

01 BACKGROUND

02 APPROACH

03 OUTCOMES

04 NEXT STEPS



01 BACKGROUND

Why are we here today?

02 APPROACH

03 OUTCOMES



### AIRBNB IS THE LEADING PLATFORM FOR RENTING OUT PRIVATE PROPERTY

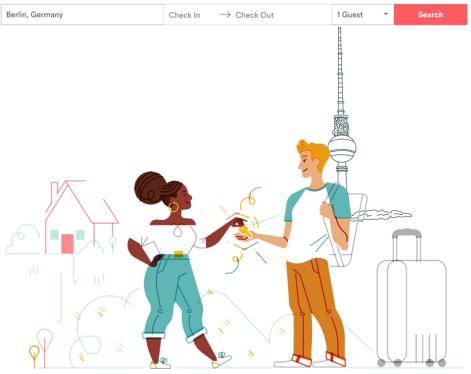
### **ABOUT AIRBNB**

Online platform offering vacation rentals for travelers - primarily homestays - as an alternative to traditional hotel or hostel stays.

Do not own or host themselves but instead act as an intermediary broker

Have ventured into complementary businesses, such as offering **experiences** at major cities

### Search over 2 million homes in over 190 countries



### CURRENTLY USERS CAN ONLY GET A PRICING ESTIMATE DURING LISTING CREATION

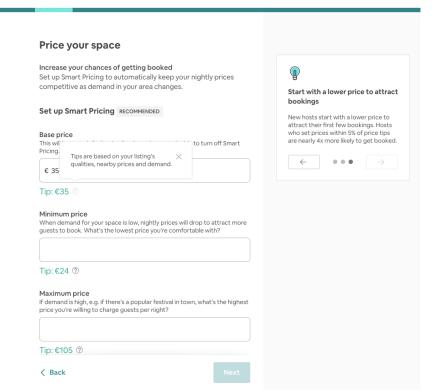
### **OPPORTUNITY**



Airbnb does not publicly publish **booking data** as this presents one of its core assets

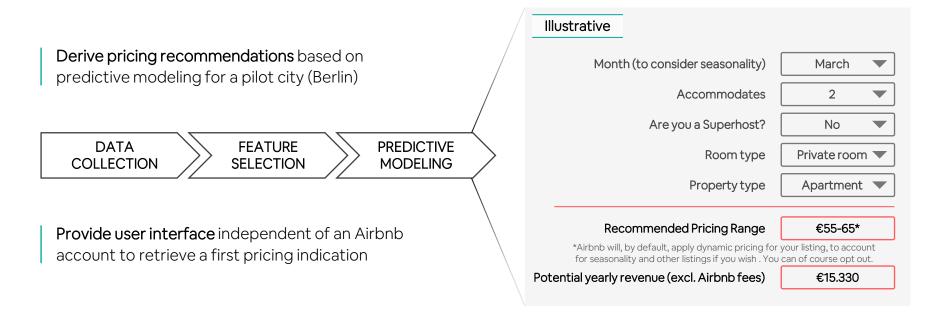
Users receive a pricing indication as part of the listing creation process

Hence it is difficult for individuals to **estimate a** good listing price before account creation



### THIS PROJECT AIMS AT CREATING A TOOL PROVIDING AN UPFRONT PRICING ESTIMATE

**CORE IDEA** 



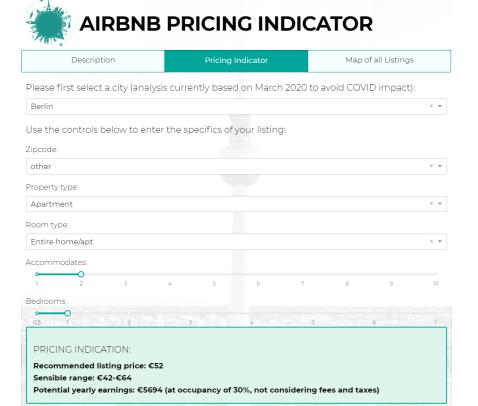
### THIS TARGET HAS NOW BEEN IMPLEMENTED IN THE FORM OF AN INTERACTIVE WEB APP

**OUTCOME** 

**Pricing recommendation** and potential earnings estimate based on up to 24 listing features

Interactive input fields with sensible default values based on average listings

**Dynamically changing listing recommendation** as soon as input values are adapted



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02 APPROACH

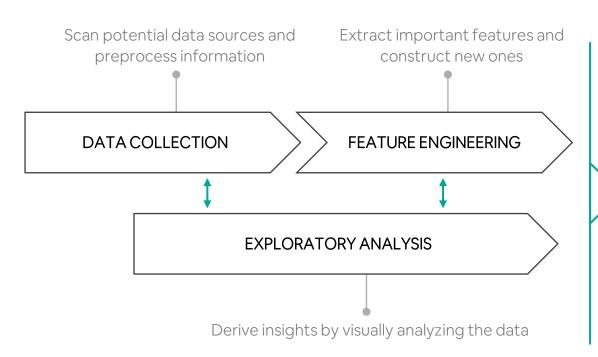
How was the problem tackled?

