Search Ranking and Personalization at Airbnb

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SEARCH RANKING TEAM



Airbnb

Worldwide Accommodations Leader

- 4M listings in 190 Countries
- 2M guest stay at Airbnb every night

2-sided Marketplace:

- Hosts: looking to rent their space
- Guests: looking for a place to stay

Our job:

Connect guests with hosts

Search At Airbnb

query: location, dates, guests, map, filters





results:



From \$79 · Perfectly located Castro
Private room · 1 bed

★★★★ 335 · Superhost



From \$149 · Grand and Cozy 1920's SF Studio Entire apartment · 2 beds

★★★★ 280 · Superhost



From \$185 · Garden Retreat steps from Haight St Entire house · 2 beds

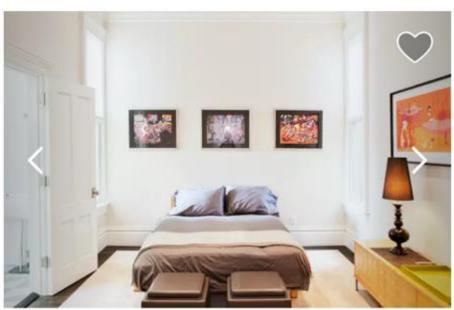
★★★★ 186 · Superhost

results:



From \$105 · Charming SF 1911 Studio
Entire condominium · 1 bed

** * * * * 136 · Superhost



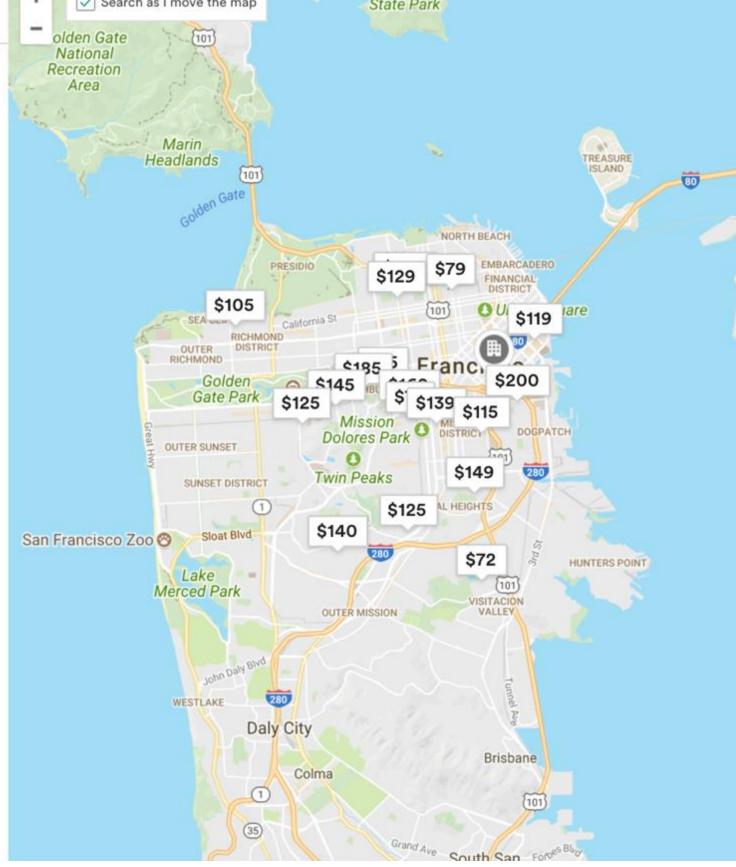
From \$129 · Upscale Private Ensuite Bed & Bath
Private room · 1 bed

★★★★ 289 · Superhost



From \$350 · Edwardian Style Pacific Heights 1BD
Entire apartment · 1 bed

*****56

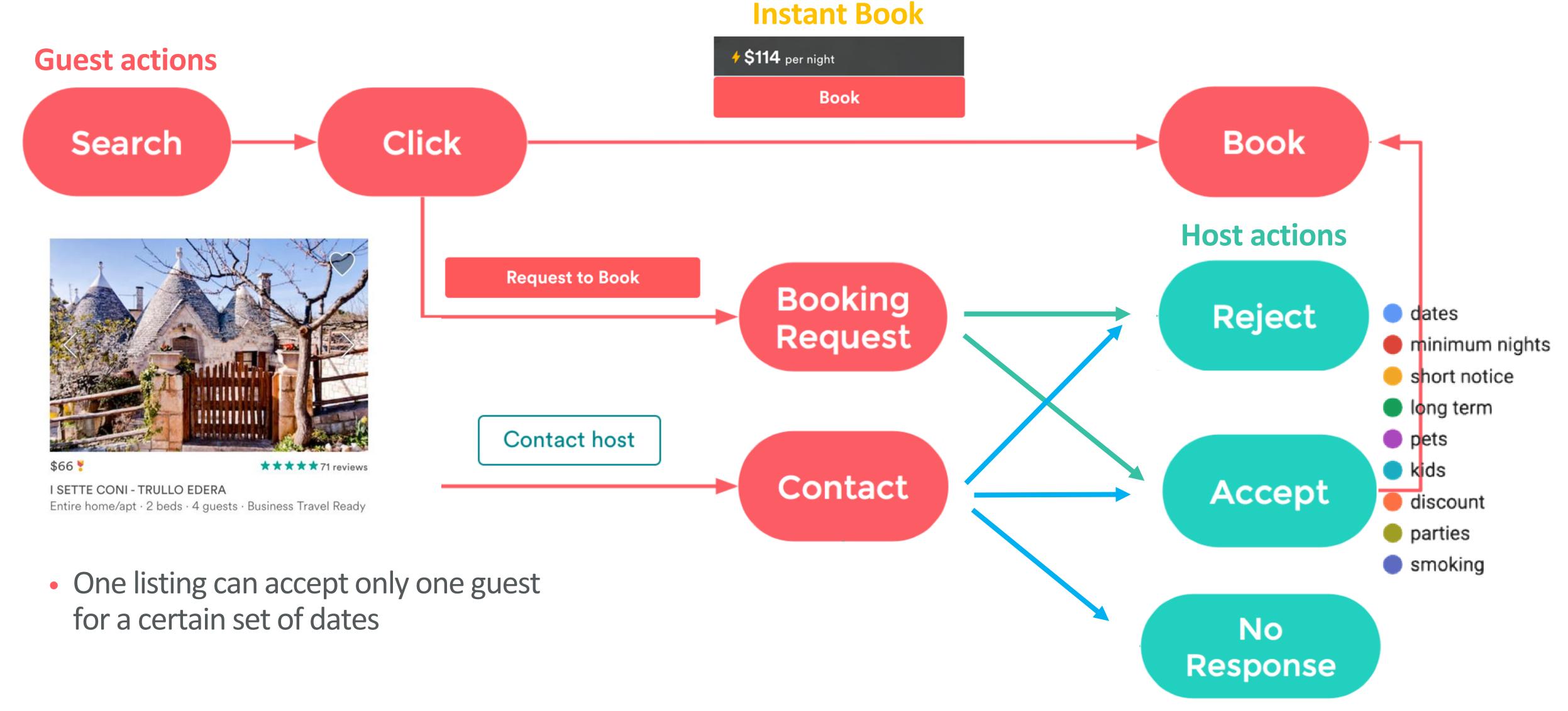


Search At Airbnb

Search Components

- Retrieval of Listings
 - Location Queries (e.g. New York): Rule based: fetch everything near by
 - Broad Queries (e.g. Skiing in France): Model based
- Ranking of Listings
 - Model based: query, user and listing features
 - Trained using user actions: clicks, wishlists, inquires, bookings and rejections
 - Tradeoffs Guest and Host preferences

Typical Guest Flow



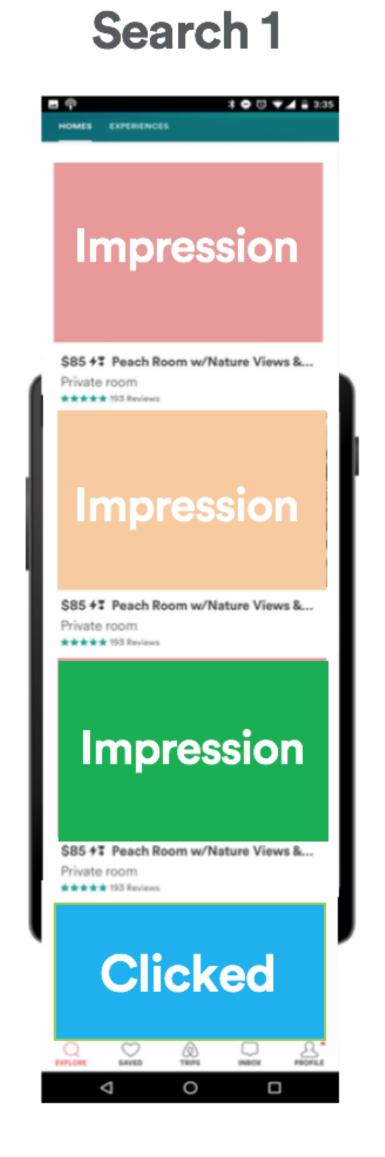
Guest Sessions

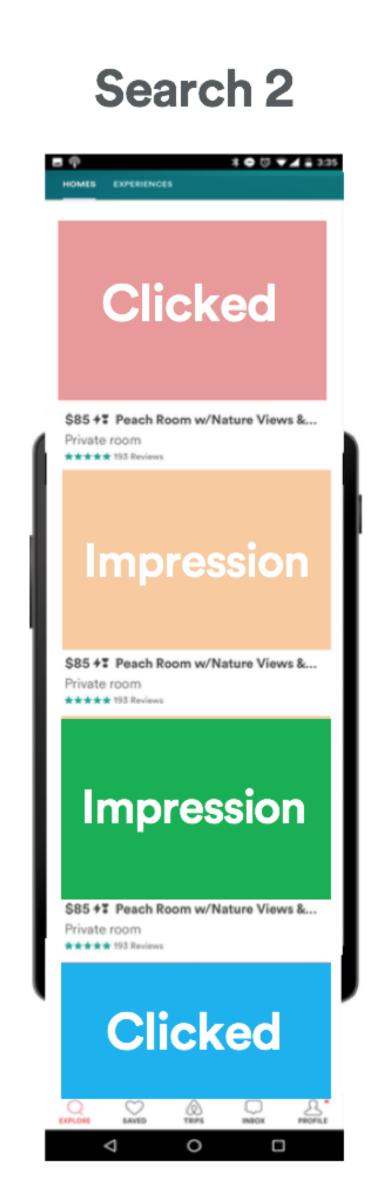
Average user does dozens of searches over multiple days before they book

