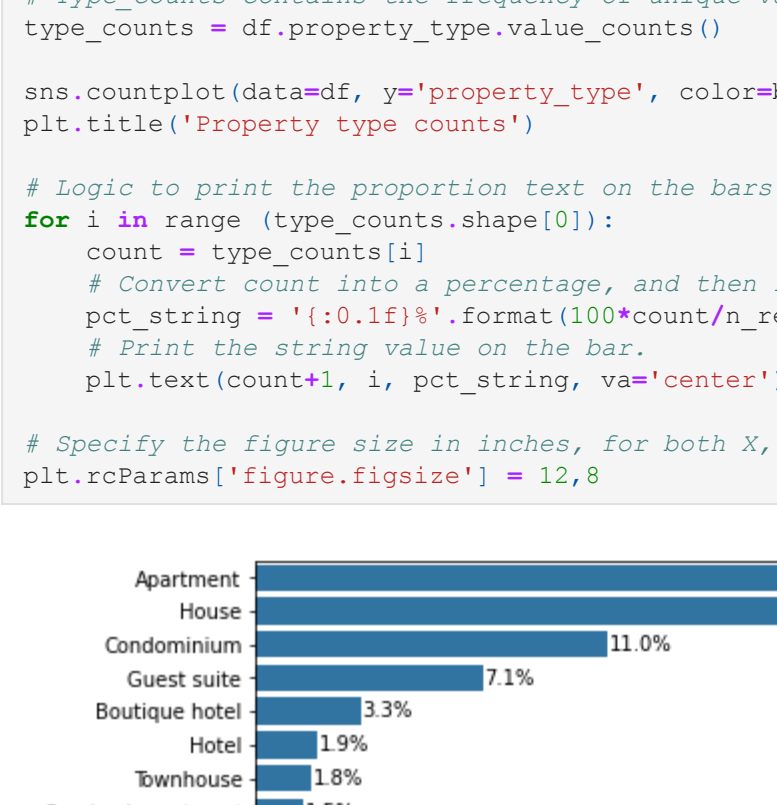
Most popular rentals have only one to two bathrooms

```
In [5]: # univariate plot of bedrooms
bins = np.arange(-5, df['bedrooms'].max()+1, 1)
plt.hist(data=df, x='bedrooms', bins=bins, rwidth=0.7)
plt.xticks(np.arange(0, 14+1, 1));
plt.xlabel('No. of bedrooms')
plt.title('Nber of bedrooms distribution')
plt.ylabel('count');
```



Most popular rentals have only one to two bedrooms

```
In [6]: # univariate plot of minimum_nights
sns.scatterplot(data=df, df['minimum_nights'].head().plot(kind='bar')
plt.xlabel('Minimum_nights')
plt.ylabel('count')
plt.title('Minimum_nights distribution')
plt.xticks(rotation=0);
```



We see that most of rentals are month rentals, followed by few 2 to 3 days rentals.

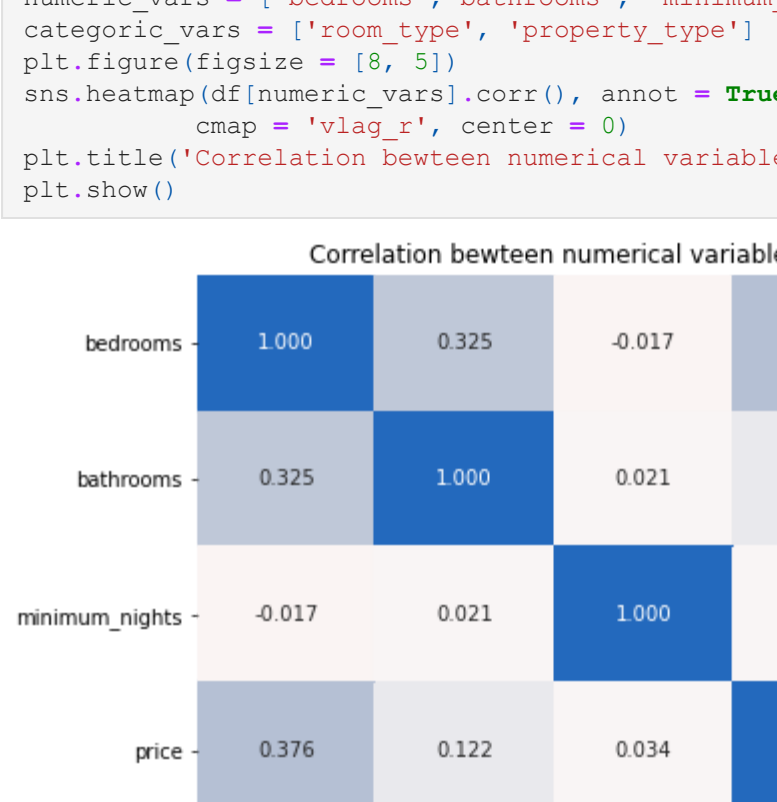
Now, let's move on to exploring the other categorical factors: room type, and property type.

```
In [7]: # univariate plot of rentals room type
base_color = sns.color_palette()[0]
type_order = df.room_type.value_counts().index
n_rentals = df.room_type.value_counts().sum()

# type_counts contains the frequency of unique values in the 'property_type' column in decreasing order.
type_counts = df.room_type.value_counts()

sns.countplot(data=df, x='room_type', color=base_color, order=type_order);
plt.title('Room type counts')

# Logic to print the proportion shape on the bars
for i in range(type_counts.shape[0]):
    count = type_counts[i]
    # Convert count into a percentage, and then into string
    pct_string = '{:0.1f}%'.format(100*count/n_rentals)
    # Print the string value on the bar.
    plt.text(i, count*50, pct_string, va='center')
```