03 OUTCOMES



## A STRUCTURED APPROACH WAS FOLLOWED FOR CREATING THE UNDERLYING MODEL

### **APPROACH**



### **FEATURE SELECTION**

Define feature set for predictive modeling

### PREDICTIVE MODELING

Train machine learning models, evaluate and make predictions

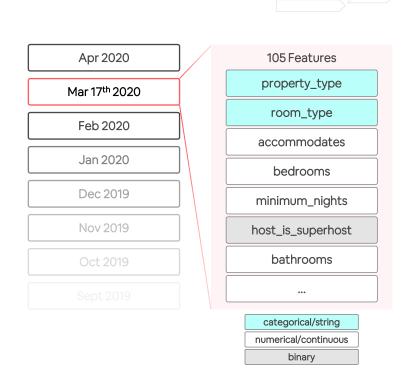
### DATA AVAILABLE AT INSIDEAIRBNB.COM LAID THE FOUNDATION FOR THE ANALYSIS

**DATA COLLECTION** 

>25.000 listings with 105 features each were retrieved for Berlin on March 17<sup>th</sup> 2020

Dataset was cleaned extensively (e.g. inactive listings, errors), reducing eventually used listings to slightly above 10.000

**Features were removed** even before data cleaning and feature engineering



### THE DATA HAS SUBSTANTIAL LIMITATIONS, BUT STILL OFFERS A RICH SET OF FEATURES

### **DATA COLLECTION**

Structured provision of publicly available data scraped from all listings at a given date

Key target and predictors available in the form of price and main listing parameters (guests, etc.)

Monthly scraping routine theoretically enables automatized actualization of prediction

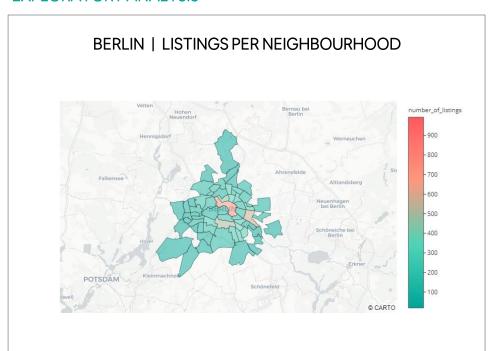
No actual booking data provided as this information is not publicly available

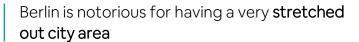
No occupation rates provided as a consequence, making only vague assumptions possible

**Price is only listed price** and might not in all cases reflect the final price at booking

## A BROAD OVERVIEW OF BERLIN HIGHLIGHTS SOME CLEAR HOTSPOTS IN TERMS OF LISTINGS

### **EXPLORATORY ANALYSIS**



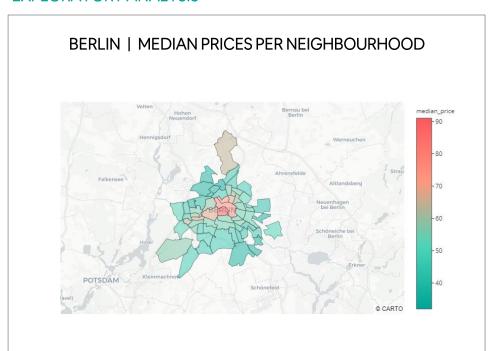


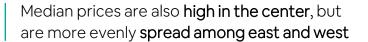
Very few neighbourhoods cover a very large portion of listings

The biggest hotspots are located in the **centre** and east side of Berlin

### HIGH PRICES ARE NOT DIRECTLY RELATED TO LISTING HOTSPOTS AND HIGHER IN THE WEST

### **EXPLORATORY ANALYSIS**





The central area with the **highest median price** is not one of the listing hotspots

Certain **outside areas** with few listings also have relatively **high median prices** 

Similarly, when analyzing **zipcodes** there is a clear span in median prices (see Appendix)

# ONE KEY FACTOR WITH HIGH CORRELATION TO PRICE IS THE NUMBER OF POTENTIAL GUESTS

### **EXPLORATORY ANALYSIS**

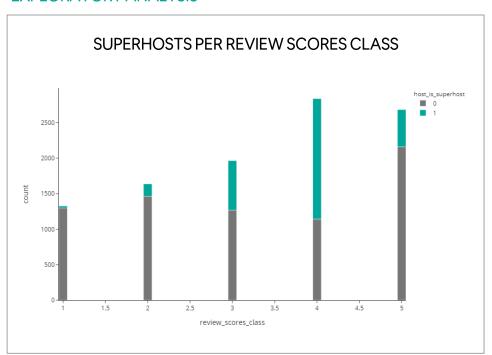


Unsurprisingly, the number of accommodated people per listing appears to be one of the key predictors for price

There is a visible and strong **positive relationship** between the number of people hosted and listing price

### HIGH RATINGS ARE NOT AUTOMATICALLY RELATED TO MORE SUCCESSFUL LISTINGS

### **EXPLORATORY ANALYSIS**





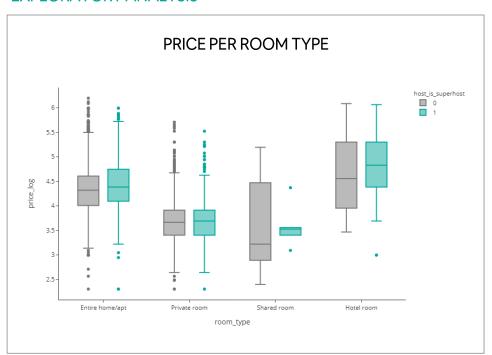
The **low occurrence of superhosts** in class 5 seems to strengthen this assumption

Further analysis reveals perfect ratings are related to newer listings (see Appendix)

Such insights need to be **considered and potentially treated** during feature engineering

### ROOM TYPE, ON THE OTHER HAND, IS CLEARLY A STRONG PREDICTOR OF PRICE

### **EXPLORATORY ANALYSIS**





There is a clear **price distinction** between the two major room types **entire home/apt** and **private room** 