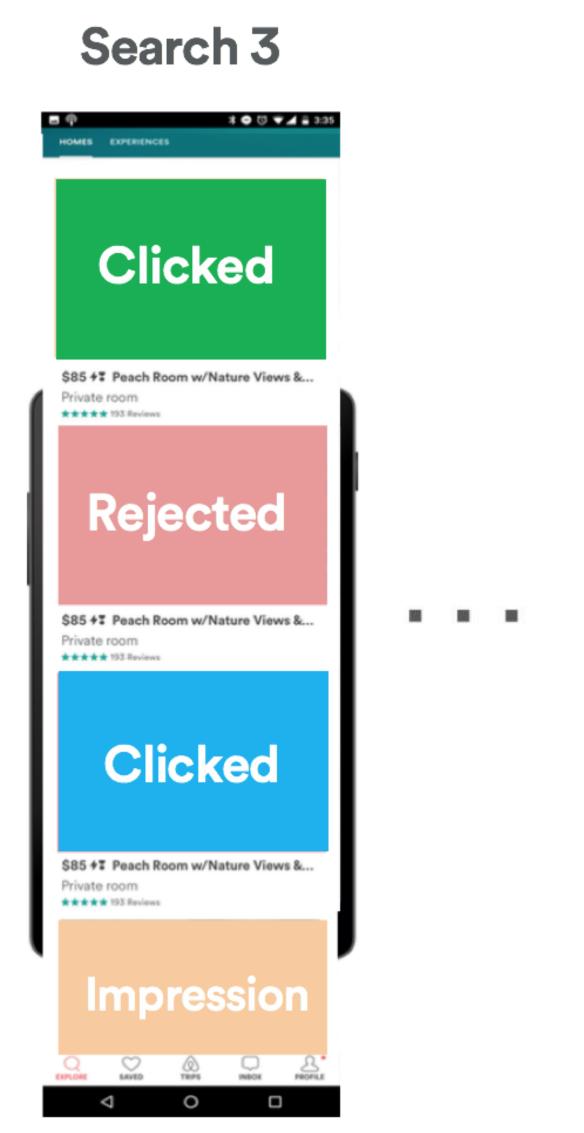


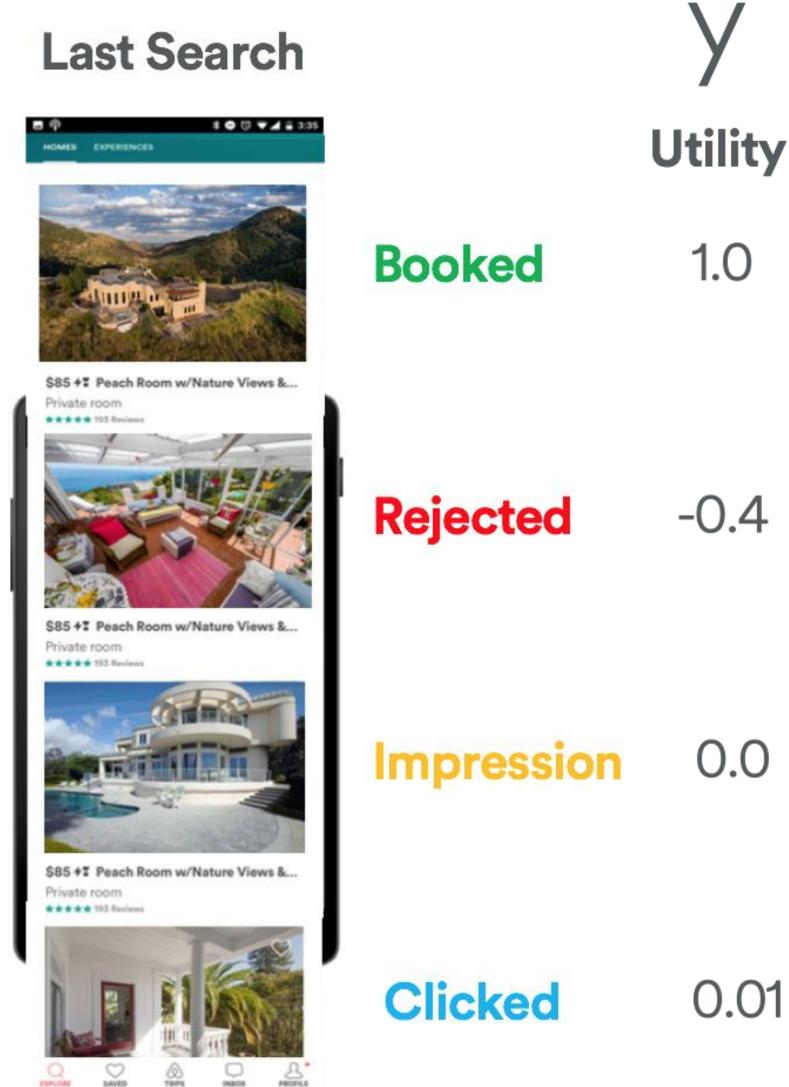
Model v0:

Last Search (label propagation) + Utility Prediction









Model v0: Last Search + Utility Prediction

- Regression: f(x) -> y Display Results in Descending Order of Predicted Utility
- Features (X):
 - Listing Features (engagement, quality, type, capacity, etc.)
 - Query Features (destination, nights, guests, etc.)
 - Guest Features (origin, past bookings, listing preferences, short term history)
 - Rejection Probability: output of separate model that predicts Host rejections

Model (f): GBM model



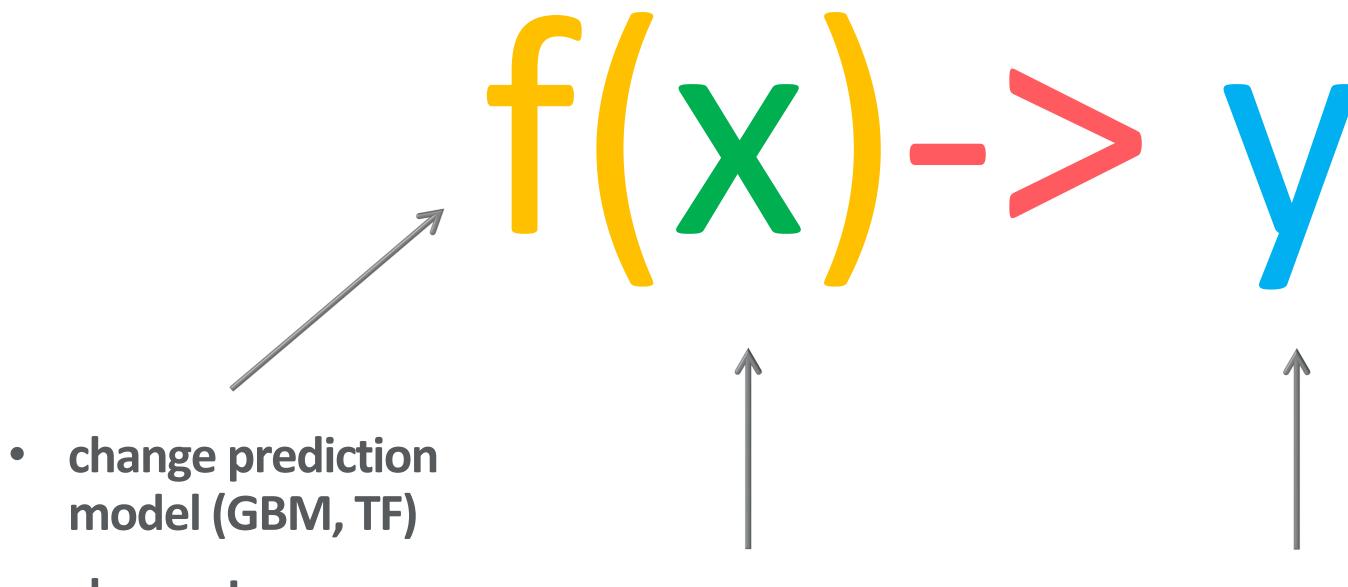
Search @ Airbnb

Airbnb 🖾

11 months ago · 4,834 views

"Search Ranking: GBM & personalization Search is a universal problem among almost all internet companies. But Airbnb's search ...