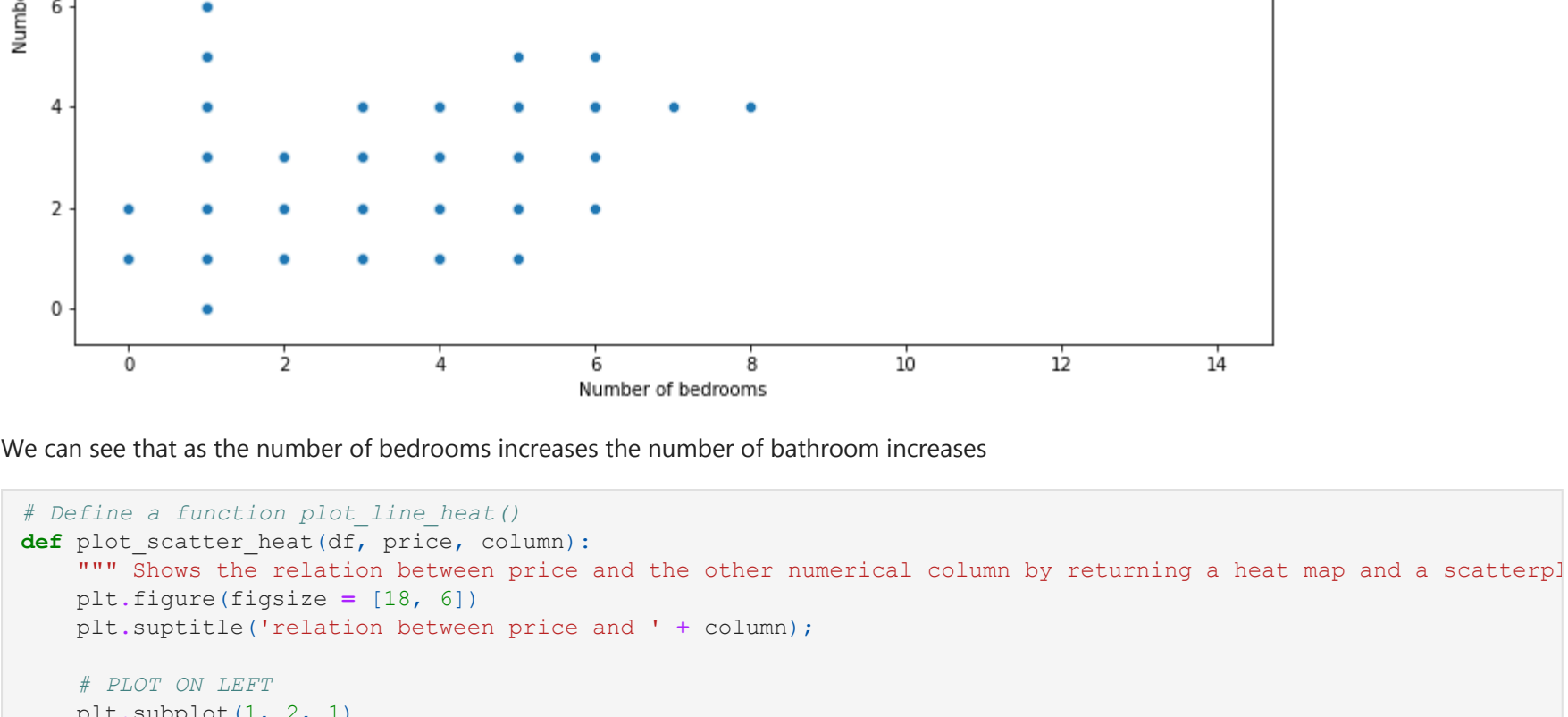


```
In [8]: # univariate plot of rentals property type
base_color = sns.color_palette()[0]
type_order = df.property_type.value_counts().index
n_rentals = df.property_type.value_counts().sum()

# type_counts contains the frequency of unique values in the 'property_type' column in decreasing order.
type_counts = df.property_type.value_counts()

sns.countplot(data=df, y='property_type', color=base_color, order=type_order);
plt.title('Property type counts')

# Logic to print the proportion shape on the bars
for i in range(type_counts.shape[0]):
    count = type_counts[i]
    # Convert count into a percentage, and then into string
    pct_string = '{:0.1f}%'.format(100*count/n_rentals)
    # Print the string value on the bar.
    plt.text(count+1, i, pct_string, va='center')
```