While this follows basic intuition, the difference is quite substantial

Hotel rooms post the highest prices while shared rooms are on the low end

### APARTMENTS ON AVERAGE ARE HIGHER PRICED COMPARED TO HOUSES

### **EXPLORATORY ANALYSIS**

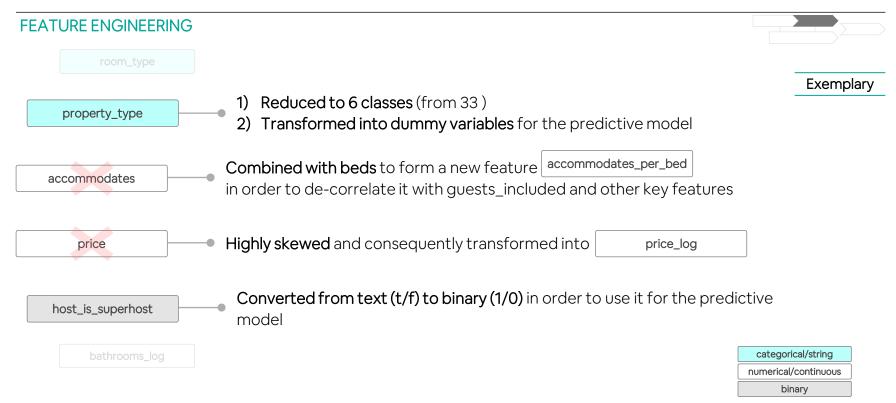




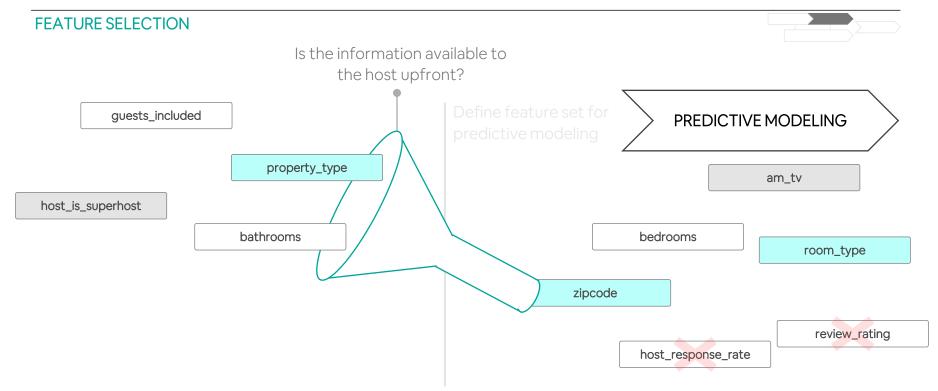
Apartments post higher prices compared to houses, which is most probably a result of superior location of the former

**Superhosts** are not necessarily associated with higher prices, which indicates that they are spread across the whole pricing spectrum

# VARIOUS FEATURES HAVE BEEN ADAPTED OR COMBINED BASED ON EXPLORATORY ANALYSIS



# INFORMATION AVAILABLE TO USERS PLAYED A KEY ROLE IN SELECTING THE FINAL FEATURES



# THE SELECTED FEATURES ENABLE A MODEL CAPABLE OF PROVIDING A FIRST INDICATION

### FEATURE IMPORTANCE AND PREDICTIVE MODELING Room type: Private room Guests included in price **Bedrooms** Room type: Shared room Property type: Boutique hotel Minimum nights Best model: XGBoost Median Average Percentage Error on price log: 5,18%

Bar length indicates feature importance, but does not represent the precise relationship in weights

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What has been achieved?



### THE MODEL IS APPLIED IN AN INTERACTIVE WEB APPLICATION ACCESSIBLE VIA ANY BROWSER

### **WEB APPLICATION**



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Where does it go from here?



### THE RESULTS OPEN UP COUNTLESS POSSIBILITIES FOR FURTHER EXPLORATION

### **FUTURE WORK**

### Predictive modeling

Apply further models and adapt current ones (e.g. NN) Examine other prediction targets (e.g. occupancy rate)

### Feature engineering

Explore NLP for text fields (descriptions, reviews, ...)
Scrape listing photos and analyze quality
Enhance current feature set

### Lean structure

Remove remaining redundancies wherever possible (e.g. pack repeated steps into functions, apply more pipelines, ...)

### Cloud

Move both model creation and app to GCP

### **Automatization**

Build a workflow to automatically retrain model monthly with new datasets

Use automated outlier detection

Use automated feature engineering

Let user input exact address and determine location

### Replicability

Apply analysis to even more cities and compare results

### Book your next vacation in ...

... a Windmill close to Amsterdam



https://www.airbnb.com/rooms/2107131

### ... an airstream



... a potato (?)



### Thank You!

Mauricio Malzer

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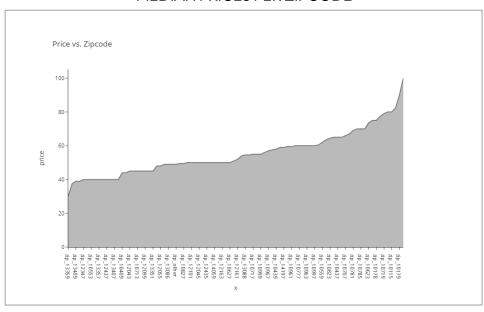
### **APPENDIX**



