Julia Kiseleva

Eindhoven University of Technology TU/e

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Outline

- Research Problem and Questions
- What is Contextual Information?
- Searching and Browsing Behavior
- Training Context-Aware System
- Applications that Benefit from Contextual Information
- Conclusion & Open Questions

bing

questions

MS Beta questions

guestions to ask at an interview

questions to ask your boyfriend

questions to ask a girl

questions to ask in an interview

questions to ask a guy

questions and answers

questions aapke answers sai baba ke

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Main Research Problem

Great imbalance between richness of information on the web and the succinctness and poverty of search requests of web users

Queries are only a partial description of the underlying complex information needs

How to **discover**, **model and use** contextual information in order to understand and improve users' searching and browsing behavior on web?

Understanding user needs



Let's give it a try...





Any time Latest Past 24 hours Past 2 days Past week Past month Past year Custom range... my keys

About 129,000,000 results (0.17 seconds)

Where Are My Keys? I Solution

Welcome to "Where are my keys?" the only site vo are currently 20378 places where what was lost cou www.them.ws/keys/index.php - Cached - Similar

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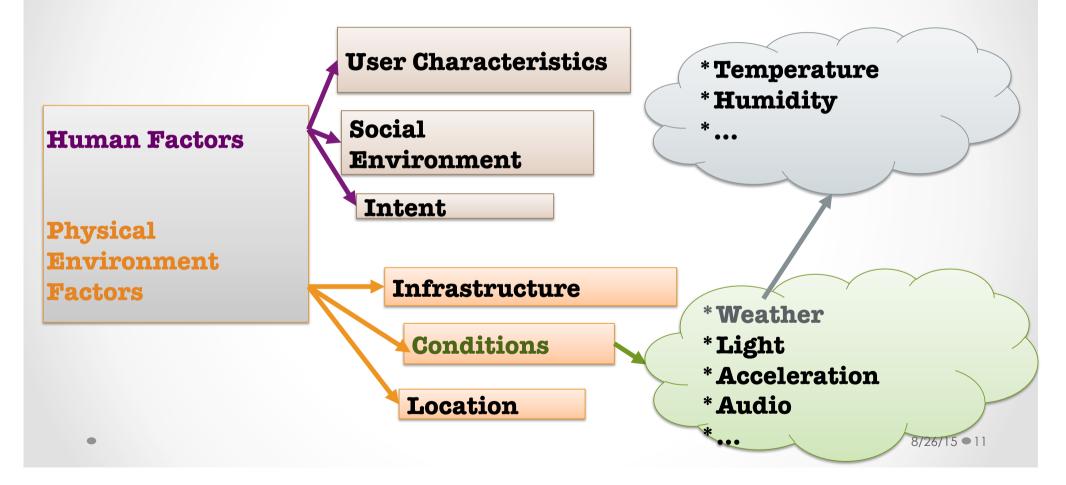
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History of context definition and discovery

Context	Year
Location	1992
Taxonomy of explicit context	1999
Predictive features vs. contextual	2002
Hidden context: (clustering, mixture models)	2004
Contextual bandits	2007
History of previous interaction	2008
Independence of predicted class	2011
Two level prediction model	2012
Focus on Context Discovery	2012 -