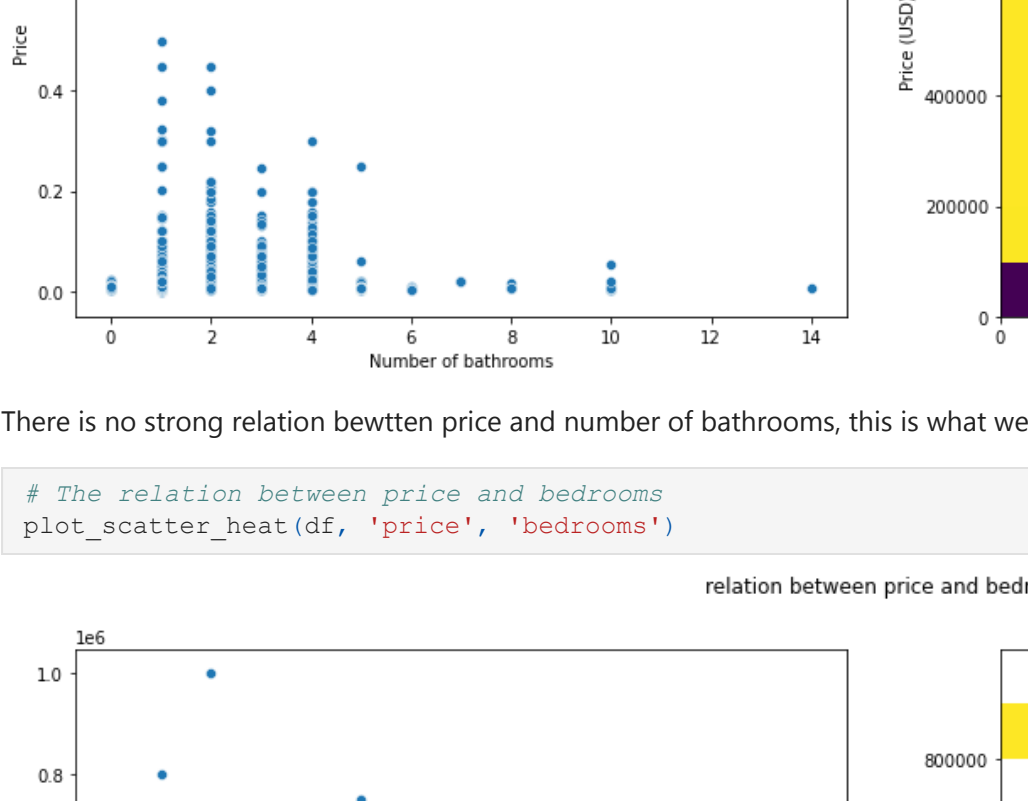


Most common room and property type are the apartment type for both the room and the property.