### MEDIAN PRICE VARIES SUBSTANTIALLY AMONG DIFFERENT ZIPCODES

### **EXPLORATORY ANALYSIS**

### MEDIAN PRICES PER ZIPCODE



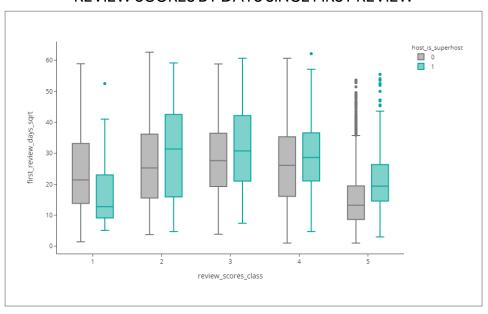
- Most zipcodes have a median price between €40 and €60
- A few primarily central and western zipcodes stand out in terms of high median prices

# HIGH RATINGS MOST FREQUENTLY OCCUR WITH VERY NEW LISTINGS

### **EXPLORATORY ANALYSIS**



### REVIEW SCORES BY DAYS SINCE FIRST REVIEW



- High ratings (class 5) also appear to be the newest listings
- Superhosts have, on average, been on the platform longer than their peers





Overview Setup Safety Financials Hosting services

The following pages lead through the listing creation process, highlighting the corresponding features in the predictive model

Features



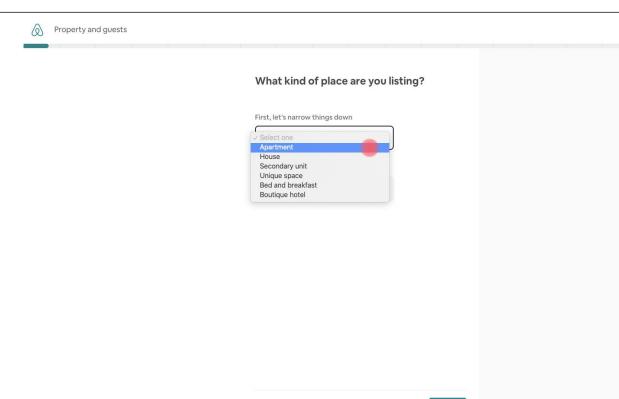
### Why host on Airbnb?

No matter what kind of home or room you have to share, Airbnb makes it simple and secure to host travelers. You're in full control of your availability, prices, house rules, and how you interact with guests.

### We have your back

To keep you, your home, and your belongings safe, we cover every booking with \$1M USD in property damage protection and another \$1M USD in insurance against accidents.

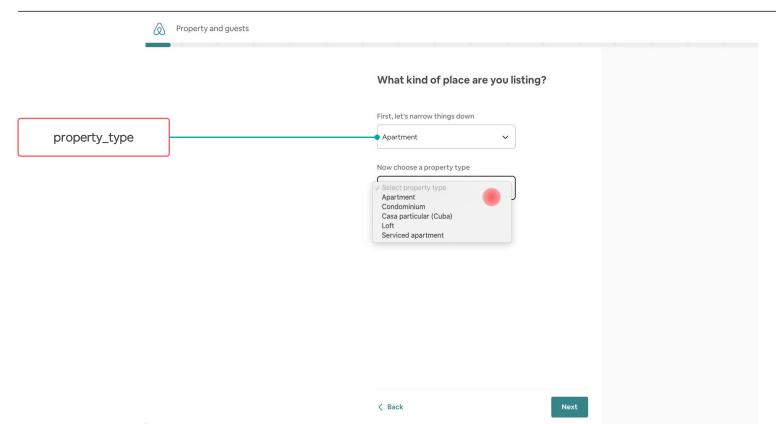




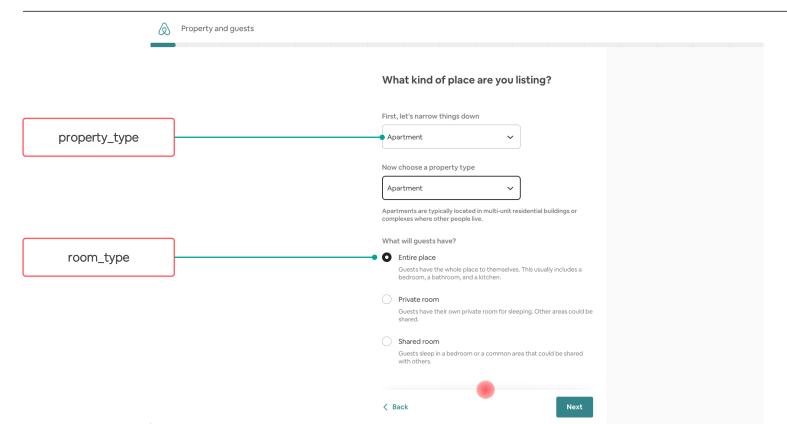
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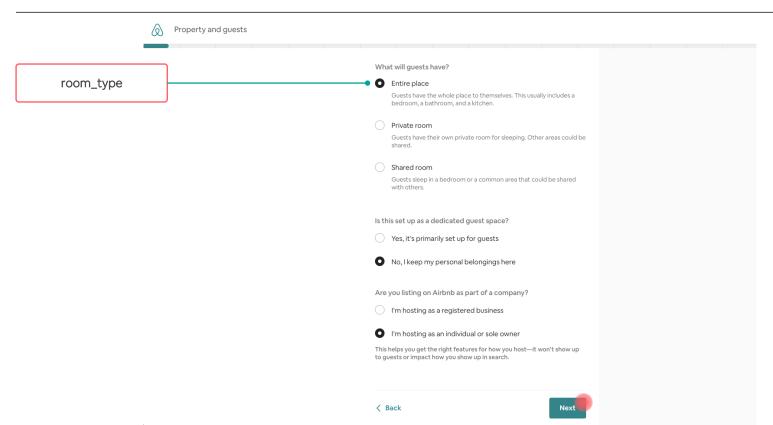