Timeline

Explicit Context Taxonomy



Type of Context

- User Context
 - o User Preferences
 - o User profiles
 - Usage of user history

Document/Product Context

- o Meta-data
- Content features

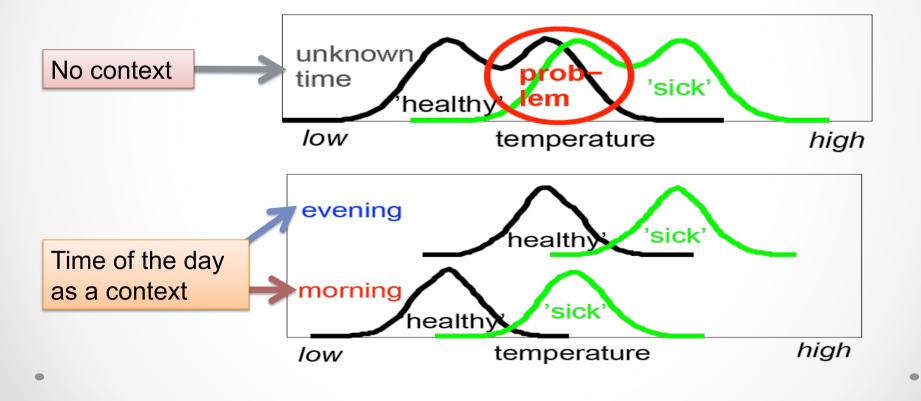
Task Context

- Current activity
- Location and etc.
- Social Context

Leveraging the social graph

Example of Context: in Diagnostics

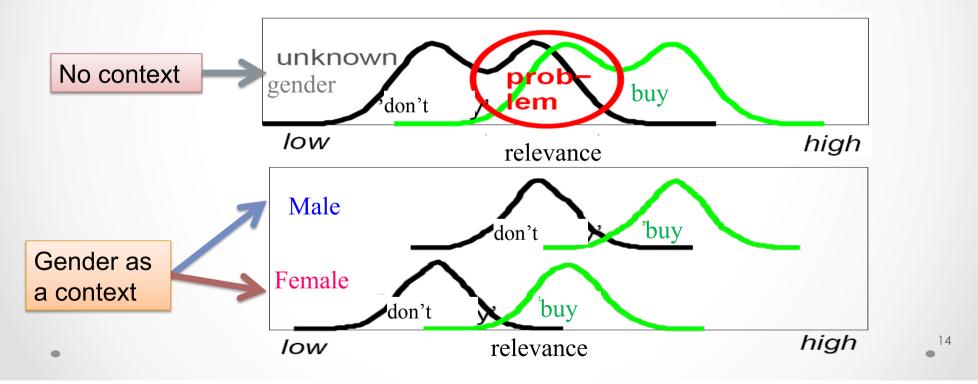
 Not predictive alone but a subset of features with the contextual attribute(s) becomes (much) more predictable



Example of Context: in Marketing

P(Purchase|gender="male")=P(Purchase|gender="female")

Model_{Male}~f(relevance); Model_{Female}~f(perceived value);



Types of User Behavior

- Searching when users are issuing queries (users have particular information needs):
 - We are trying to improve search results (SERP) taking context into account
- Browsing when users are surfing a website:
 o we are analyzing their movements utilizing context

Contextual Information affects user behavior!

Web Predictive Analytics

- Web predictive analytics aims to predict individual and aggregated characteristics indicating visitor behavior for purposes of understanding and optimizing web usage
- Application
 - Search engines (Bing, Yandex)
 - Recommender System (Booking.com, Yahoo News)
 - Computational Advertisement (Amazon.com)
- Tasks
 - Online shop's recommendations
 - Users' next action prediction
 - Users' intention predicting
 - Personalized search result page
 - 0 ...

Modeling User Behavior: What?

- Predicting Users Next Action
 - Click prediction on display advertising
 - Drop out prediction
 - o User Trail prediction
- Predicting Information Needs
 - Navigational vs. explorative vs. purchase
 - o Changes in user intent
 - Personalization based on context
 - Personalization based on changed context

Modeling User Behavior: Why?

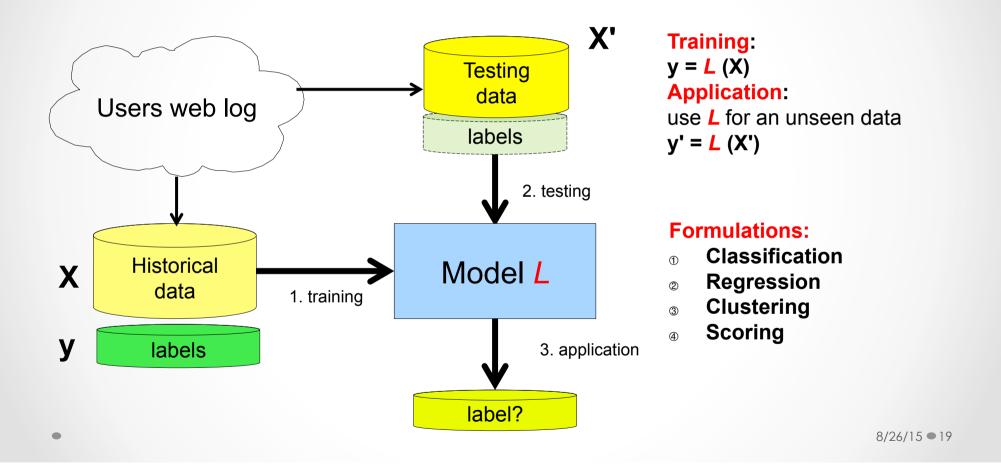
- Understanding How Satisfied are Users
 - o redesign website
 - o diversified search
 - o search recommendations

• To better use advertisement budget

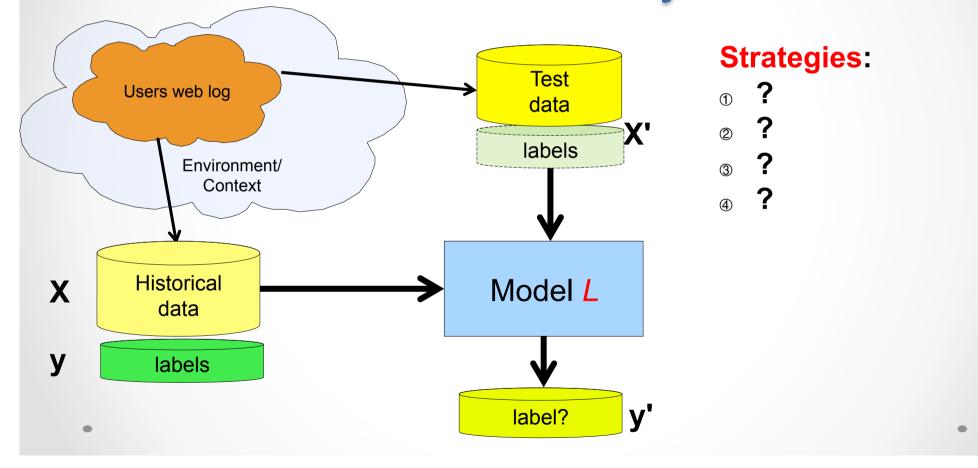
- o When serve ads?
- What type of ads to serve?
- brand awareness CPM or convergence CPC
- To `manipulate' user is it worth giving a promotion?