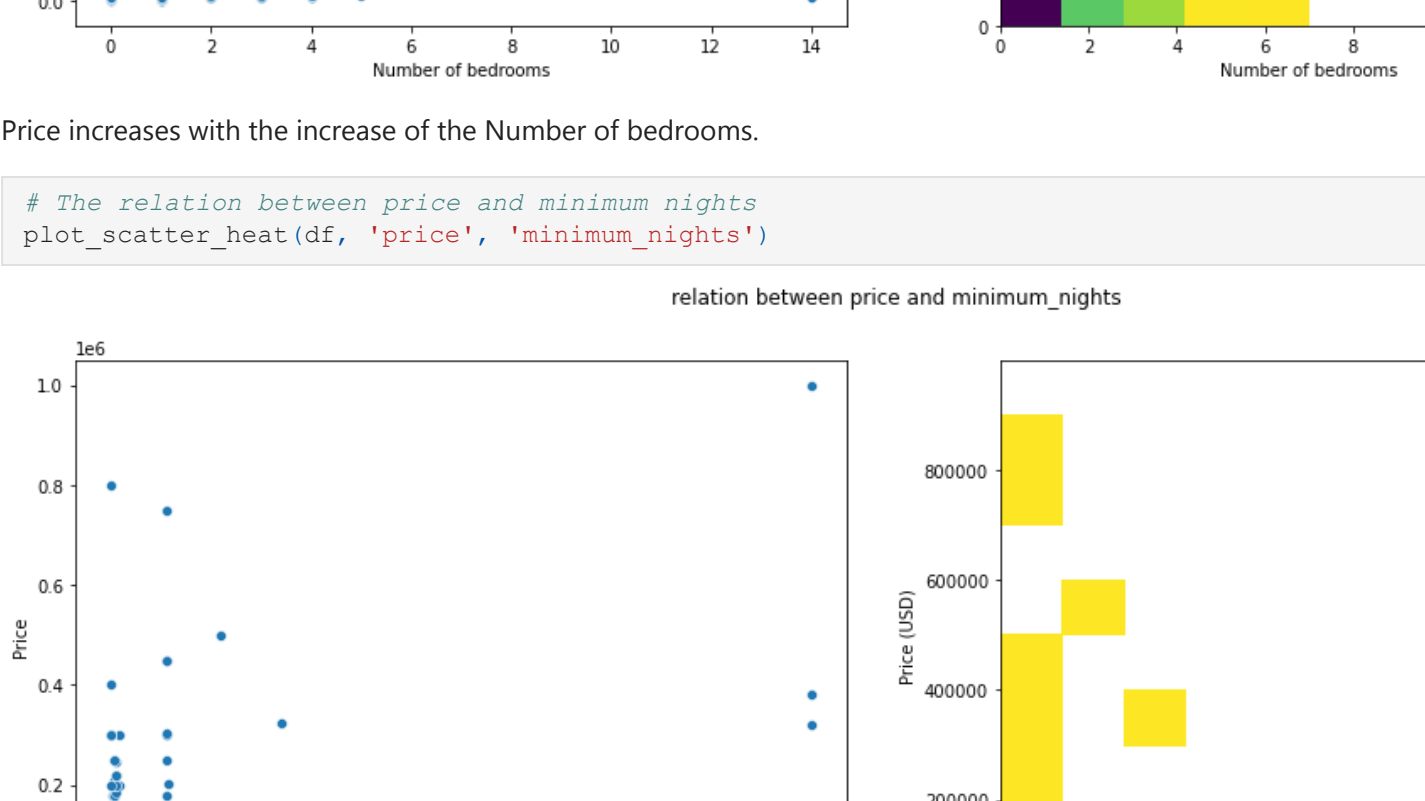
Bivariate Exploration

Lets start with the numerical values effect on price

```
In [9]: # correlation plot
numeric_vars = ['bedrooms', 'bathrooms', 'minimum_nights', 'price']
categorical_vars = ['room_type', 'property_type']
plt.figure(figsize=(8, 5))
sns.heatmap(df[numeric_vars].corr(), annot=True, fmt='.3f',
           cmap='viridis', center=0)
plt.title('Correlation between numerical variables')
plt.show()
```



```
In [10]: # Visualizing the relation between bedrooms and bathrooms
df.groupby(['room_type', 'property_type'])['price'].mean().plot(kind='bar')
plt.xlabel('Number of bedrooms')
plt.ylabel('Mean price (USD)')
plt.title('relation between bedrooms and bathrooms');
```