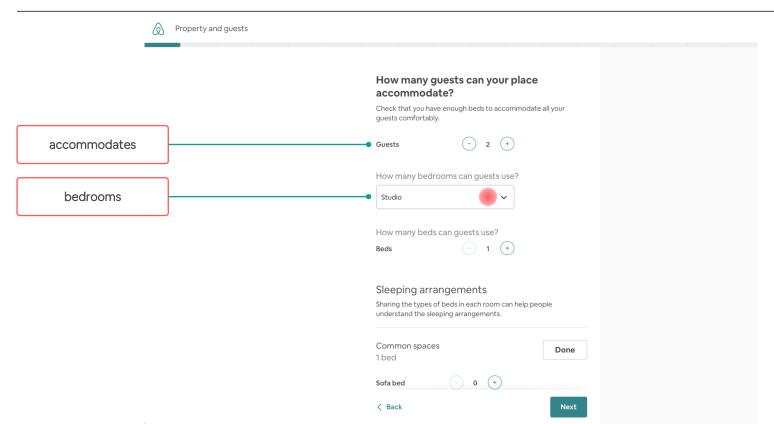




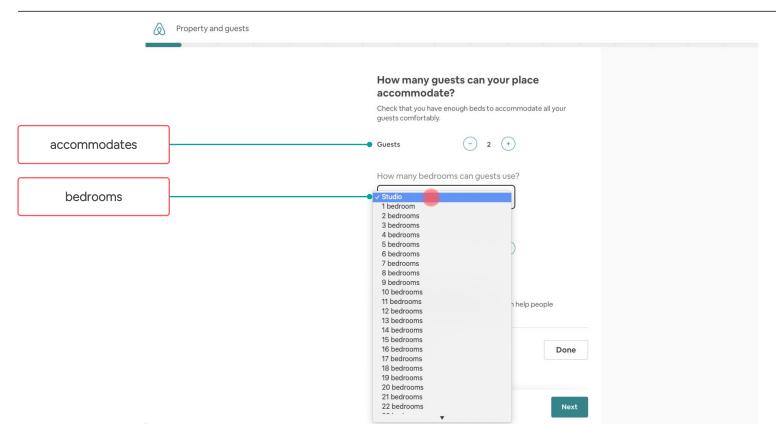












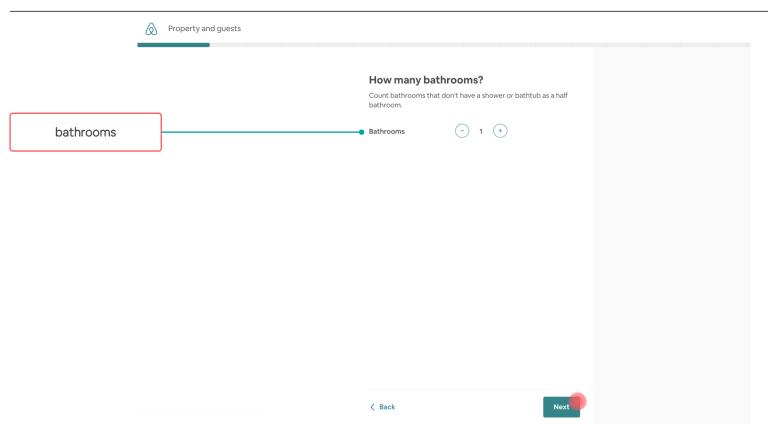


<b></b>	Property and guests					
		Studio		~		
		How many beds ca	5711	1 (+)		
	Sleeping arrangements Sharing the types of beds in each room can help people understand the sleeping arrangements.					
beds	•	Common spaces 0 beds			Done	
	,	Sofa bed	<u> </u>	+		
		Couch	<u> </u>	+		
		Floor mattress	- 0	+		
		Bunk bed	<u> </u>			
		Add another bed		~		
		< Back			Next	



<u> </u>	Property and guests				
		Studio	~		
		How many beds can gue	ests use?		
		Sleeping arrangem Sharing the types of beds in understand the sleeping arra	each room can help peopl	е	
beds	•	Common spaces 1 bed		Done	
		Sofa bed	0 (+)		
		Couch -	0 (+)		
		Bunk bed -	1 (+)		
		Add another bed	~		
		< Back		Next	