How we iterate?



change Loss function

 change how we create training data

add new features

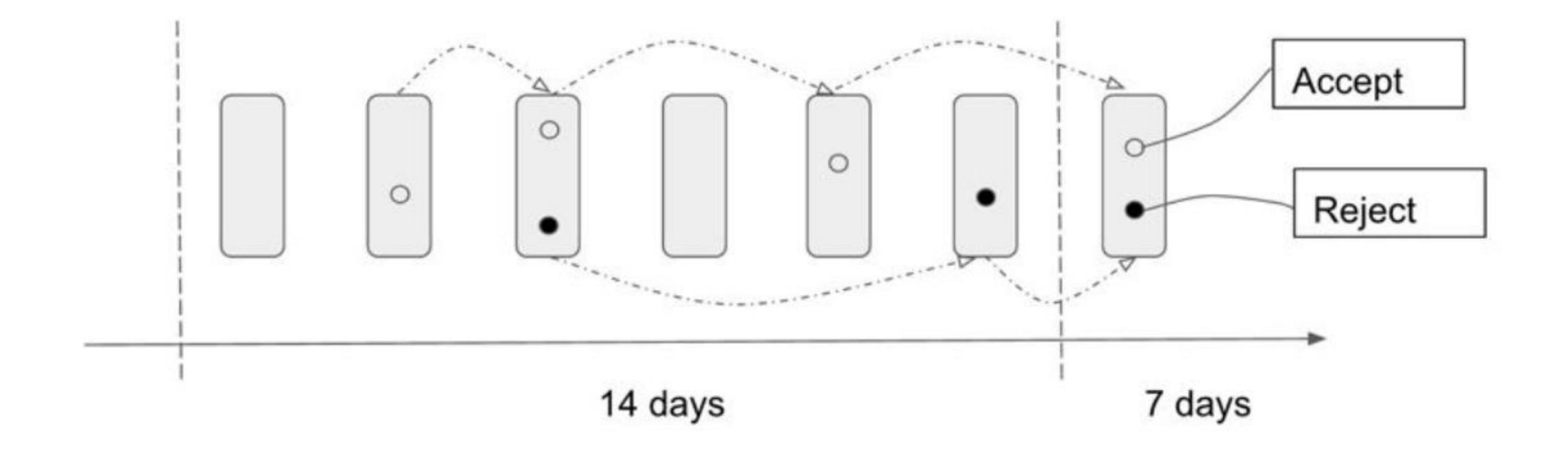
change what we predict

Examples of Successful Iterations

Significant Booking Gains

1. Early Attribution

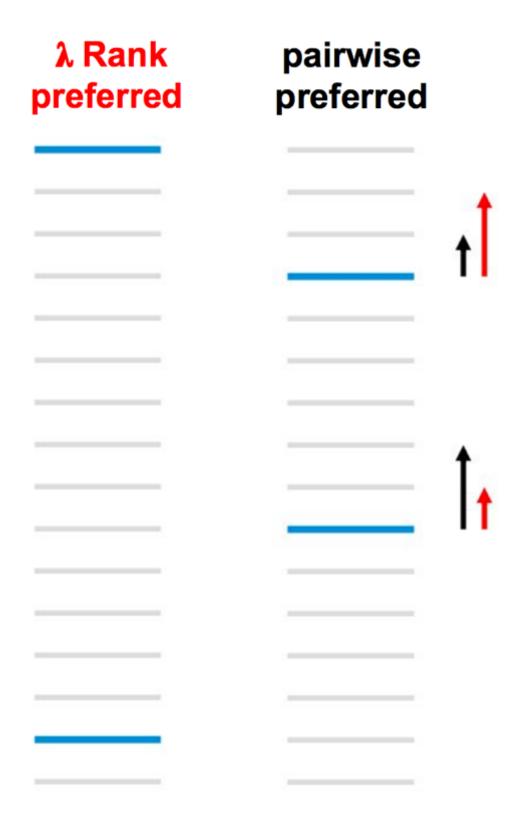
- Instead of only Last Search use Earlier Searches as well
- Propagate labels back



Examples of Successful Iterations

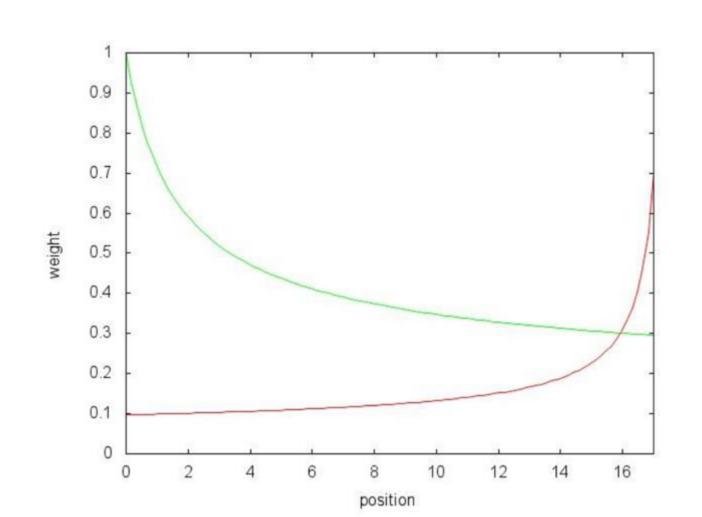
2. Pointwise to Pairwise

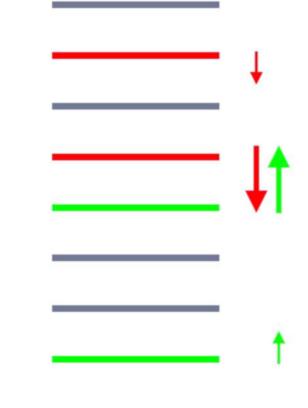
Weighted Pairwise (LambdaRank)



Weight Top Positions
More in Pairwise
Loss

- Dual Discount Curve Pairwise (Modified LambdaRank)
 - Lambda Rank focus only on pushing good outcomes to the top
 - But its equally important to push bad outcomes to the bottom





Green: Good Outcomes Red: Bad Outcomes

Examples of Successful Iterations

3. Real-time Personalization

Main Challenges

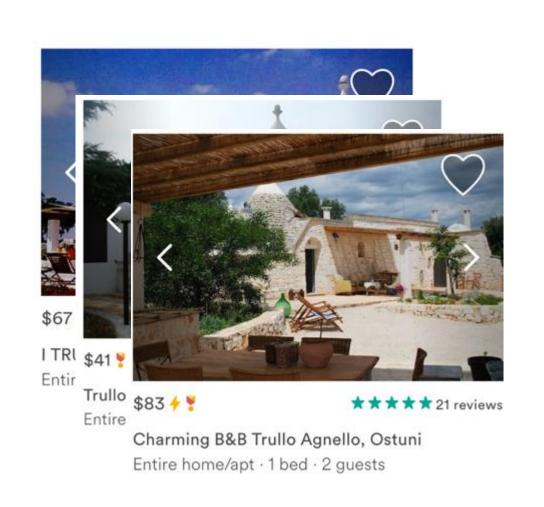
- Guests do not visit every day (most travel up to 2-3 times a year)
- Preferences may change from visit to visit
- Rarely go to same location twice

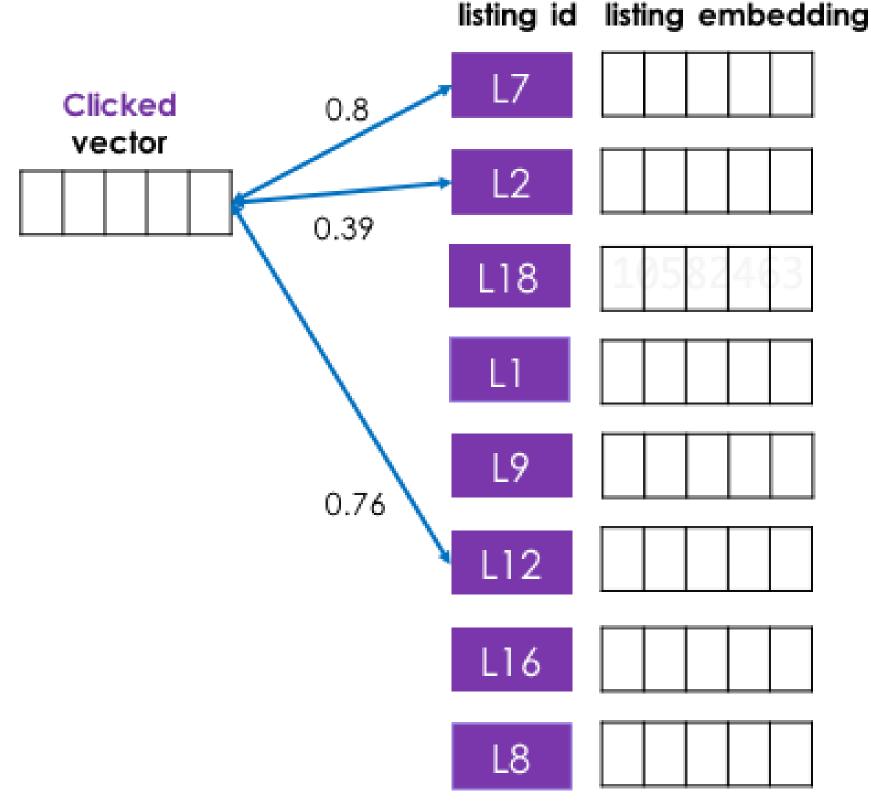
Approach: In-session personalization

- 1. Memorize listings you clicked on recently
- 2. Show similar listings to the one you liked (wishlisted, long clicked, contacted)

Real-time Personalization using Embeddings

recently long-clicked listings by user (>60 sec)