 personalize with intent of converging to a desired action
 personalized suggestions based on user context

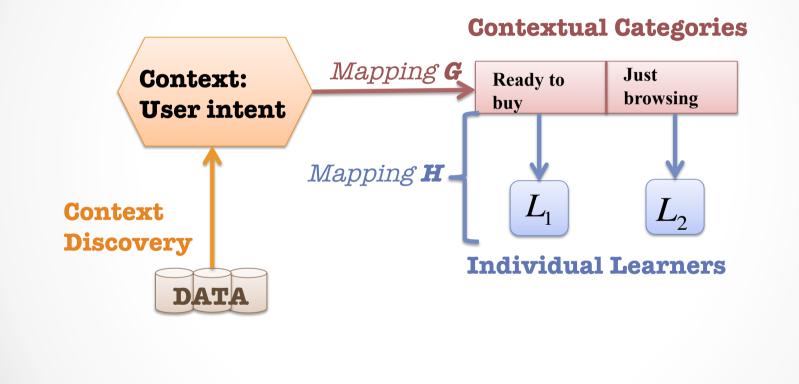
User Intent Modeling: How?



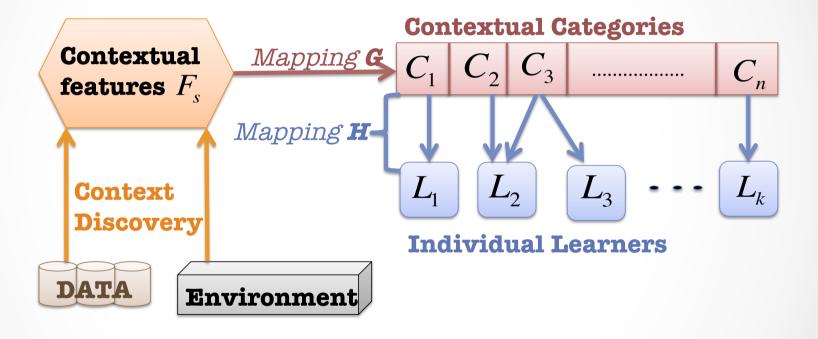
Context-Awareness in Web Predictive Analytics