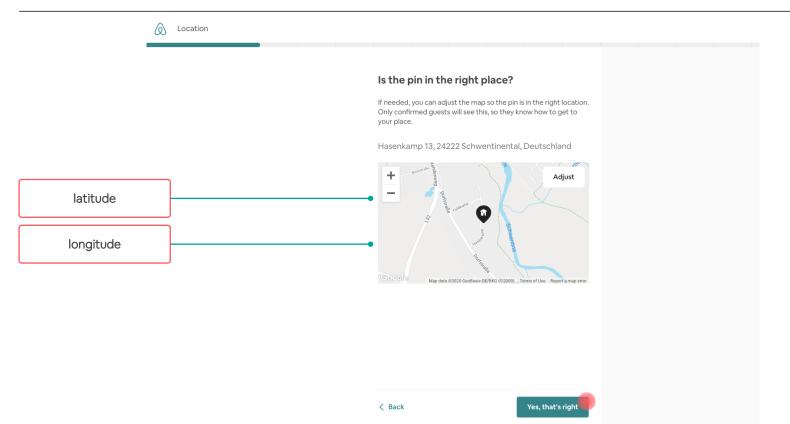




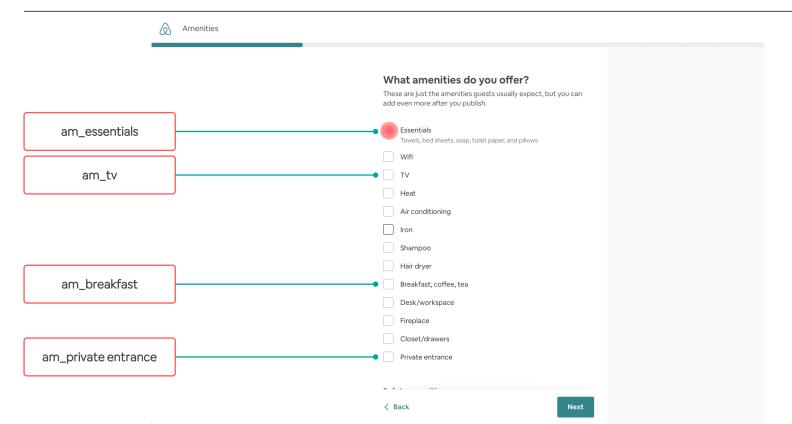


<u></u>	Location					
			Where's your pla Guests will only get your reservation.	exact address once they've booked a		
			Country / Region			
			Germany Street address	<u> </u>		
			Hasenkamp 13			
			z. B Kurfürstendamm 67  Apt, suite. (optional)			
			z. B Gebäude 1			
			City	ZIP code		
zipcode –			Sehwentinental	24222		
			z. B Berlin	z. B 10719		
			< Back	Next		

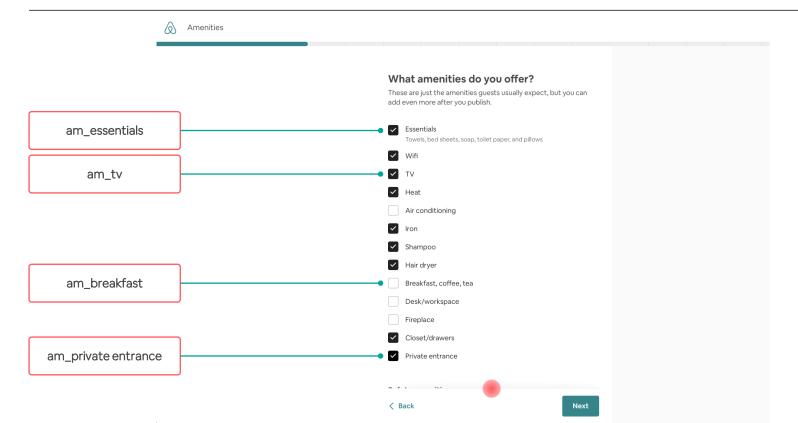




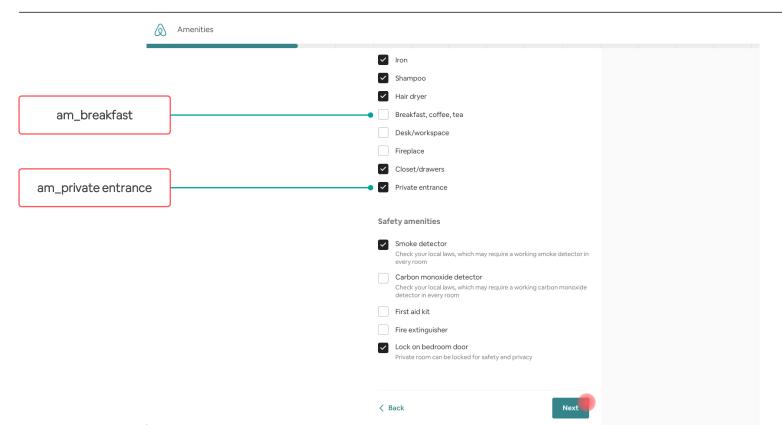




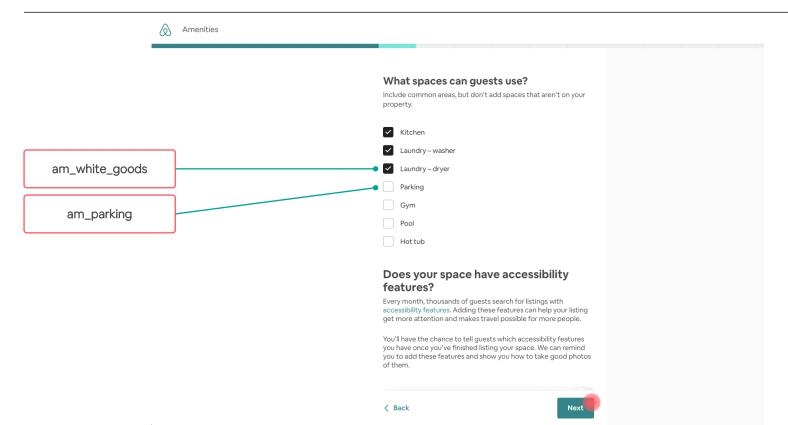
















#### Liven up your listing with photos

Take photos using a phone or camera. Upload at least one photo to publish your listing—you can always add more or edit later.