CANDIDATE LISTINGS

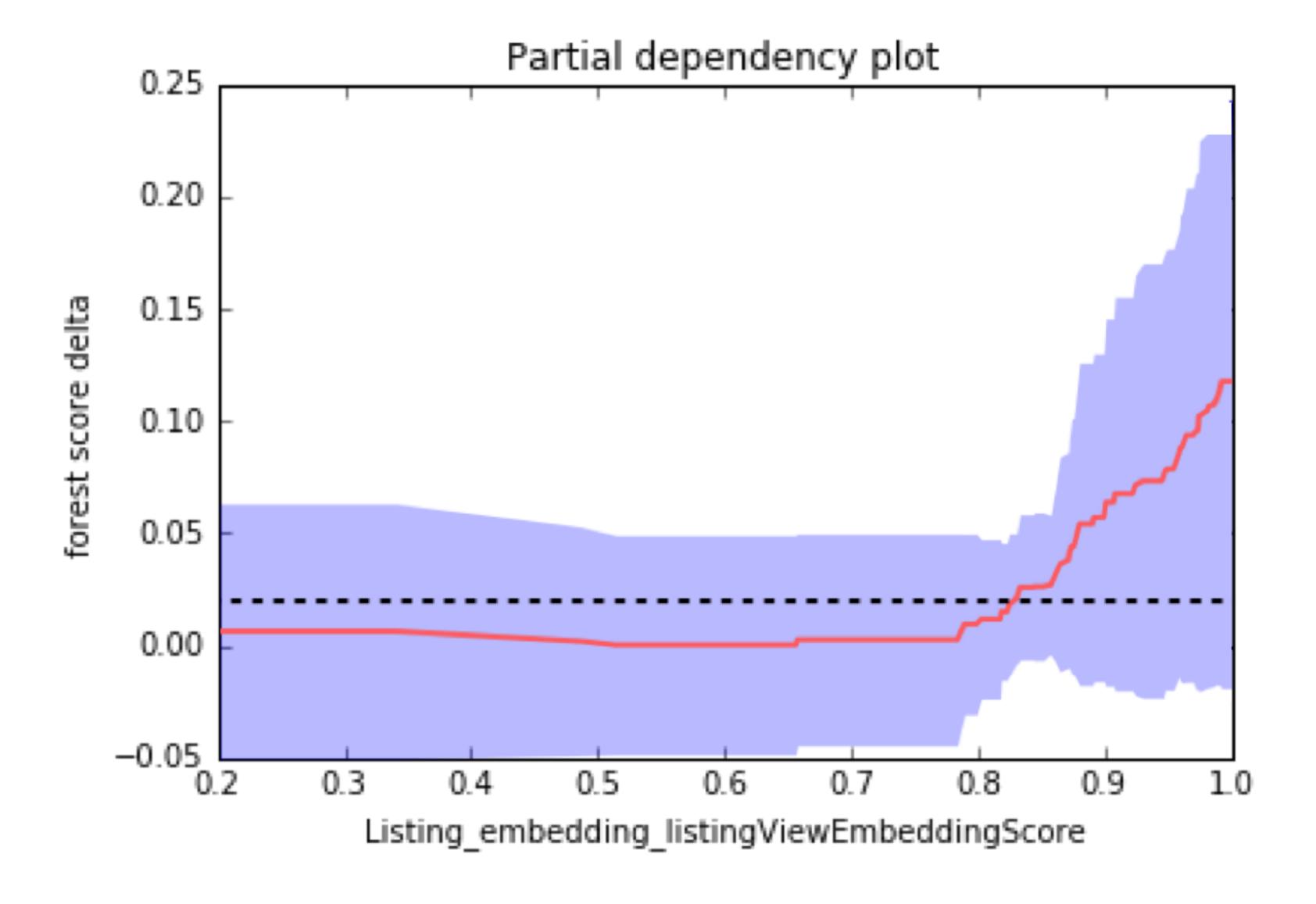
Embedding Features:

- 1. longViewEmbeddingScore
- 2. wishlistEmbeddingScore
- 3. contactEmbeddingScore

Real-time Personalization using Embeddings

Partial dependency plot

Feature Importance: top 10

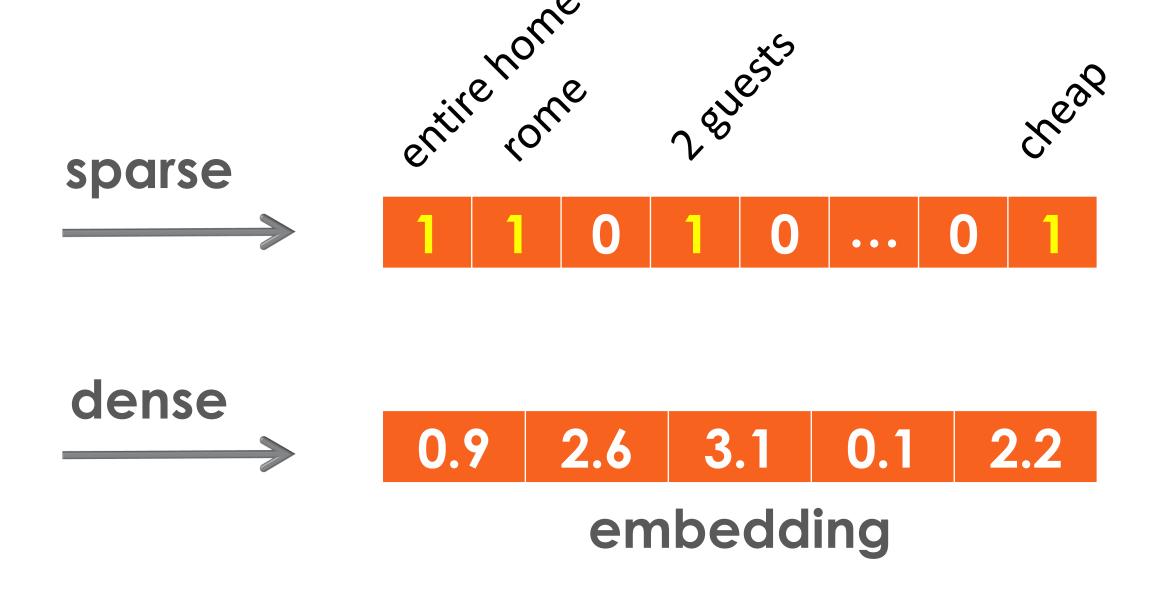




Listing Embeddings

How to represent a listing?

Listing
Casa Vacanze in Rome



- ☐ Represent listings as numeric vectors
- Vectors need to be learned using training data (search sessions)
- We want listings with similar contexts to have similar vectors
- □ Popular tool for training: word2vec

Search Sessions

Actions by single user (listing clicks, inquiries, bookings) ordered in time

```
748612 4160766
           2582727 5473823 2582727
                                     5473823
   6251934 9257649
   7924193 10116733 8998529 9075420 4113166
           4986017 5503478 5503478 5879904
   5503478
S6 10237904 8680483 8718513 11691507 4831342 8004575 7866901
```

Search Sessions

We only use clicks that have >30 sec page view time (no accidental clicks)



Session ends when there is more than 30 min inactivity



Training

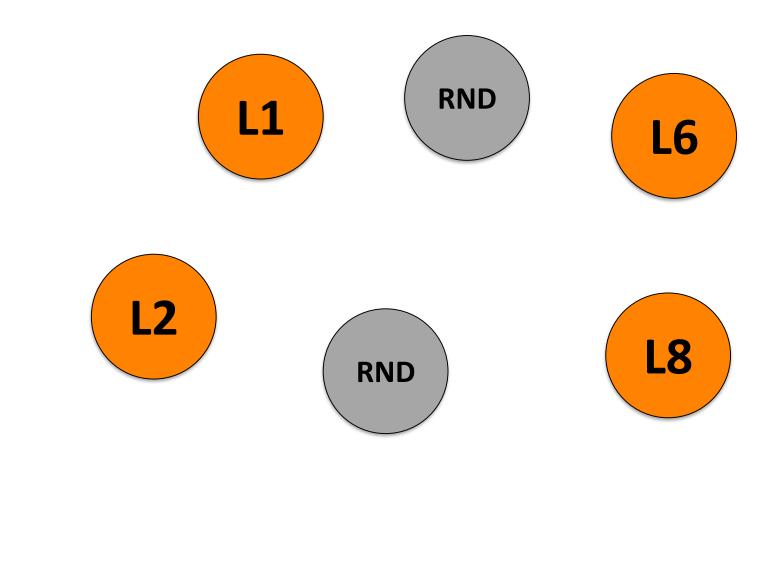
Example search session:

Listing8, Listing1, Listing2, Listing6



$$\mathbf{v}_{i}^{new} = \mathbf{v}_{i} + h \times (1 - S(\mathbf{v}_{i}^{T}\mathbf{u}_{neigh})) \times \mathbf{u}_{neigh}$$