We can see that as the number of bedrooms increases the number of bathroom increases

```
In [11]: # Define a function plot_line_heat()
def plot_scatter_heat(df, price, column):
    """ Shows the relation between price and the other numerical column by returning a heatmap and a scatter plot.
    plt.figure(figsize=(18, 6))
    plt.suptitle('relation between price and ' + column);

    # PLOT ON LEFT
    plt.subplot(1, 2, 1)
    sns.scatterplot(data=df, x=column, y=price, ci=False)
    plt.xlabel('Number of ' + column)
    plt.ylabel('Price')

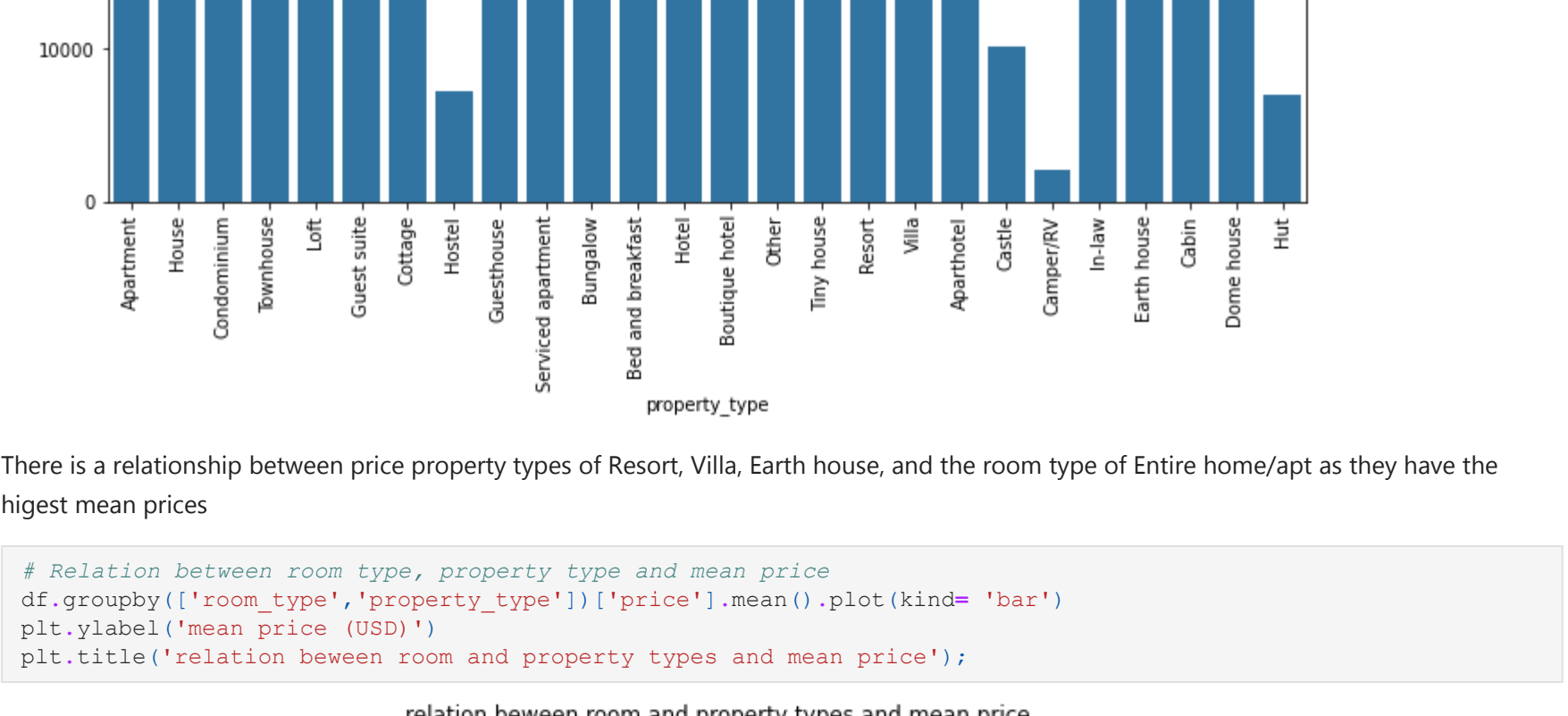
    # PLOT ON RIGHT
    plt.subplot(1, 2, 2)
    plt.hist2d(data=df, x=column, y=price, cmin=0.5, cmap='viridis_r')
    plt.colorbar()
    plt.xlabel('Number of ' + column)
    plt.ylabel('Price (USD)')
```

```
In [12]: # The relation between price and bathrooms
plot_scatter_heat(df, 'price', 'bathrooms')
```



