Learning to Rank journey at GetYourGuide: Our Logbook

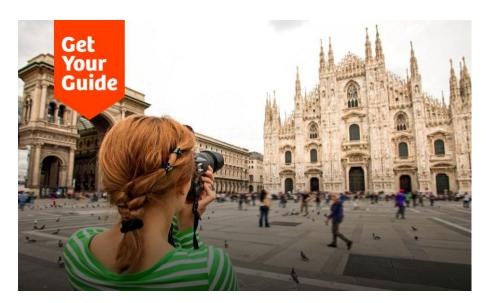
Felipe Besson • 11 Jun 2018



About GetYourGuide

GYG is the leading global marketplace for tours and activities

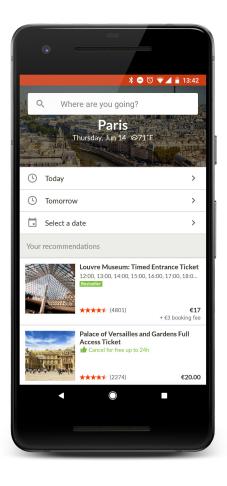
- 33k+ activities
- 7K+ destinations
- 415+ employes
 - 20 in the Data team
- Based in Berlin and Zurich

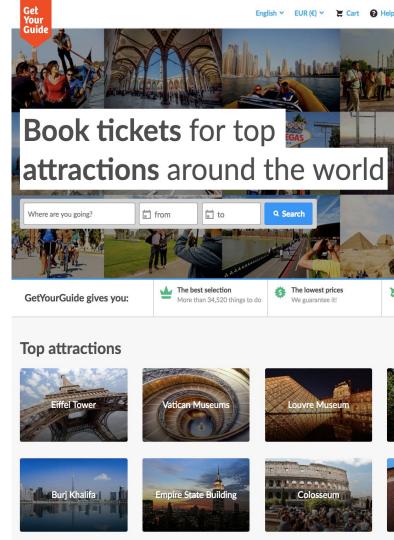


Full-text Search

Location drivenDiscovery

Rank	
Business metrics)
+	
Text Relevance)





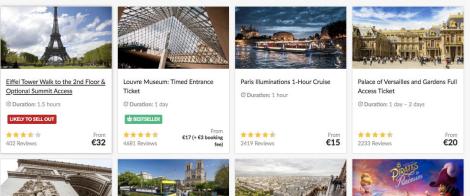
Location Pages (LPs)

- Location driven
- Dates are **very** important
- **High-intent customers**
- **Paid traffic**





Top attractions in Paris











Eiffel Tower: Second Floor Priority Disneyland Paris 1-Day Ticket



Paris: Skip-the-Line Arc de Triomphe **Rooftop Tickets**

Paris: 1, 2, or 3-Day Hop-on Hop-off Ticket

Access

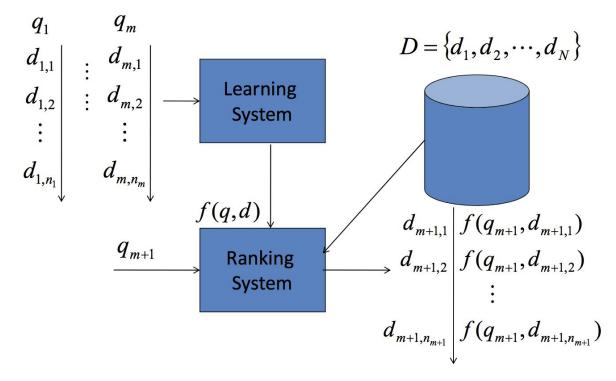
C Duration: 10 hours - 1 day

Problems with LP Ranking

- Focus on business metrics
- Customer intentions (search keywords)
 - □ "Eiffel Tower ticket" = "Eiffel Tower restaurant"
- Difficult to introduce new and diverse products

We needed to learn how to rank activities in LPs!

Learning to Rank (LTR)



Extracted from ACML 2009 Tutorial Nov. 2, 2009 Nanjing

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Berlin Buzzwords 2018

Our Logbook

Scope

- Decisions
- Tools and data pipelines
- Lessons learned

Not covered by this talk

Heavy statistical modelingEvaluation of models



First iteration (V1) Scope and decisions

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Learning to Rank (LTR) at GYG

Apply Machine Learning to introduce **relevance** factors into our ranking formula

Use our user intention data to have a dynamic LP ranking

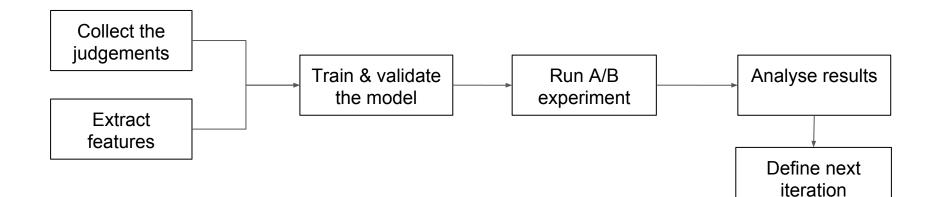
V1 Focus

- Vertical: Points of Interest
 - Ticket, Tour, Museum, Historic site, park, ...
- Only in English (we have 22 languages)
- Location pages have no explicit user query
 - Search Keywords:



MVP mindset

Follow the standard steps of a LTR solution



We started the journey!