Upload Photos

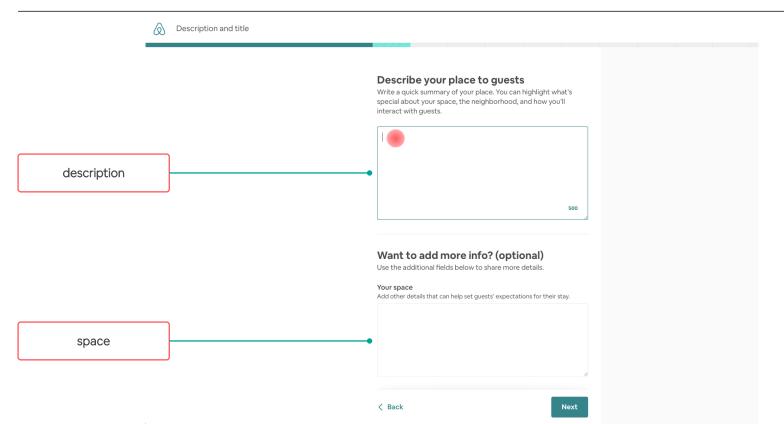
Or drag them in from your desktop



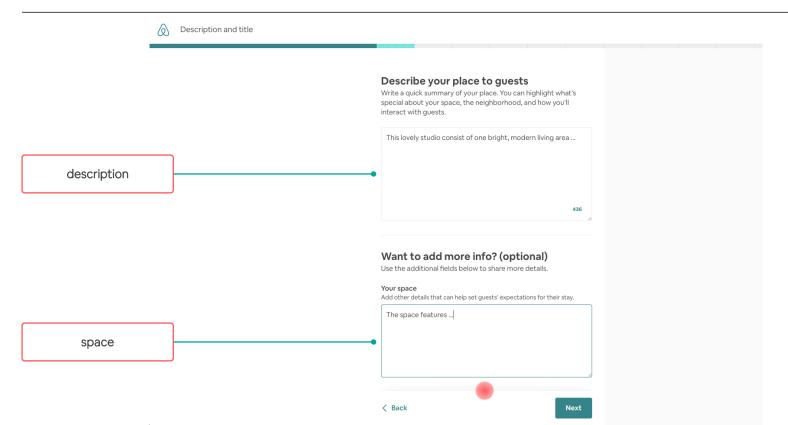
#### Quick tips for quality photos

- · Declutter your space
- Use natural daylight and avoid flash
- Shoot from room corners in landscape mode
- Balance visible floor and ceiling space
- Highlight special decor and amenities
- Add photos of every room guests can access

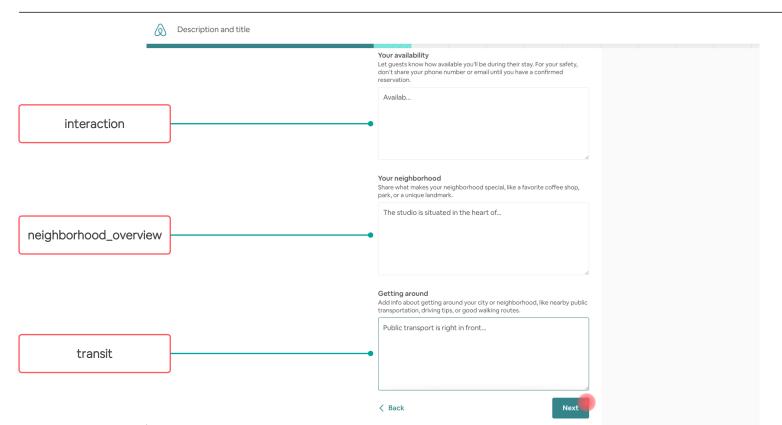




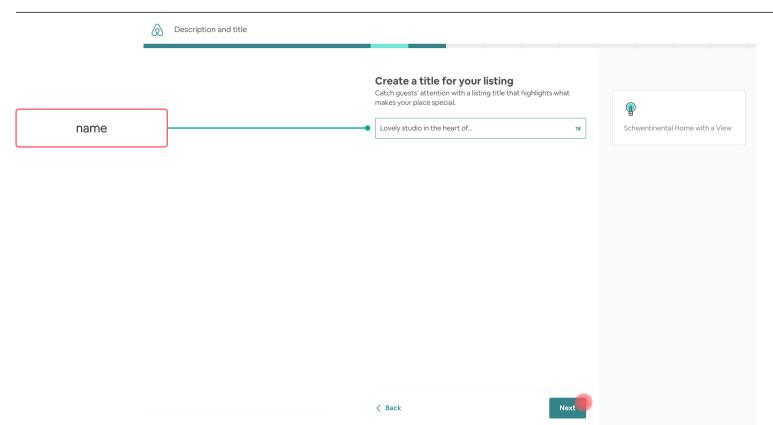








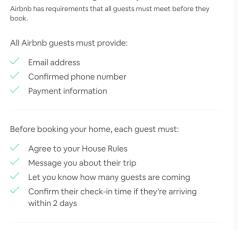








Booking settings

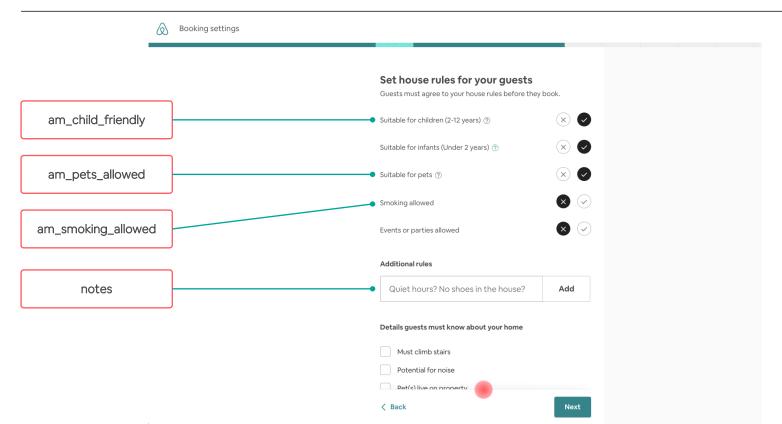


Add additional requirements

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Review Airbnb's guest requirements