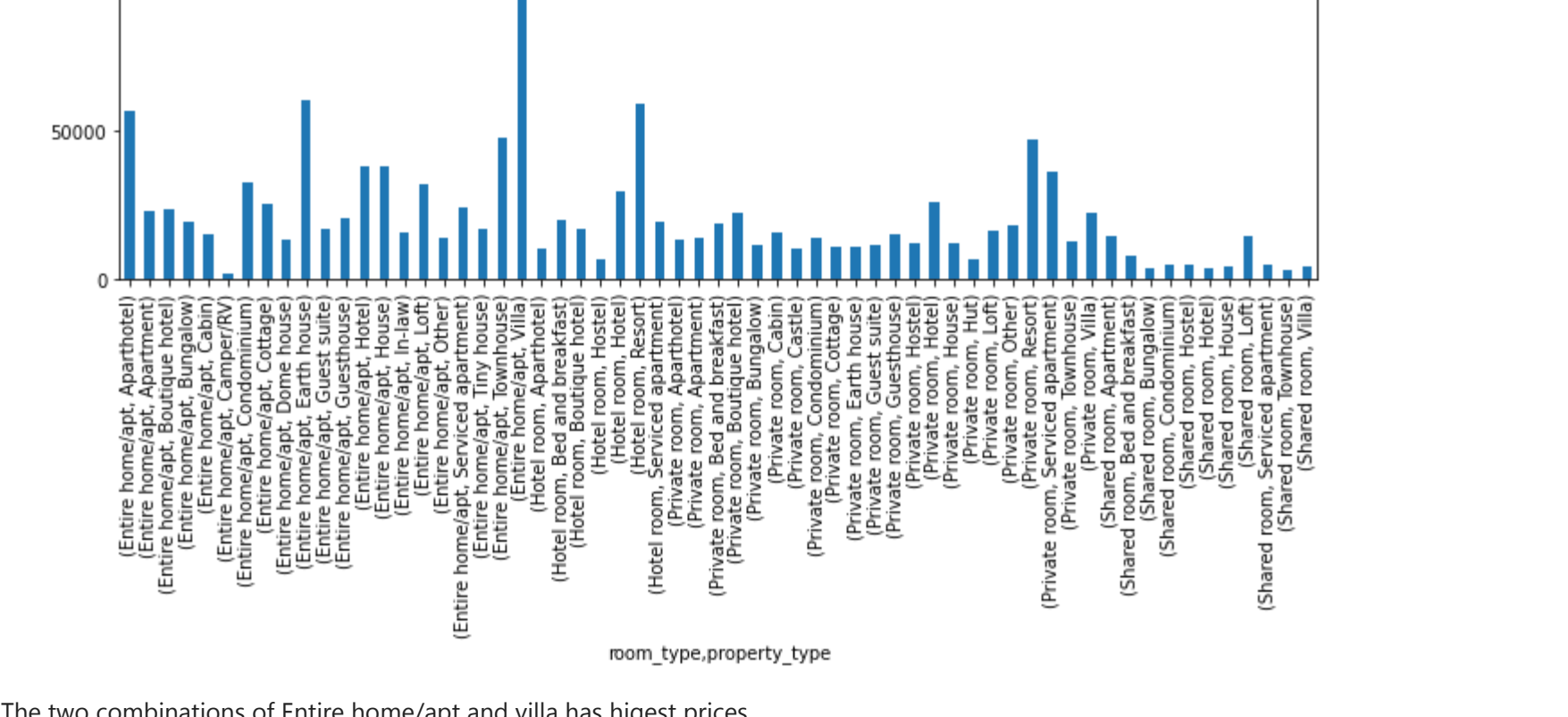
There is no strong relation between price and number of bathrooms, this is what we concluded before from the statistical approach.

```
In [13]: # The relation between price and bedrooms
plot_scatter_heat(df, 'price', 'bedrooms')
```



Price increases with the increase of the Number of bedrooms.

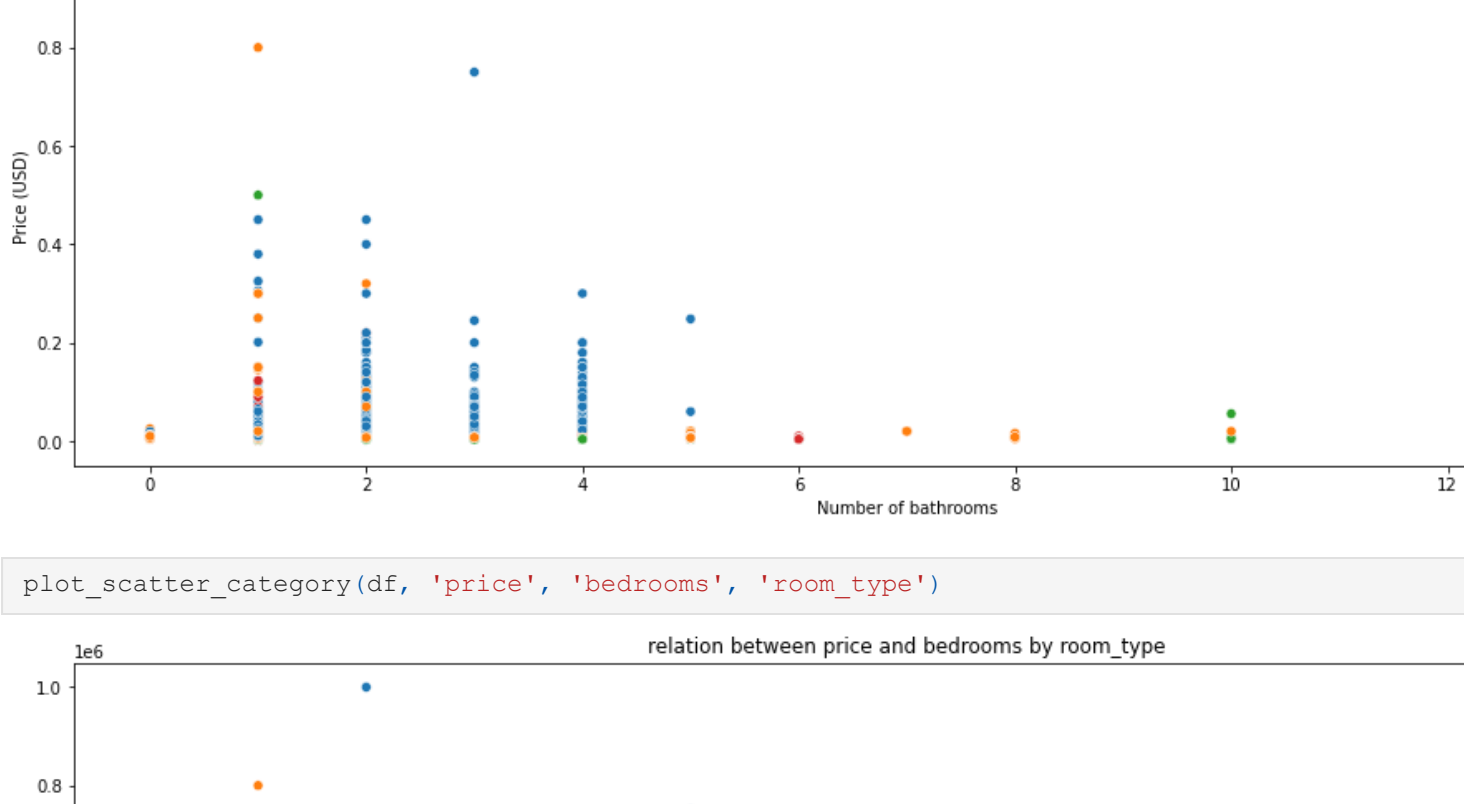
```
In [14]: # The relation between price and minimum_nights
plot_scatter_heat(df, 'price', 'minimum_nights')
```