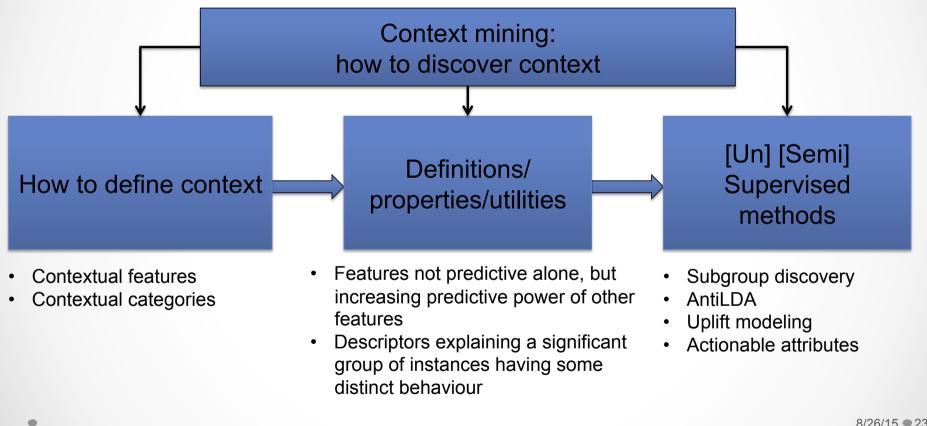
Defining Useful Context



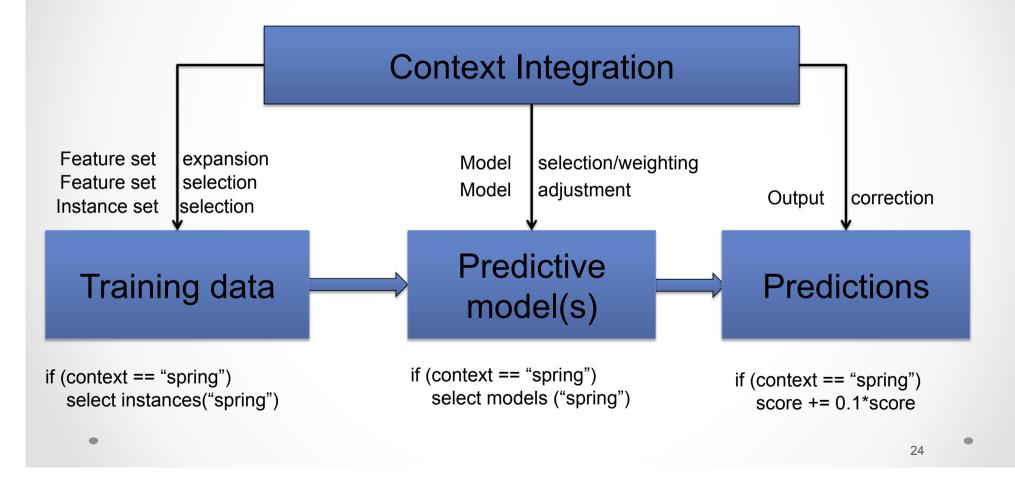
Defining Useful Context

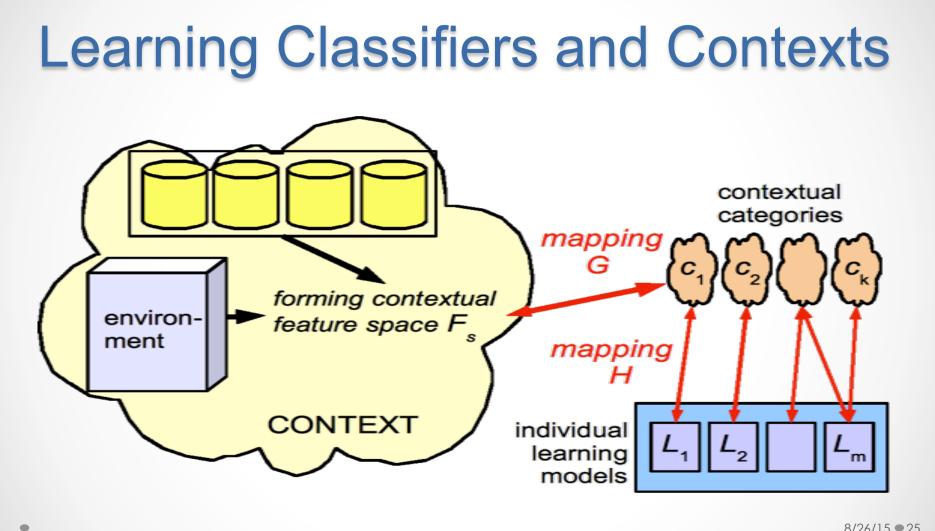


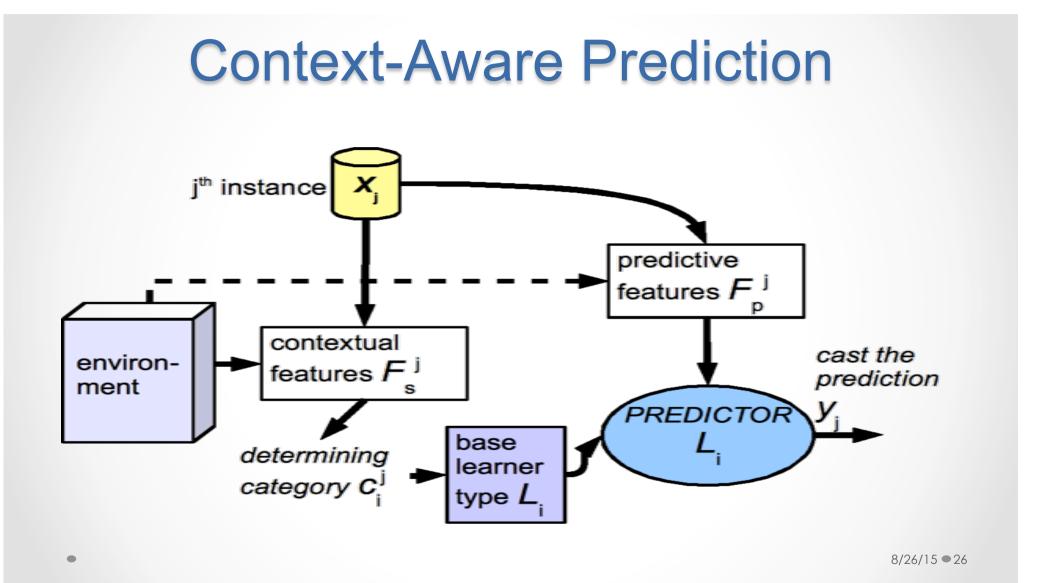
Strategies for Context Discovery

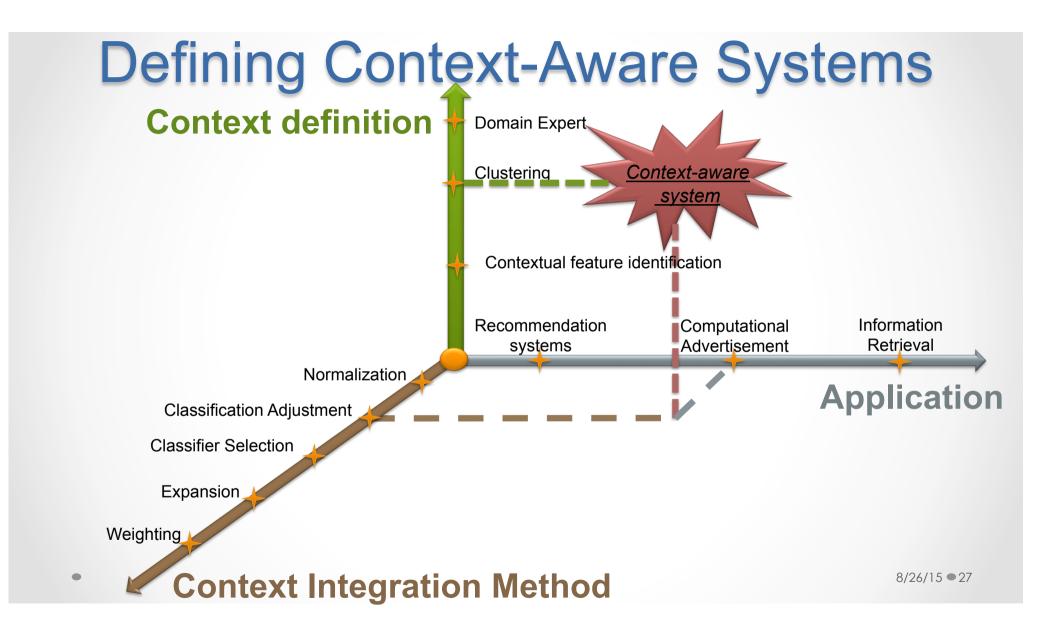


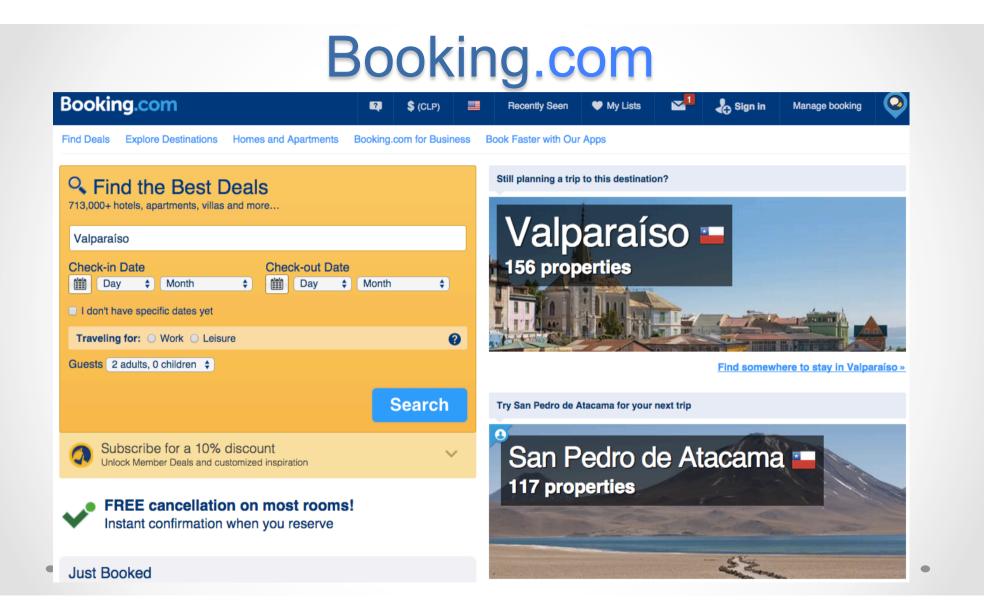
Strategies for Context Integration