Judgement List

- Ground truth set (Golden set)
- Highest traffic keywords

keyword	location id	doc id	judgement
eiffel tower summit access	2600 #Eiffel tower	42 #Optional Summit Access	?
eiffel tower summit access	2600 #Eiffel tower	24 #Paris cooking class	?
audio guide	2600 #Eiffel tower	41 #Audio Guided Eiffel Tower	?
<u>΄</u>			<u> </u>

query (qid)

Human labeling judgement list

- Judgements were collected from Domain Experts
 - Internal stakeholders of GYG
- Judgement scale from 0 3
- ~ 30k judgements
- Pre analysis of current rank
 - □ NDCG@7 = **0.55**

Human labeling judgement list



- ✓ Good approach when data is incomplete/inconsistent
- When what is a relevant result is still unclear
- No need to normalize queries deeply

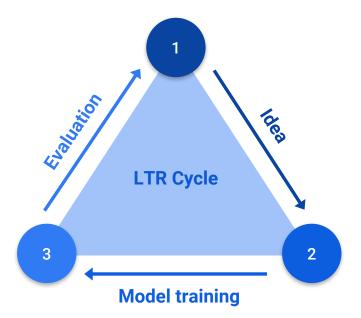
- X Relevance is subjective from user to user
- X Hard to scale
- X Crowdsourcing is expensive

Enriching Judgements with features

Feature Engineering

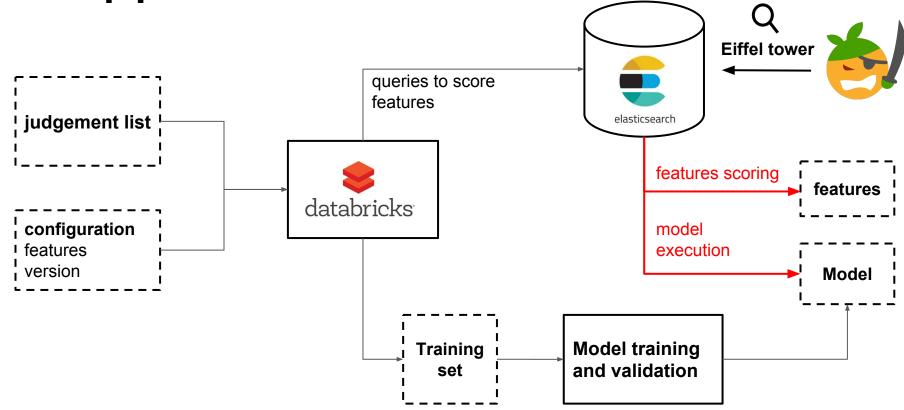
Query document	Business metrics	Document
 BM25 of single text fields 	 Raw metrics: clicks, bookings, impressions 	 Activity attributes: price, duration, # reviews
 Multi-match combinations 	• Rates: CTR, CR	 Categories (Dummy variables)

Why to develop a data pipelines?



- **Reliable**
- Reproducible
- Fast Iterations

First pipeline to collect features



Problems

Managing requests to ES

 300 features x 30k rows = 9mi queries

 Versioning features and training sets

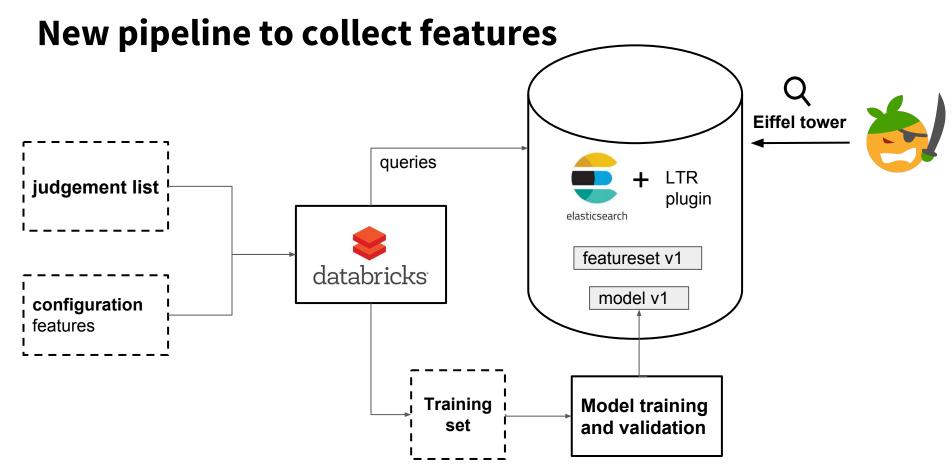
How to execute complex models on query time ?



