

Airbnb

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Abstract

Here is the abstract.

Introduction

[leíró statok: eu bizottság felhívása]

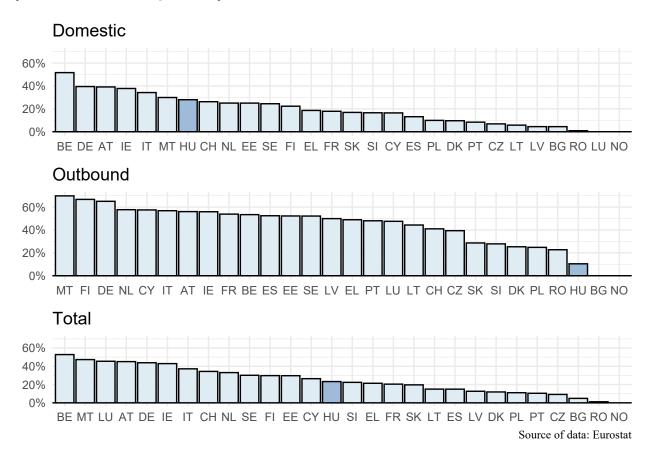


Figure 1: Proportion of internet bookings of the main means of accommodation by countries and the partner type

[1] "sf" "data.frame"

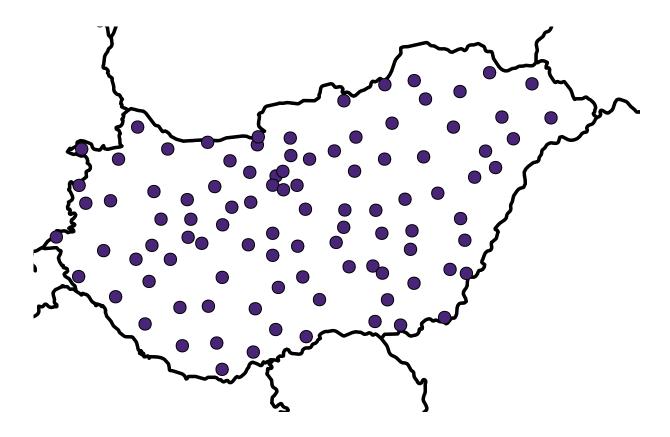


Figure 2: Starting points to our scrapping algorithm

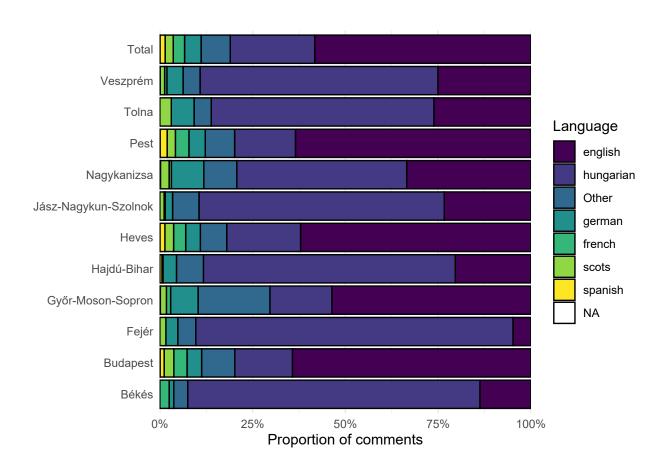


Figure 3: Most common languages found in the comments by counties



Figure 4: Words which correspond to high or low rating

Theoretical consideration

Data

Explore data

Model building

Consequences

Summary

References

Silge, J. & Robinson, D. (2017), Text mining with R: A tidy approach, O'Reilly Media, Sebastopol, CA.

Zervas, G., Proserpio, D. & Byers, J. W. (2017), 'The rise of the sharing economy: Estimating the impact of airbnb on the hotel industry', *Journal of Marketing Research* **54**(5), 687–705.