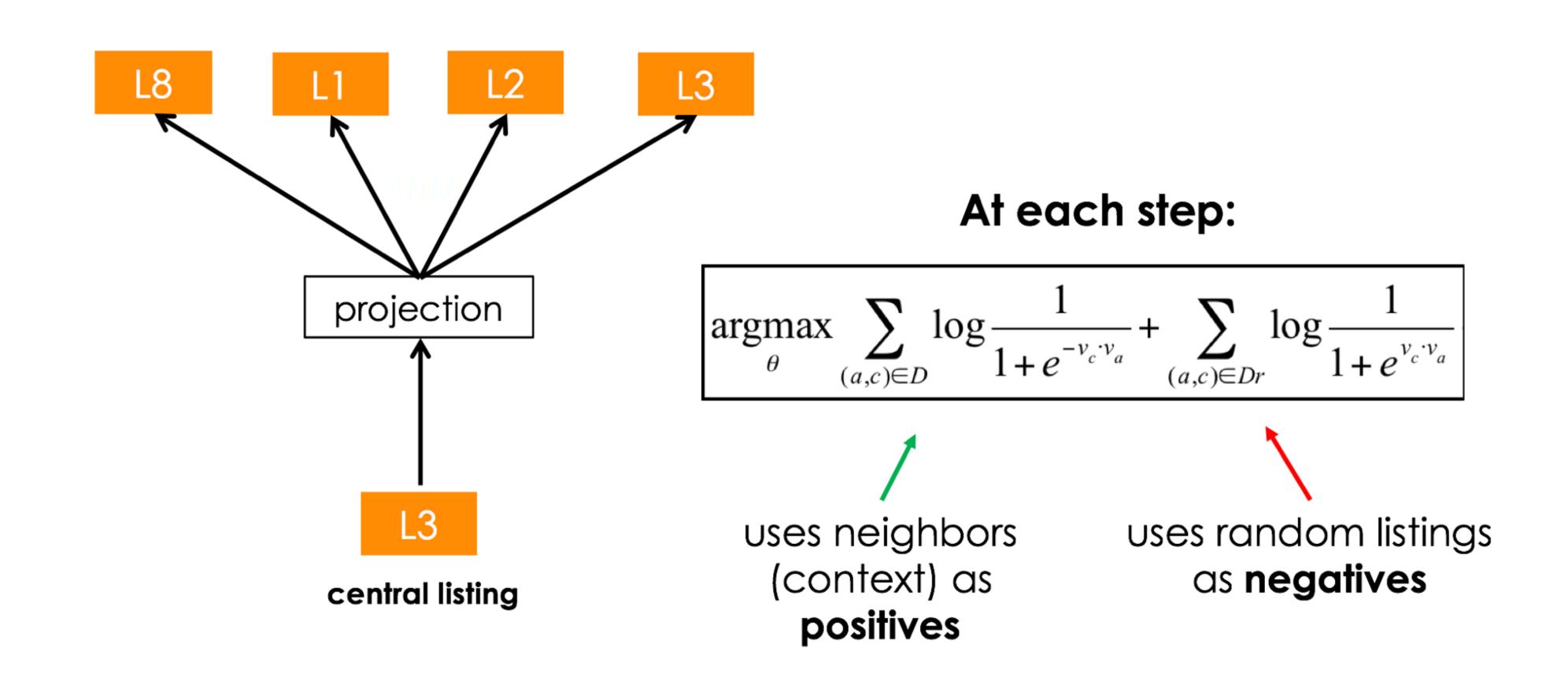
$$\mathbf{v}_{i}^{new} = \mathbf{v}_{i} - h \times S(\mathbf{v}_{i}^{T} \mathbf{u}_{neg}) \times \mathbf{u}_{neg}$$



embedding space

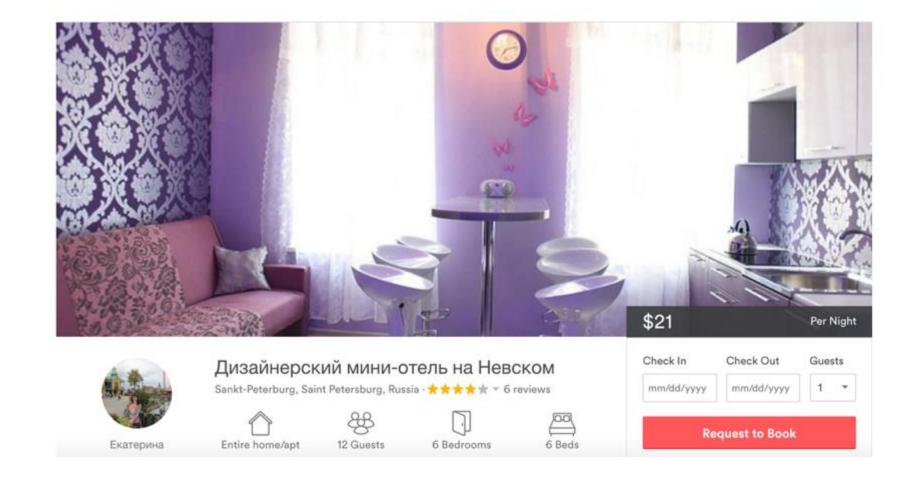
Training

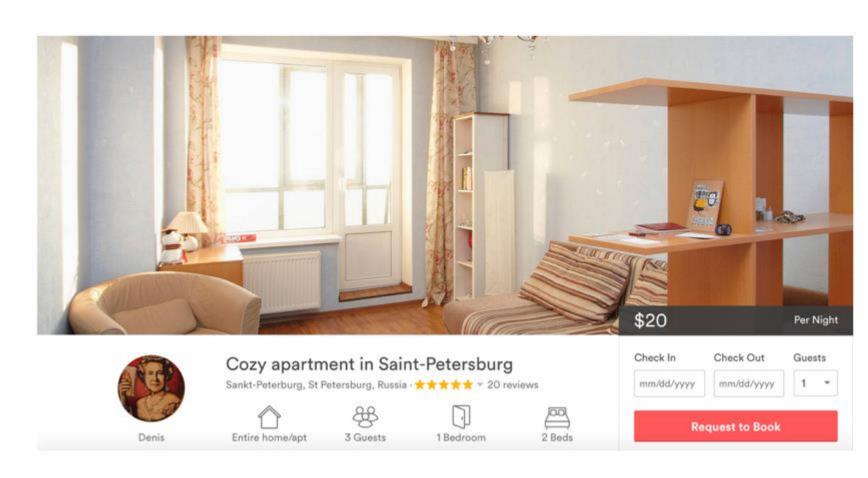
Skip-gram model (word2vec)

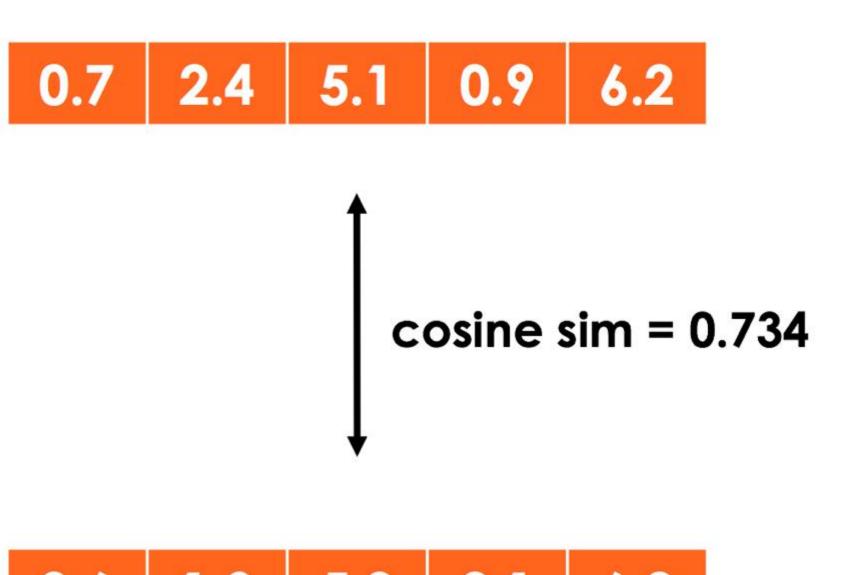


After Training

Can calculate similarity between any two listings

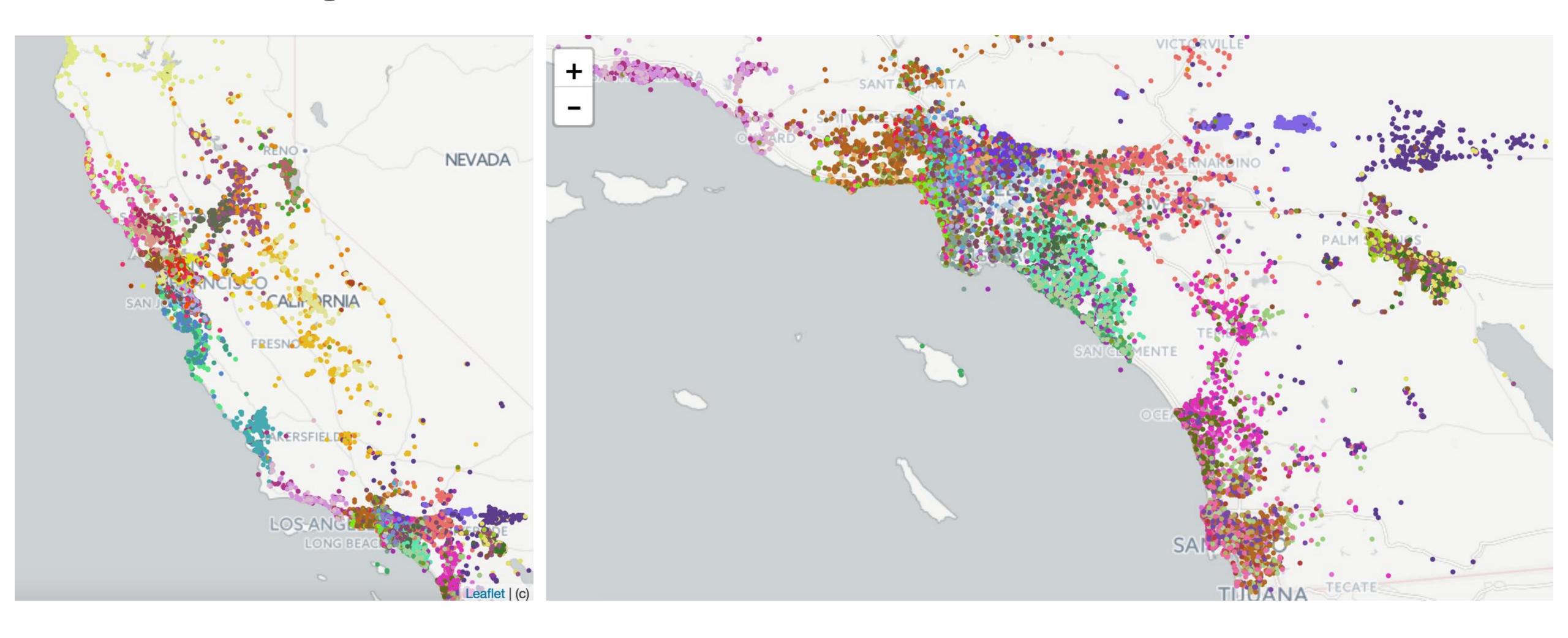






After Training

Can cluster listings: California



word2vec Modifications that Worked for us

1. Not all interactions between two listings are the same

co-click session: 5473823 2582727 5473823 2582727 5473823 5473823

co-inquiry session: 6251934 9257649

sessions that end in booking: 5473823 2582727 5473823 2582727 5473823 5473823

Positive interactions

- co-click default weight
- co-inquiry larger weight
- booking sessions larger weight