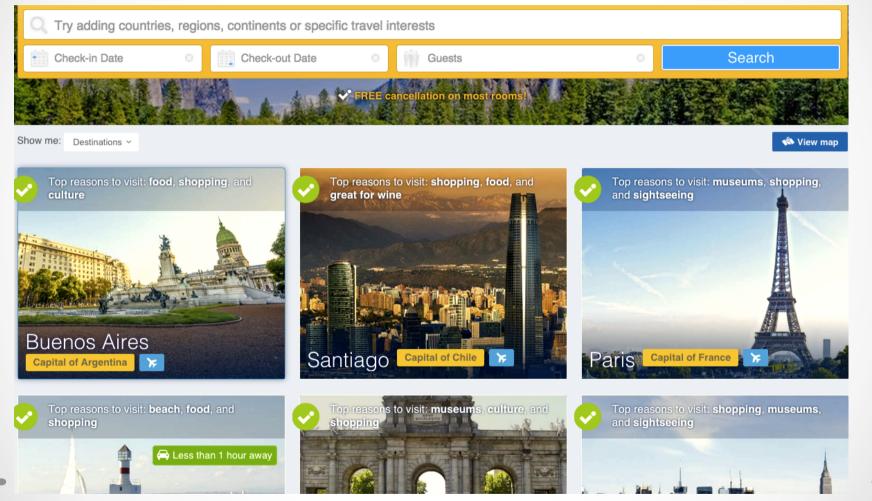




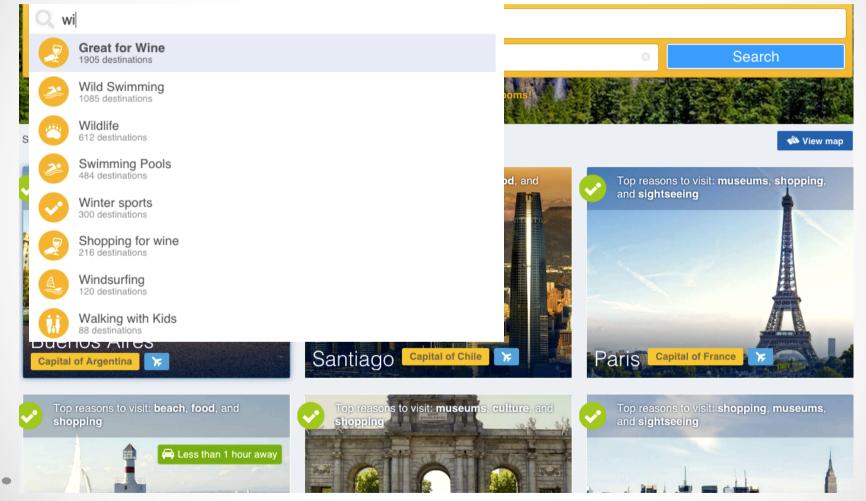
Booking.com

- World largest online travel agent
- > 220 countries
- > 81.000 destinations
- > 710.000 bookable hotels worldwide
- > 30.000.000 unique users
- >> 100.000 unique visitors per day

Destination Finder



Destination Finder



Destination Finder