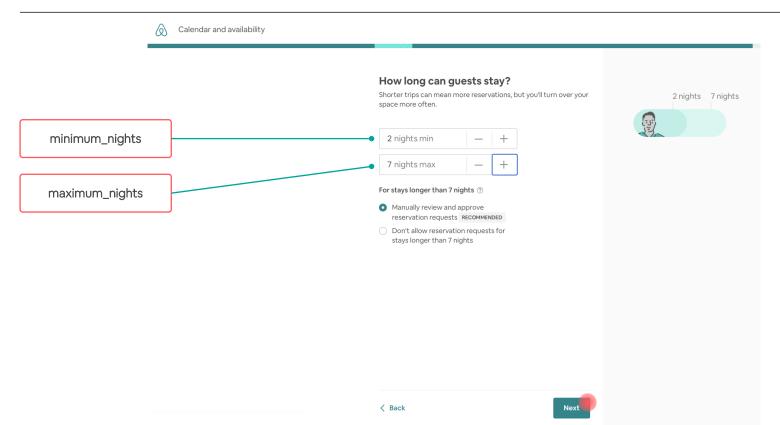






	٨	Booking settings			
			Additional rules		
notes		•	Quiet hours? No shoes in the house?	Add	
			Details guests must know about your home		
			Must climb stairs		
			Potential for noise		
			Pet(s) live on property		
			✓ No parking on property		
			Describe the parking situation around your listing		
			Parking on the streets		
			Some spaces are shared		
			Amenity limitations		
			Surveillance or recording devices on property		
			Weapons on property		
			Dangerous animals on property		
			< Back	Next	

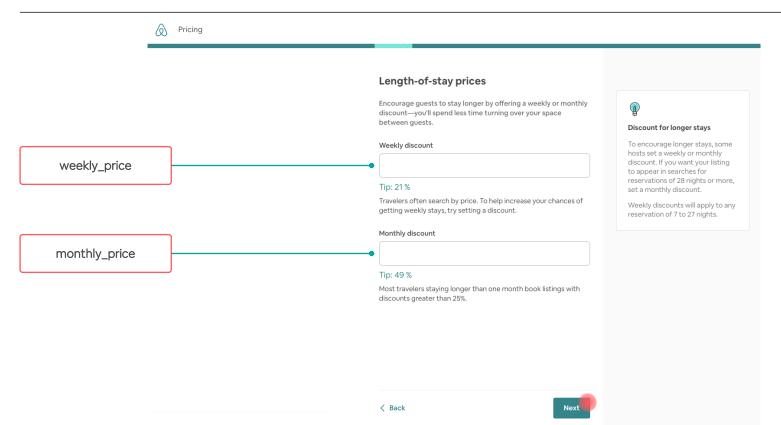






Price your space  Increase your chances of getting booked Set up Smart Pricing to automatically keep your nightly prices competitive as demand in your area changes.  Set up Smart Pricing RECOMMENDED  Base price This will Pricing. Tips are based on your listing's  qualities, nearby prices and demand.  Tips are based on your listing's  qualities, nearby prices and demand.  Tips are based on your listing's  qualities, nearby prices and demand.  Tips are based on your listing's  qualities, nearby prices and demand for your space is low, nightly prices will drop to attract more guests to book. What's the lowest price you're comfortable with?		Pricing		
Set up Smart Pricing to automatically keep your nightly prices competitive as demand in your area changes.  Set up Smart Pricing RECOMMENDED  Set up Smart Pricing RECOMMENDED  Base price This will Pricing.  Tips are based on your listing's  qualities, nearby prices and demand.  Tip: €35 ⑦  Minimum price When demand for your space is low, nightly prices will drop to attract more			Price your space	
Minimum price When demand for your space is low, nightly prices will drop to attract more	price	•	Increase your chances of getting booked Set up Smart Pricing to automatically keep your nightly prices competitive as demand in your area changes.  Set up Smart Pricing RECOMMENDED  Base price This wil to turn off Smart Pricing. Tips are based on your listing's your guildies, nearby prices and demand.	Start with a lower price to attract bookings  New hosts start with a lower price to attract their first few bookings. Hosts who set prices within 5% of price tips are nearly 4x more likely to get booked.
guests to book. What's the lowest price you're comfortable with?			Minimum price	
			guests to book. What's the lowest price you're comfortable with?	
Tip: €24 ⑦  Maximum price  If demand is high, e.g. if there's a popular festival in town, what's the highest price you're willing to charge guests per night?			Maximum price  If demand is high, e.g. if there's a popular festival in town, what's the highest	
Tip: €105 ⑦  < Back  Next				





#### THIS SUMMARY SHOWS THE FEATURES RETRIEVED DURING LISTING CREATION

#### OTHER SOURCES (E.G. PROFILE)

property\_type

room\_type

accommodates

bedrooms

beds

bathrooms

zipcode

latitude

longitude

am essentials

am\_tv

am\_breakfast

LISTING CREATION

am\_private entrance

am\_white\_goods

am\_parking

description

space

interaction

neighborhood\_overview

transit

name

am\_child\_friendly

am\_pets\_allowed

am\_smoking\_allowed

notes

minimum\_nights

maximum\_nights

price

weekly\_price

monthly\_price

host\_is\_superhost

host\_identity\_verified

is\_location\_exact

security\_deposit

cleaning\_fee

availability\_365

review\_scores\_rating

instant\_bookable

cancellation\_policy

calc\_host\_listings

numerical/continuous

categorical/string

binary