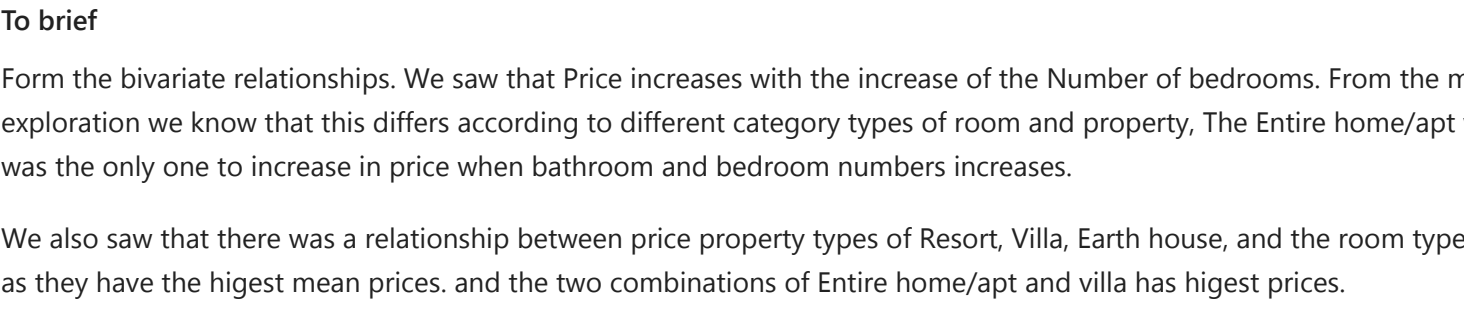
Number of minimum nights is less than a week and near to month are more common and prices.

Now let's take a look at the relationship between price and the other two categorical features

```
In [15]: # The relation between room type and mean price
base_color = sns.color_palette()[0]
sns.barplot(data=df, x='room_type', y='price', color=base_color)
plt.ylabel('mean price (USD)')
plt.title('relation between room type and mean price');
```



```
In [16]: # The relation between property type and price
base_color = sns.color_palette()[0]
sns.barplot(data=df, x='property_type', y='price', color=base_color, ci=False)
plt.xticks(rotation=90)
plt.ylabel('mean price (USD)')
plt.title('relation between property type and mean price');
```



There is a relationship between property types of Resort, Villa, Earth house, and the room type of Entire home/apt as they have the highest mean prices

```
In [17]: # Relation between room type, property type and mean price
df.groupby(['room_type', 'property_type'])['price'].mean().plot(kind='bar')
plt.xlabel('mean price (USD)')
plt.title('relation between room and property types and mean price');
```


The two combinations of Entire home/apt and villa has highest prices.