Q Grea	t for Wine 🛛 Foo	od 🛛 Add more	Э			
Check-	in Date ©	Check-o	out Date 🛛 🛇	Guests	0	Search
		141 A		cancellation on most room	ısı	
Sort by: Releva	ance ~ Show me: [Sale of the second	2. 王东、梁东 法专		🐝 View map

Best places in the world for "Great for Wine" and "Food" - based on traveler recommendations



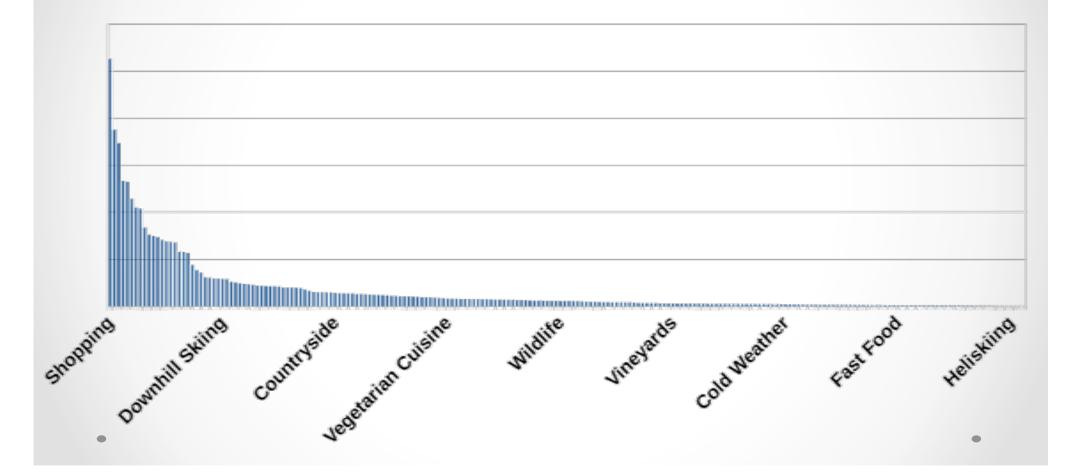
How Do We Get Endorsements?