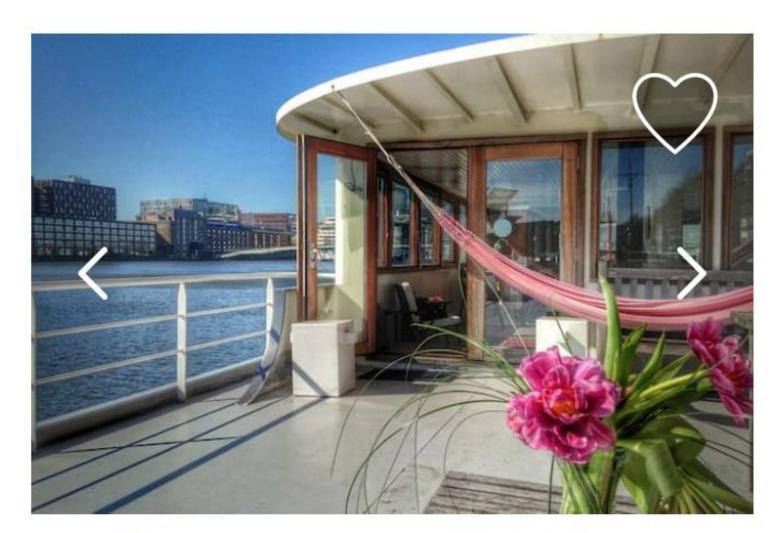
word2vec Modifications that Worked for us

2. Better negative sampling

- Random negatives: probability of sampling a negative from same city very low (not a problem if you have 1-2 cities but we have thousands)
- Random negatives from same city: leads to better within city listing similarities
- Analogous to learning word vectors when data comes from many languages

Applications

Can Find Similar Listings



\$156 🦞

★★★★★ 237 reviews

LUXURY INDEPENDENT STUDIO on SHIP : fr...

Private room · 2 beds · 3 guests

Listing	Cosine distance
579851	0.989346
851044	0.982895
64736	0.979351
2024269	0.976072
802052	0.974991
884329	0.974928
3850766	0.972117
5207034	0.968444
2126459	0.963931
1025482	0.963376



43 ★ ★ ★ ★ 249 reviews

Studio on Spectacular Houseboat Private room · 1 bed · 2 guests



\$133 ***** ★ ★ ★ ★ 318 reviews

HOUSEBOAT NOVA 80m2 + FREE BIKES
Private room · 3 beds · 4 guests



\$100 ₹ ★★★ 263 reviews

Sleeping on a ship in the centre.

Private room · 1 bed · 2 guests



\$98 **→** ★★★★ 306 reviews

Houseboat, garden, center, free parking Private room · 1 bed · 2 guests



\$138 **→ ★ ★ ★ ★ 2**45 reviews

Authentic houseboat with privacy and comfort Private room · 1 bed · 2 guests



\$161 ₹ ★★★ 71 reviews

Luxury Houseboat

Entire home/apt · 1 bed · 2 guests

Applications

Similar Listings



\$66 🖁

★★★★★ 71 reviews

I SETTE CONI - TRULLO EDERA

Entire home/apt · 2 beds · 4 guests · Business Travel Ready

Similar listings



\$55 TRADITIONAL HOUSE IN OSTUNI!!!
Entire house · 1 bed

★★★★ 55



\$57 PIETRACHIARAHOUSE PANORAMIC TE...
Entire house · 1 bed

★★★★★53



\$67 REhoUSE Ostuni
Entire apartment · 2 beds

★★★★★ 56

Before

Similar Listings

\$99 4



Trullo of 1800 in the Itria Valley
Entire home/apt · 5 beds · 5 guests





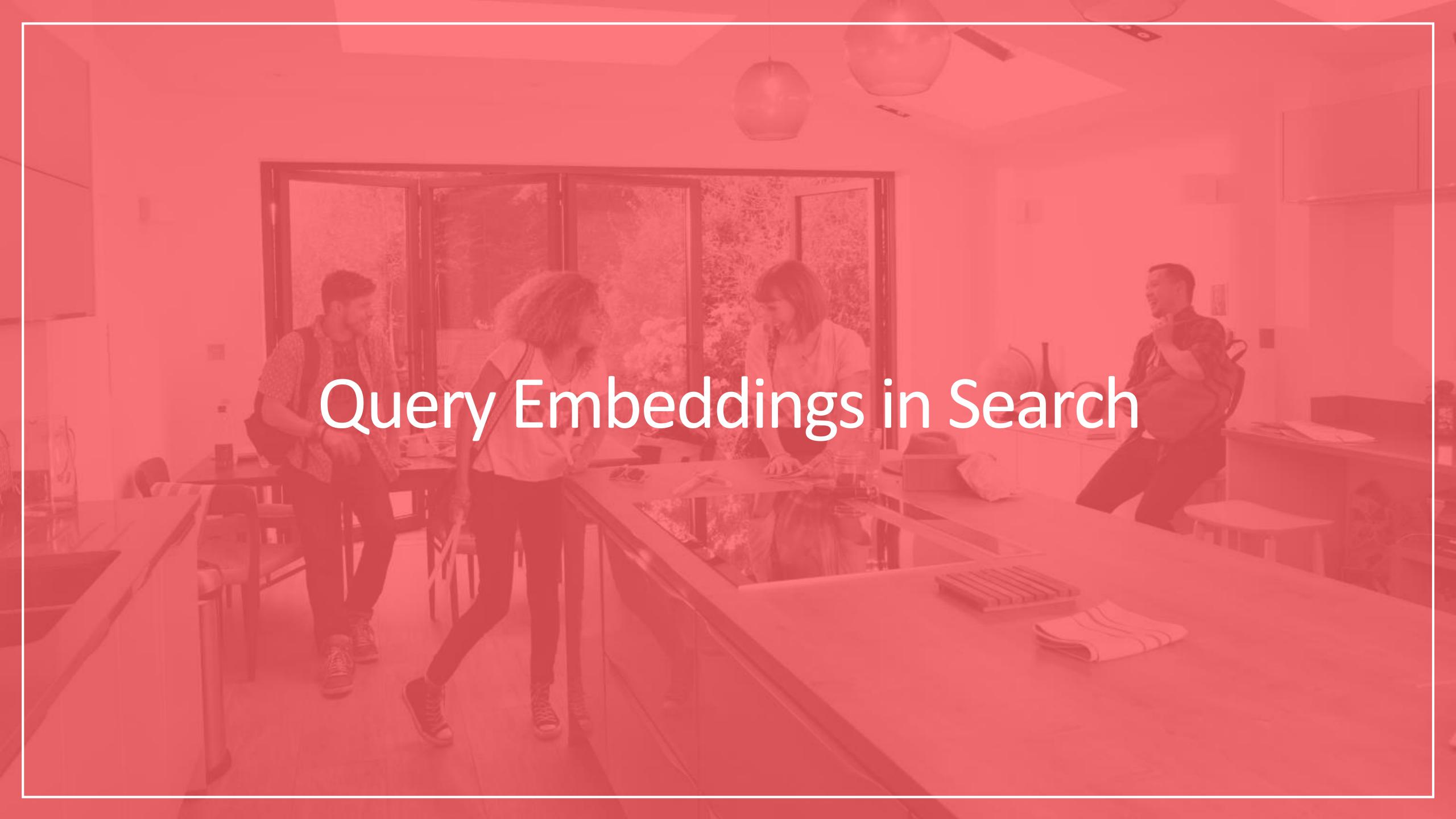


I TRULLINI OSTUNI - MARTINA FRANCA Entire home/apt · 1 bed · 2 guests

*** 30 reviews

\$67 4 🖁

After



Query Embeddings

Query (geo_id available)
San Francisco, California, USA

embedding:

embedding:

embedding:

embedding:

O.2 1.1 7.2 0.8 3.1

embedding:

O.7 3.4 2.8 0.1 2.2

- Represent queries as numeric vectors
- Vectors need to be learned using training data (search sessions)
- □ We want queries with similar contexts to have similar vector