Elasticsearch LTR plugin

- Developed by OpenSource Connections
- RankLib and XGBoost
- Features and models storage (versioning)
- Endpoint for scoring features
 - Fewer queries to ES (one per judgement)

repo: <u>https://github.com/o19s/elasticsearch-learning-to-rank</u>



Our pipeline

- ✓ Iterate faster to get new datasets
- Versioning of models and feature sets
- ✓ Simplicity to run models



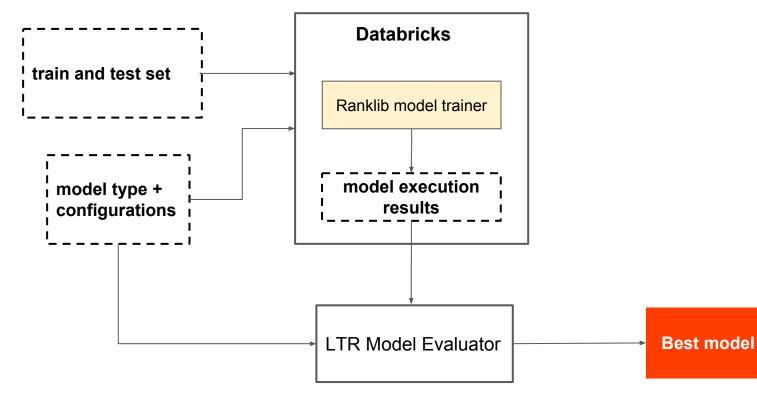
 Not automated from end to end (but good enough)

Training and validating Models

Goals

- Have a reasonable model for location pages
 - relevance + business metrics
- Evaluation metric: NDCG@10
- Success (business): **CTR** (Click-Through Rate)
- Constraints
 - Do not include user features

Pipeline to train and test Ranklib models



Model training pipeline

Iterate faster to try new modelsAutomate models evaluation



- Initial overhead to distribute
 Ranklib, we could use XGBoost
- X Analyze feature importance with Ranklib

Best V1 Model

LambdaMart \square NDCG@10 = 0.9282

Query document	Business metrics	Document
 Title Highlight Description Best field multi-match 	 Clicks Bookings Impressions CR 	 #Reviews Review rating Deal price Best seller



Best V1 model didn't work

Eiffel Tower Walk to the 2nd Floor & Optional Summit Access Duration: 1.5 hours LIKELY TO SELL OUT Eiffel Tower Guided Climb with Optional Summit Access Duration: 2. hours LIKELY TO SELL OUT Paris Elffel Tower Skip-the-Line Tour and Summit Access	From US\$40.22 From US\$42.73	1	
Arrow 240 Duration: 2 hours LIKELY TO SELL OUT Paris Eiffel Tower Skip-the-Line Tour and Summit Access			
Duration: 90 minutes	From US\$86.72		Gene
Paris: Elffel Tower by Night Skip-the-Line Guided Tour	From US\$45.25 US\$56.56		
Skip-the-Line Eiffel Tower Ticket and 2nd Level Tour	From US\$57.81		
Eiffel Tower: Skip-the-Line Summit Tour ***** 6 Duration: 90 minutes	From US\$75.28		
Eiffel Tower Priority Access to the Second Floor	From US\$61.58		a ar

Q "Eiffel tower skip-the-line ticket"

	Skip The Line: 2-Hour Guided Eiffel Tower Tour ****** 198 Duration: 2 hours LIKELY TO SELLOUT	From US\$86	
S	Elffel Tower Dinner With River Cruise	From US\$209	
<u>F</u> L	Guided Tour to Eiffel Tower 2nd Floor with Priority Access No reviews yet Duration: 90 minutes NEW ACTIVITY	Intro Special From US\$55	_
and	Paris: City Tour and Skip-the-Line Eiffel Tower Ticket Duration: 2 hours	From US\$273	
	Louvre, Notre Dame, Seine Cruise & Eiffel Tower Lunch	From US\$215	
	Eiffel Tower, Hop on Hop Off & Option for Cruise Ticket *****1 Duration: 1 day	From US\$110	
	Paris: Eiffel Tower Dinner & Seine River Cruise	From US\$219	
A	Eiffel Tower: Skip-the-Line Summit Tour	From US\$75.28	
à	Priority Eiffel Tower Tickets & Audio Guided Paris Bus Tour	From US\$86	

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> From US\$50.90

Main lessons learned

- Relevance of results for LP
- Judgement list extraction
- Quality of our queries
- Distribution of judgements

What is relevance for our business?