Multivariate Exploration

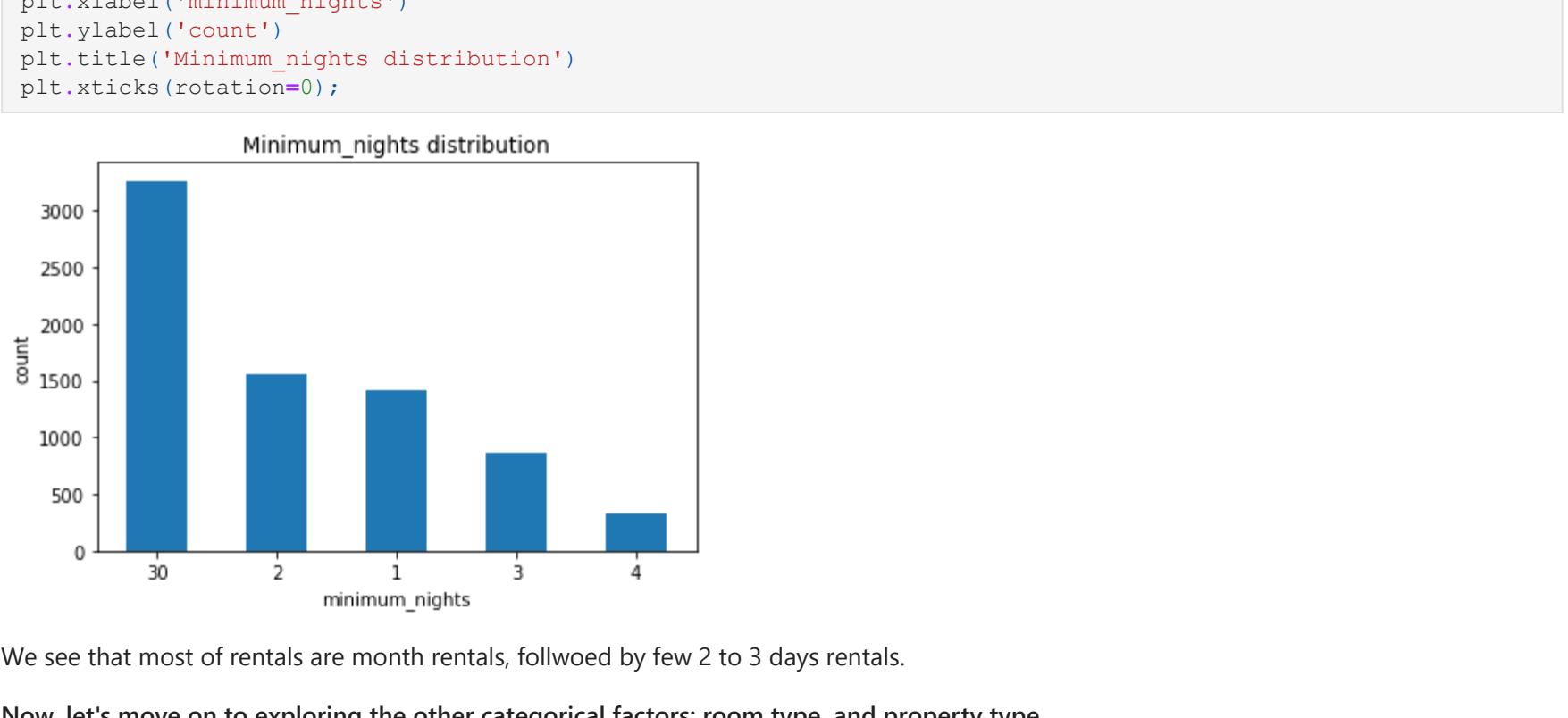
Lets create a plot of the relationship between price, bathrooms, bedrooms, and room types

```
In [18]: # Define a function plot_line_heat()
def plot_scatter_category(df, price, column, category):
    """ Shows the relation between price and the other numerical column according to a specific category type.
    plt.figure(figsize=(18, 6))
    plt.title('relation between price and ' + column + ' by ' + category);
    sns.scatterplot(data=df, x=column, y=price, hue=category, ci=False)
    plt.xlabel('Number of ' + column)
    plt.ylabel('Price (USD)')
```

```
In [19]: plot_scatter_category(df, 'price', 'bathrooms', 'room_type')
```



```
In [20]: plot_scatter_category(df, 'price', 'bedrooms', 'room_type')
```



To brief

From the bivariate relationships. We saw that Price increases with the increase of the Number of bedrooms. From the multivariate exploration we know that this differs according to different category types of room and property. The Entire home/apt was the room type was the only one to increase in price when bathroom and bedroom numbers increases.

We also saw that there was a relationship between price property types of Resort, Villa, Earth house, and the room type of Entire home/apt as they have the highest mean prices. and the two combinations of Entire home/apt and villa has highest prices.