Search Sessions

Actions by single user (queries) ordered in time

```
Garmisch-Partenkirchen
```

- S1 skiing_germany geo_fs2342f61lHgQ
- S2 winter_wanderland winter_wonderland winter_wonderland_neitherlands skiing geo_f23432dfHf

sa portugal_surfing portugal_surf geo_MgsRQfBOHA geo_Hw0RU41424Da

Peniche, Portugal

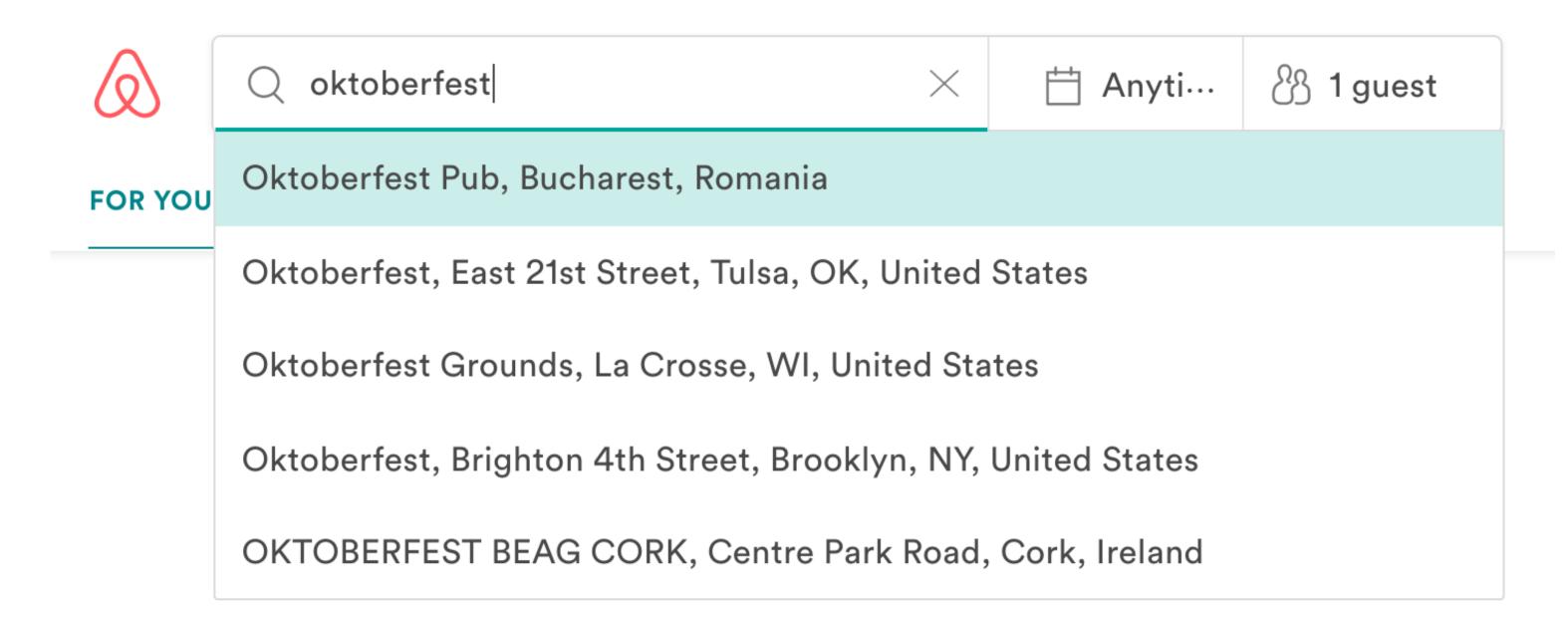
- 54 grant_park_chicago lollapalooza grant_park_chicago geo_Llq58sDogRHf1
- geo_Kyvb07vZg french_quarter_mardi_gras geo_ChIJf5407f1

After Training

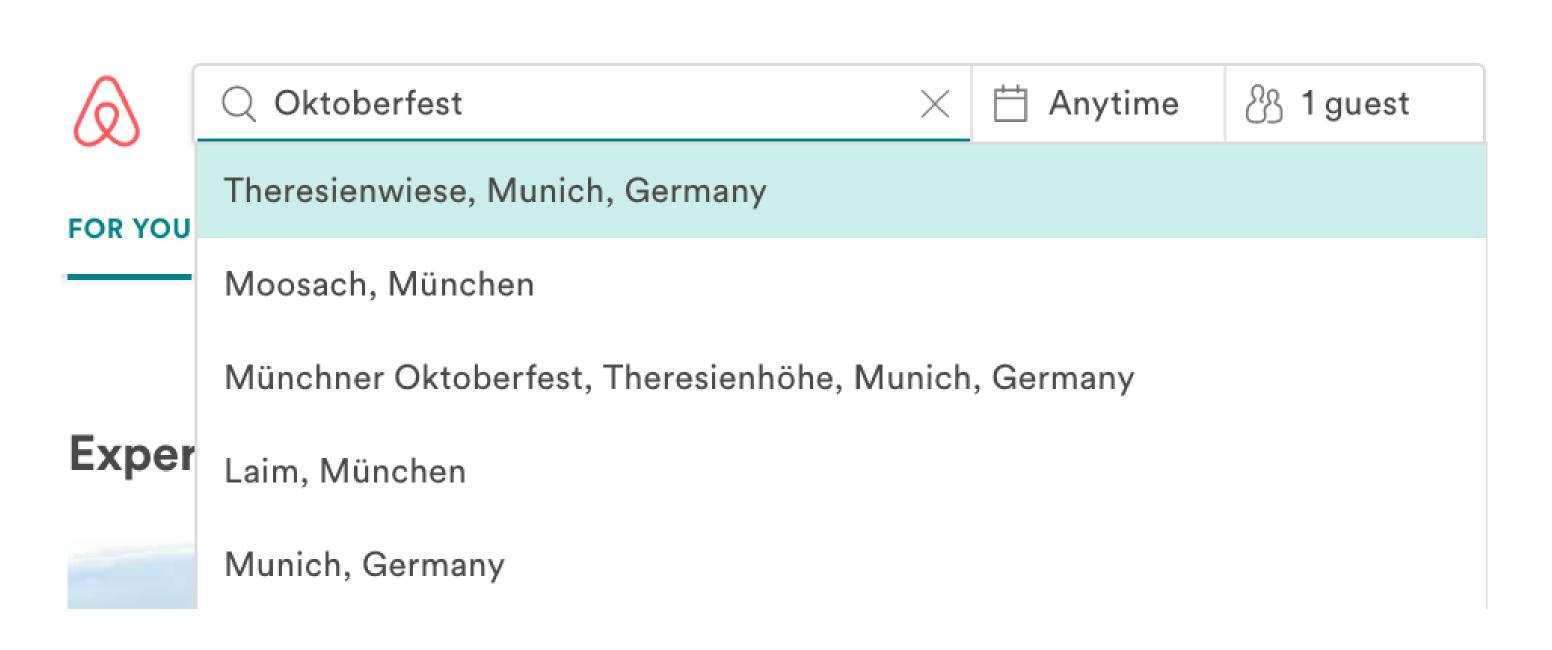
We can calculate similarities between queries

Application: Autocomplete (Type-ahead)

Before



After







Greek Islands



Ħ Anyti...

🖄 1 guest

FOR YOU

Greek Islands, South Halsted Street, Chicago, IL, United States

Greek Islands Taverna, North Ocean Boulevard, Fort Lauderdale, FL, United States

Greek Islands, East 22nd Street, Lombard, IL, United States

Greek Islands, Center Street, Omaha, NE, United States

Greek Islands, East Andreas Road, Palm Springs, CA, United States

After



Greek Islands



🛱 Anyti...