Our use case: Location pages

- First point of contact of many visitors
- Few rank positions to change
- Business metrics matter (e.g., revenue)

Domain Experts labeling

- □ This document is relevant for this query ? 0 3
- This document is a potential conversion ?



Another approach

Data approach for e-commerces

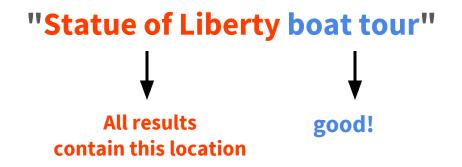
- Perceived utility of:
 - search results (Click through rate)
 - product page (Add-to-cart)
- Overall user satisfaction (Conversion)
- Business value (Revenue)

Experts could refine judgements collected from data

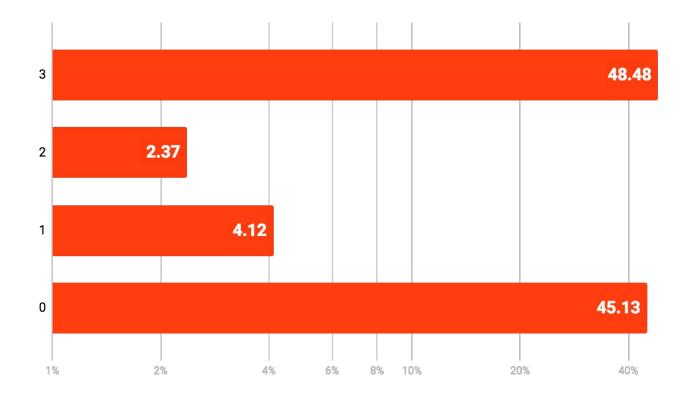
Reference: On Application of Learning to Rank for E-Commerce Search by Santu, Sondhi and Zhai (2017)

Quality of our queries

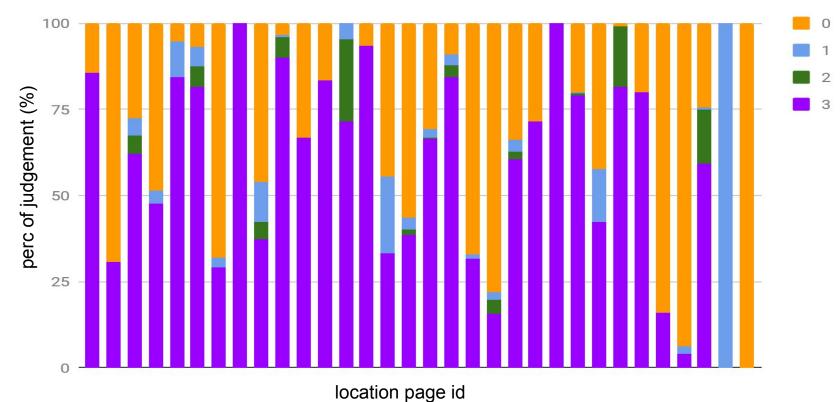
- Didn't consider real user query but the keyword search engine matches
- Location part is not relevant for scoring some queries



Distribution of our Judgements



Distribution of our Judgements per page

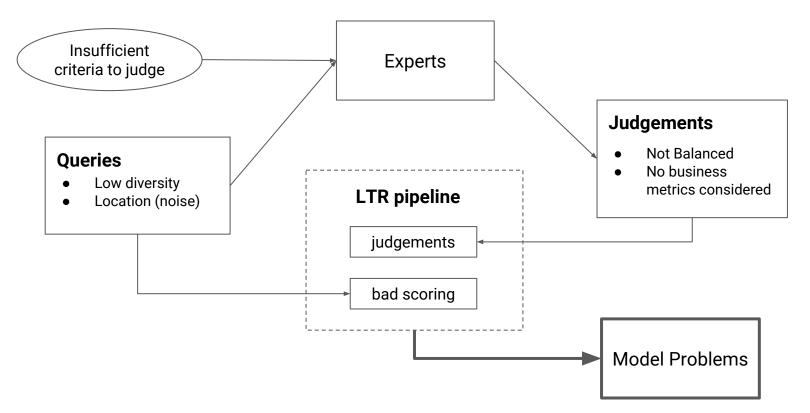


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We have very similar items in LP!

- In general: judgements didn't reflect those differences (binary relevance)
- In each page: very different and not balanced judgement distribution

Everything is connected



Next steps for V2

- Collect judgements from data
- Redefine our criteria for measuring relevance
- Apply LTR in another GYG search features
- Extract the intentions from the keywords
 - Query understanding might help
- Judge the judgements very often

We need to keep sailing!

Thank you!

Questions?

A 6 B

