

# Point-of-interest recommendation for travelers. Special Edition of the Text REtrieval Conference (TREC) Contextual Suggestion Track at RuSSIR 2015.

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## Overview

The goal of the hackathon was to elaborate recommendation for three users based on the history of their preferences: for each visited .

## Data munging

The test data emulates most of the real world problems such as:

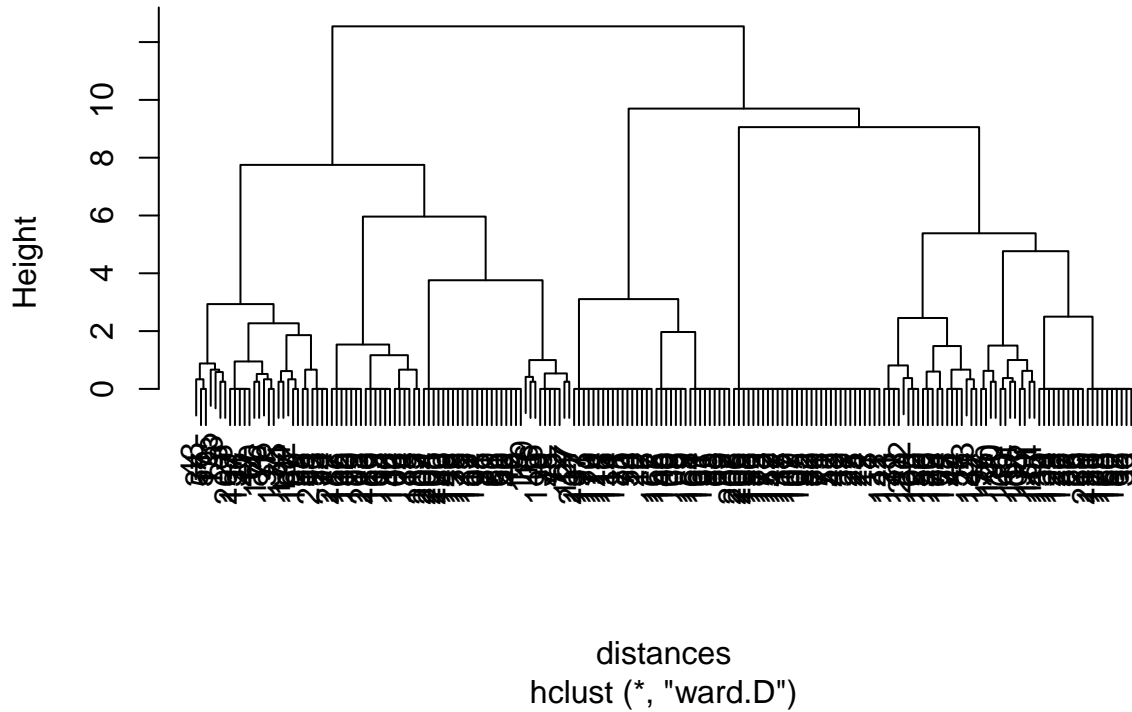
- original data provided in JSON format and need to be converted before analysis,
- missed data (there are 214 profiles but only 194 of them are complete, also one of the target user doesn't have any ratings for visited place),
- inconsistent data (e.g. gender might be “mail”, “femail” or “f”),
- data duplication (e.g. there are many very similar categories: “Art Galleries”, “Museums” and “Fine Art Museums” or “Coffee” and “Cafes” etc.)

Some of these issues were solved by using R while others were fixed manually.

## Split users into clusters

Users were split into five clusters by their gender, group, trip season, duration and type. Although it wasn't obvious while looking at hierarchical clustering dendrogram, the goal was to produce ad hoc solution such that target users with similar traits end up in the same cluster, but not the all three target users:

## User clustering



Target users:

	id	age	gender	group	season	duration	trip.type	cluster
212	1234567	50	male	friends	summer	longer	holiday	3
213	1234568	47	male	friends	autumn	longer	holiday	3
214	1234569	30	female	alone	summer	weekend trip	holiday	2

## Calculate target users preferences:

Weights of tags for user with *id=1234567* were calculated by averaging ranks of tags for *cluster 3* because there are no any tag for this user:

	newtags	rating
60	nature	3.159575
65	tropical	3.000000
49	cafs	2.771930
62	restaurant	2.628141
52	drinking	2.586956
45	art	2.581967
58	museum	2.549618

	newtags	rating
55	history	2.500000
61	relax	2.462428
46	authentic	2.336956
57	leisure	2.242991
44	active	2.177632
59	music	2.160000
50	city	2.104651
64	sport	2.026316
63	shopping	2.015544
51	cold weather	2.000000
56	hot weather	2.000000
66	warm weather	2.000000
53	family friendly	1.980392
54	fast food	1.900000
48	business	1.684211
67	wellness	1.642857
47	budget friendly	1.000000

Weights of tags for user with  $id=1234568$ :

	newtags	rating
2093	art	3.000000
2094	authentic	3.000000
2099	museum	2.750000
2098	leisure	2.666667
2103	restaurant	2.451613
2092	active	2.333333
2096	drinking	2.000000
2101	nature	2.000000
2105	sport	2.000000
2100	music	1.800000
2102	relax	1.800000
2095	city	1.500000
2104	shopping	1.333333
2097	fast food	0.000000

Weights of tags for user with  $id=1234569$ :

	newtags	rating
2107	art	4.000000
2109	leisure	4.000000
2111	music	3.000000
2112	relax	3.000000
2113	restaurant	2.916667
2110	museum	2.500000
2114	sport	2.000000
2106	active	0.000000
2108	authentic	0.000000

## Make predictions

Recommendations for user with  $id=1234567$ :

[illegible]

Recommendations for user with  $id=1234568$ :

	name	url
16	Grand Maket Russia Interactive Museum	<a href="http://www.tripadvisor.com/Attraction_Review-g298507-d3396387-F">http://www.tripadvisor.com/Attraction_Review-g298507-d3396387-F</a>
26	Faberge Museum	<a href="http://www.tripadvisor.com/Attraction_Review-g298507-d6580510-F">http://www.tripadvisor.com/Attraction_Review-g298507-d6580510-F</a>
44	Mikhailovsky Theatre / Mikhailovsky Theatre	<a href="https://foursquare.com/v/%D0%BC%D0%B8%D1%85%D0%B0%D0%">https://foursquare.com/v/%D0%BC%D0%B8%D1%85%D0%B0%D0%</a>
68	Grand Model Russia	<a href="https://foursquare.com/v/%D0%B3%D1%80%D0%B0%D0%BD%D0%">https://foursquare.com/v/%D0%B3%D1%80%D0%B0%D0%BD%D0%</a>
32	Percorso	<a href="http://www.tripadvisor.com/Restaurant_Review-g298507-d5247712-1">http://www.tripadvisor.com/Restaurant_Review-g298507-d5247712-1</a>
41	Severyanin	<a href="http://www.tripadvisor.com/Restaurant_Review-g298507-d2645717-1">http://www.tripadvisor.com/Restaurant_Review-g298507-d2645717-1</a>