🖄 1 guest

Santorini, Thira, Greece

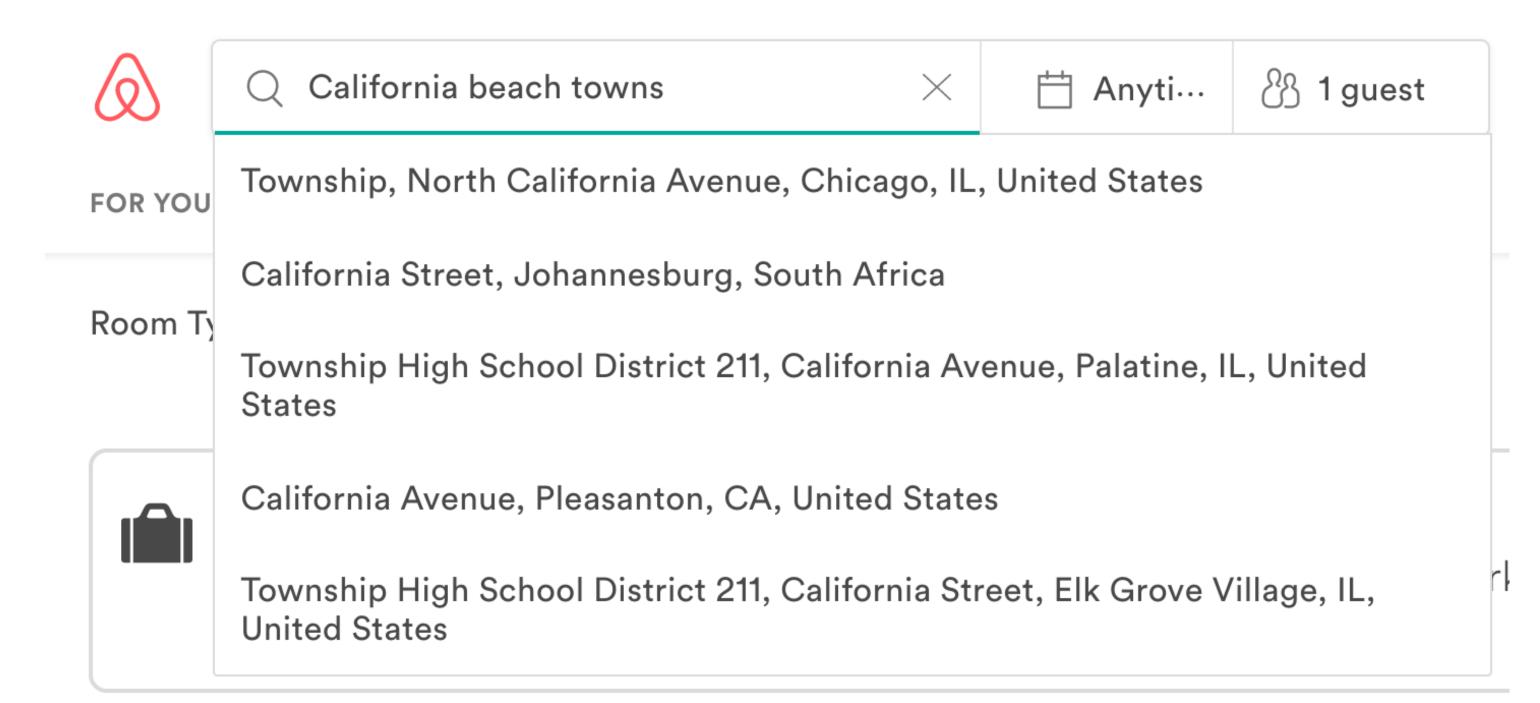
Mykonos, Greece

Ionian Islands, Greece

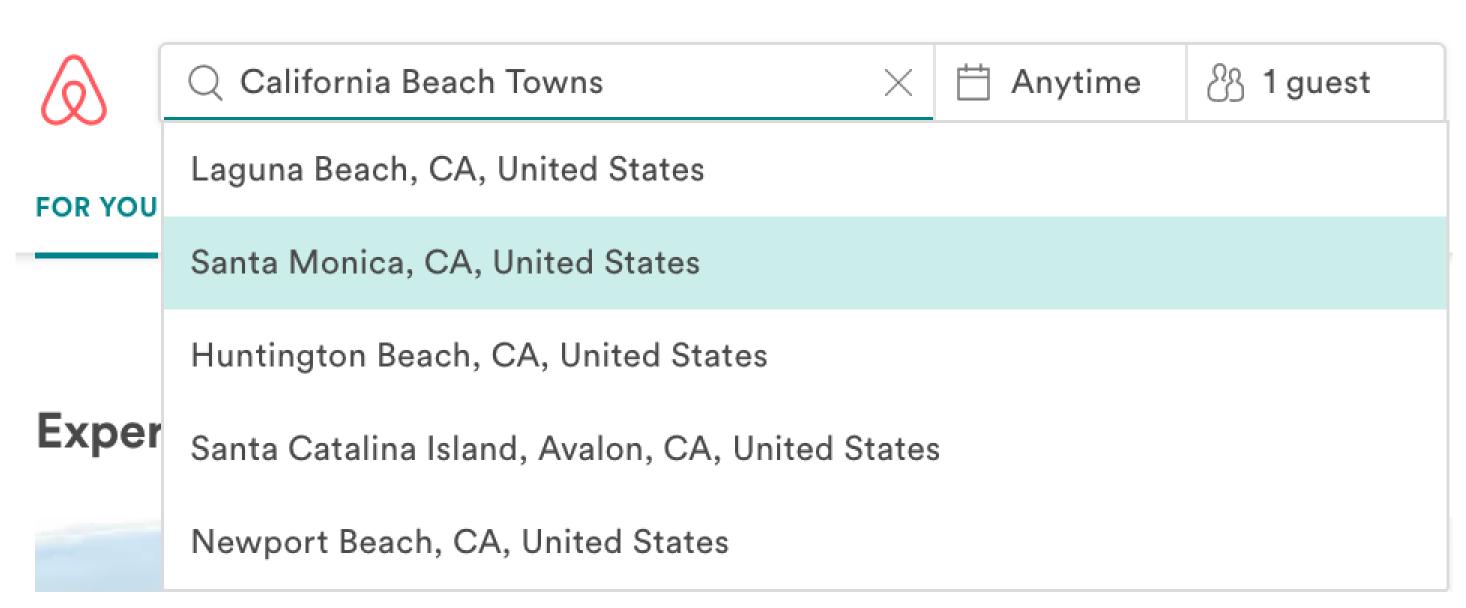
Lesvos, Greece

Cephalonia, Greece

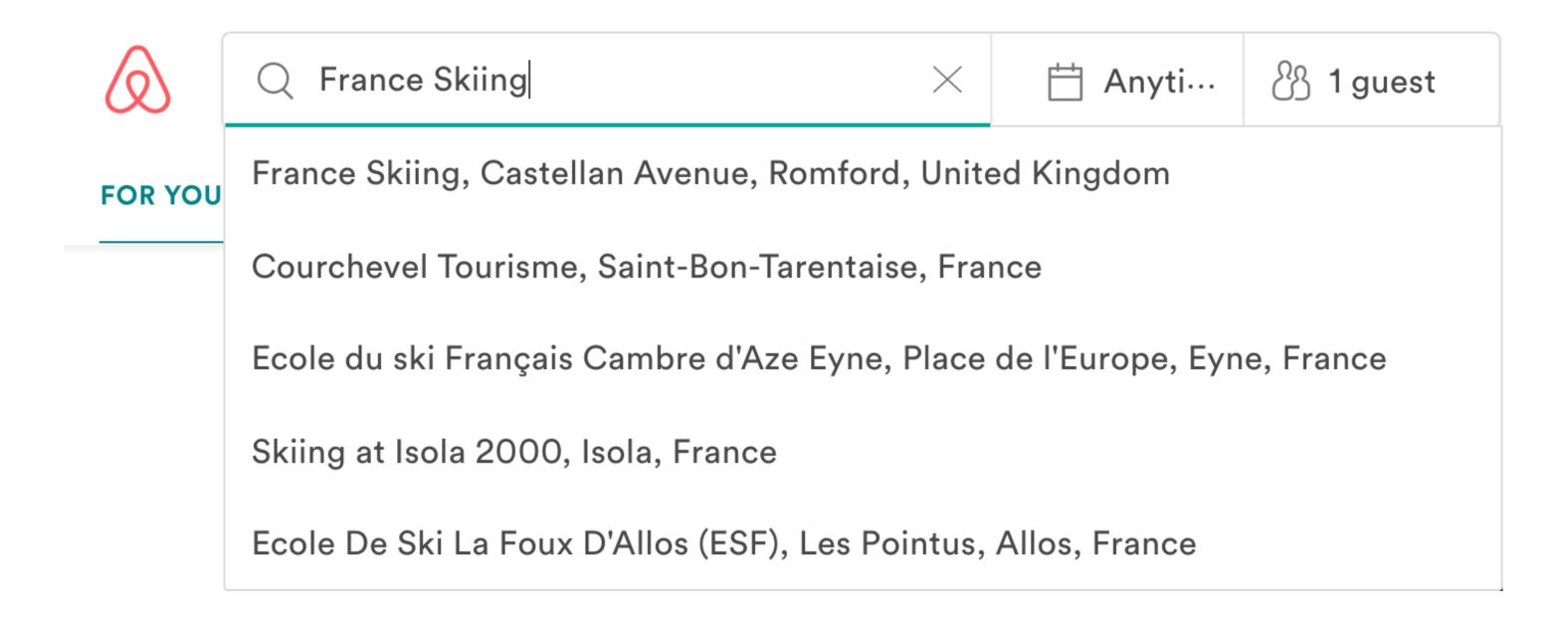
Before



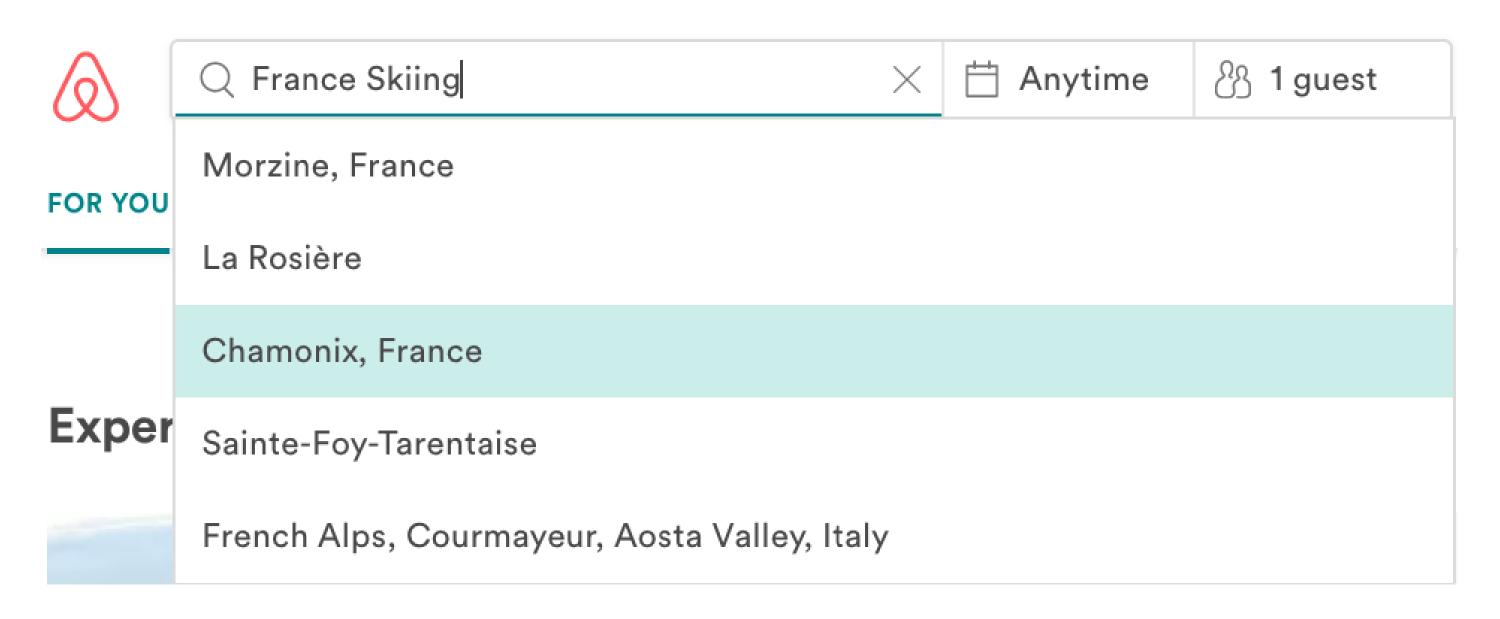
After



Before



After



Questions?

Press questions:

press@airbnb.com

We are hiring ML Scientists

Contact me:

mihajlo.grbovic@airbnb.com