- Only users who stayed at a hotel in a destination can endorse it
- Free text endorsements since 2013
- Since 2014 free text endorsements standardized to 256 canonical tags
- Used NLP techniques to extract the canonical base

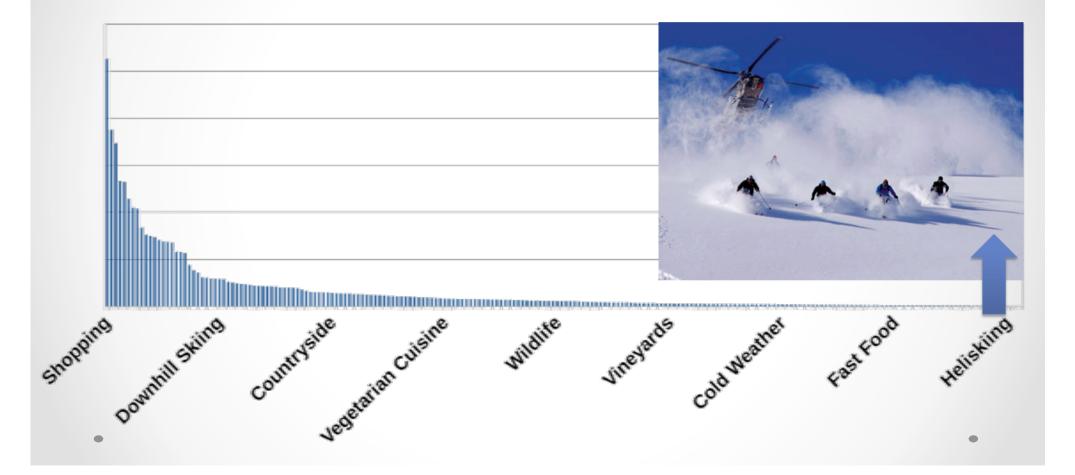
More Numbers:

- > 13.000.000 total unique endorsements
- > 60.000 destinations

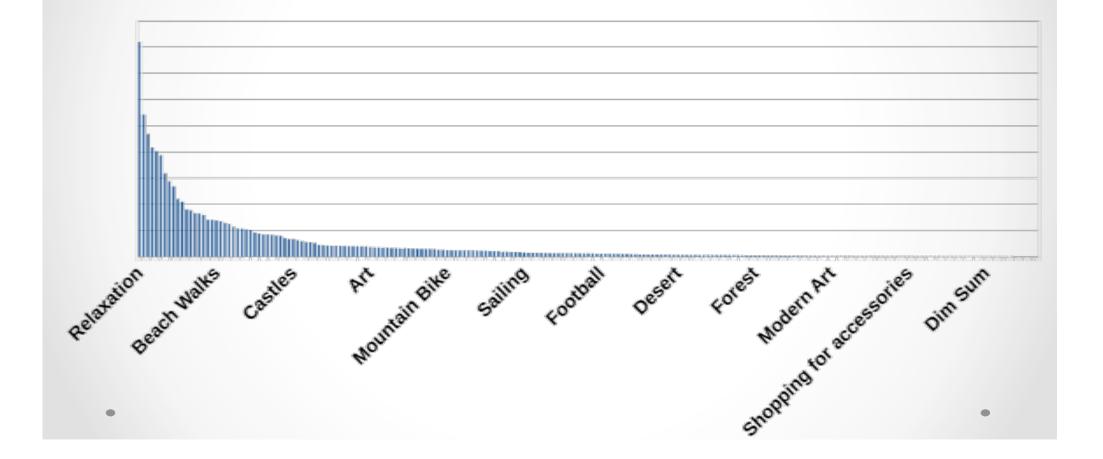
Endorsement Histogram



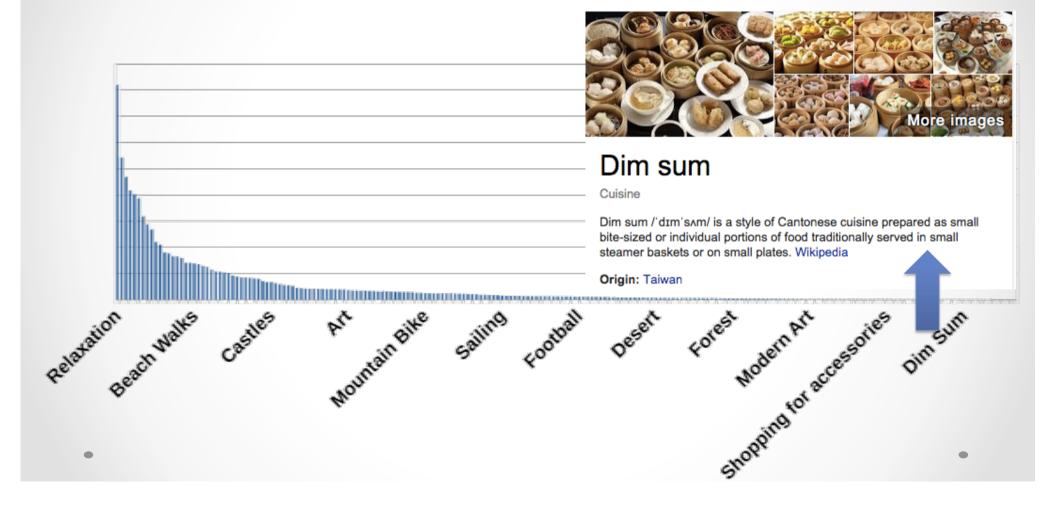
Endorsement Histogram



Endorsement-Destination Histogram

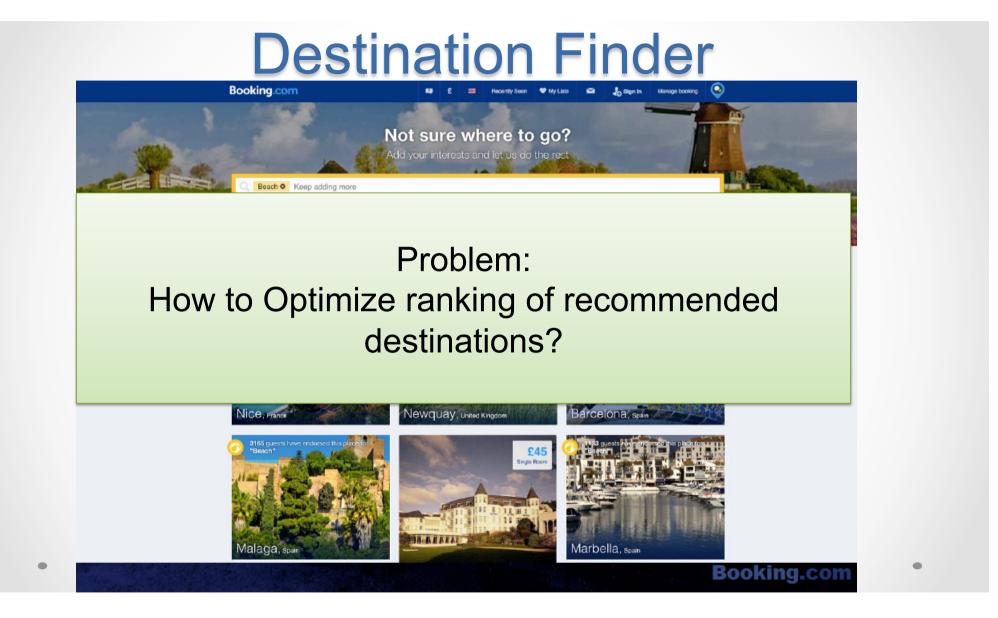


Endorsement-Destination Histogram



Endorsements for Santiago

	Popularity	Name
	1	Shopping
	2	Wine Travel
	3	Food
	4	Museums
TI.	5	Culture
	6	Gourmet Food
	7	Parks
	8	Restaurants
	9	City Trip
In the second seco	10	Architecture
	11	Friendly People
ng hocal Food Nature Beach Adventure Food Sunsets	alleries	ered door skiing weath spas indie Rock



Problem Setup

• Challenge:

o Recommender or Information Retrieval System?

Information Retrieval:

O Users have information need which is expressed as a query
O System has to satisfy this information needs

Recommender:

System predicts what users might be interested

We have both: 1) Users search for activities 2) Users don't know where they want to go

Why is It Hard?

Problem Characteristics:

- S Sparsity: new or rare users/destinations
- V Volatility: users' interests/endorsements of destinations change over time
- I Identity: a failure to match data from the same users
- P Personas: users have different interests at different, possibly closely points in time

Continuous Cold Start Problem!

Ranking Destinations for 'Beach'

- Keep it simple!!! We care about performance!
- Naïve Bayes:

P(Miami, Beach) = P(Miami) * P(Miami | Beach)

P(Miami Beach) = _	# 'Beach' endorsements for Miami		
	# 'Beach' endorsements		

What and How to Compare?

- Booking.com Baseline
- Random

- Most Popular Destination
- Naïve Bayes

Objection: Increase User Engagement (Clickers per SERP)

A/B Testing Setup

- 50/50 traffic split
- Experiments run for *N* full weeks according to desired power and significance levels
- Hypothesis tests are performed according to targeted metrics (G-test in our case)

Version A



1202 guests have endorsed this place for "Sand Beaches"



Version B





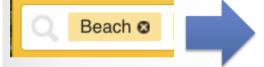


Version C









<u>Ranker</u>	<u># Users</u>	<u>Engagement</u>	
Baseline	9928		
Random	10079	Amount of	
Popularity	9838	clickers	
Naïve Bayes	9895		

<u>Ranker</u>	<u># Users</u>	<u>Engagement</u>
Baseline	9928	25.61% +/- 0.72%
Random	10079	24.46% +/- 0.71%
Popularity	9838	25.50% +/- 0.73%
Naïve Bayes	9895	26.73% +/- 0.73%

<u>Ranker</u>	<u># Users</u>	<u>Engagement</u>
Baseline	9928	25.61% +/- 0.72%
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Back to Question Why Is It Hard

Conclusion and Future Work

- Interesting application
- Surprise 1: Keep simple baseline in production system
- Surprise 2: Random performed not bad => effect serendipity

For the Future:

Improve the ranking by taking contexts into account

Thank you!

- Context identification and integration it into prediction models
- Accurately predicting users' desired actions and understanding behavioral patterns of users in various web-applications
- Personalization and adaptation to diverse customer needs and preferences
- Accounting for the practical needs within the considered application areas