Appendix: R codes

```
# Packages ---
   library(tidyverse)
   library(knitr)
   library(tidytext)
   load('dat.RData')
   update geom defaults("point", list(fill = "#B1339E",
10
                                      shape = 21,
                                      color = "black".
12
                                      size = 1.4))
13
   update_geom_defaults("line",
14
                        list(color = "midnightblue", size = 1.4))
15
16
   update_geom_defaults("smooth", list(color = "red4", size = 1.4))
17
   update_geom_defaults("density",
19
                        list(color = "midnightblue", fill = "midnightblue",
20
                             alpha = .3, size = 1.4))
21
22
   extrafont::loadfonts(device="win")
23
   theme set(theme minimal() + theme(
25
     legend.direction = "vertical",
     plot.caption = element text(family = "serif")
27
   ))
29
   f.plot_eurostat <- function(x){</pre>
30
     eurostat::get_eurostat('tour_dem_ttorg', time_format = 'num') %>%
31
       filter(trip_arr %in% c('ACC_WEB', 'TOTAL') & duration == 'N1-3' & time == 2017 & purpose == 'TOTAL'
32
       pivot_wider(names_from = trip_arr, values_from = values) %>%
33
       mutate(
34
         value = ACC_WEB/TOTAL
35
       ) %>%
36
       filter(partner == x) %>%
37
       mutate(geo = fct reorder(geo, -value)) %>%
38
       ggplot() +
       aes(geo, value, fill = geo == 'HU') +
40
       geom_hline(yintercept = 0) +
41
       geom col(color = 'black') +
42
       scale_fill_brewer(palette = 3, guide = F) +
       scale_y_continuous(labels = scales::percent, limits = c(0, .7)) +
44
       labs(
45
         x = NULL, y = NULL, title = case_when(
46
           x == 'DOM' ~ 'Domestic',
           x == 'OUT' ~ 'Outbound',
48
           T ~ 'Total'
49
         )
50
51
52
```

```
53
    ggpubr::ggarrange(
      f.plot eurostat('DOM'),
55
      f.plot_eurostat('OUT'),
      f.plot_eurostat('WORLD') +
57
        labs(caption = 'Source of data: Eurostat'),
      ncol = 1
59
    )
60
61
    hun_cities <- read_csv("worldcities.csv") %>%
62
      filter(country == "Hungary") %>%
63
      select(city, lat, lng, admin name, population)
64
65
    cities <- readxl::read excel("cities.xlsx")</pre>
66
    library("rnaturalearth")
68
    library("rnaturalearthdata")
70
    world <- ne_countries(scale = "large", returnclass = "sf")</pre>
    class(world)
72
73
    merge(cities, hun_cities, by = 'city') %>%
74
      tibble() %>%
75
      ggplot() +
76
      geom sf(data = world, size = 1.2, fill = 'white', color = 'black') +
      coord_sf(xlim = c(16, 23.4), ylim = c(45.5, 48.7), expand = FALSE) +
78
      geom_point(aes(x = lng, y = lat), size = 4,
79
                  shape = 21, fill = viridis::viridis(1, begin = .1)) +
80
      theme(
81
        text = element blank()
82
83
84
    lapply(dat, function(x) {
85
      tibble(city = x[["source"]][["city"]], comments = x$comments) %>%
         mutate(language = textcat::textcat(comments))
87
    }) %>%
      reduce(rbind) %>%
89
      mutate(
        language = fct_lump(language, n = 6) %>%
91
           fct_infreq()
      ) %>%
93
      merge(hun cities) %>%
      {rbind(., mutate(., admin name = 'Total'))} %>%
95
      mutate(admin_name = fct_reorder(admin_name, admin_name == 'Total')) %>%
96
      ggplot() +
97
      aes(y = admin name, fill = language) +
      scale_x_continuous(labels = scales::percent, limits = c(0,1), expand = c(0,0)) +
99
      geom_bar(color = "black", position = position_fill()) +
100
      scale_fill_viridis_d() +
101
      labs(x = 'Proportion of comments', y = NULL, fill = "Language")
102
103
    dat_words <- lapply(dat, function(x) {</pre>
104
      tryCatch({
105
        tibble(comments = x$comments) %>%
106
```

```
mutate(language = textcat::textcat(comments)) %>%
107
           filter(language == "english") %>%
108
           tail(-2) %>%
109
           tidytext::unnest_tokens(words, comments, to_lower = T) %>%
110
           mutate(assesment = x[["source"]][["assesment"]], n_reviews = x[["source"]][["n_reviews"]])
      }, error = function(e) NULL)
112
113
    ) %>%
114
      {reduce(Filter(f = Negate(is.null), .), rbind)}
116
    dat_words <- dat_words %>%
117
      filter(n_reviews > 10 & words != 'bővebben') %>%
118
      mutate(type = case_when(
119
        assesment >= quantile(assesment, .99) ~ 'High',
120
        assesment <= quantile(assesment, .01) ~ 'Low',
121
        T ~ 'Middle'
122
      ))
123
124
    dat_words_High <- dat_words %>%
125
      filter(type == 'High') %>%
      group by(words) %>%
127
      summarise(n = n()) \%
      ungroup() %>%
129
      merge(dat_words %>%
               group by(words) %>%
131
               summarise(total = n()) %>%
               ungroup()) %>%
133
      filter(n >= 10) %>%
134
      mutate(
135
        rtf = n/total, type = 'High'
136
      ) %>%
137
      arrange(desc(rtf)) %>%
138
      anti_join(data.frame(words = c(stopwords::stopwords(), "also", "can"))) %>%
139
      head(50)
140
    dat_words_Low <- dat_words %>%
142
      filter(type == 'Low') %>%
143
      group by(words) %>%
144
      summarise(n = n()) \%
      ungroup() %>%
146
      merge(dat_words %>%
147
               group by(words) %>%
148
               summarise(total = n()) %>%
               ungroup()) %>%
150
      filter(n >= 10) %>%
151
      mutate(
152
        rtf = n/total, type = 'Low'
153
154
      arrange(desc(rtf)) %>%
155
      anti_join(data.frame(words = c(stopwords::stopwords(), "also", "can"))) %>%
156
      head(50)
157
    rbind(dat_words_Low, dat_words_High) %>%
      reshape2::acast(words ~ type, value.var = "rtf", fill = 0) %>%
159
```

```
wordcloud::comparison.cloud(colors = viridis::viridis(2, direction = -1, end = .7),
max.words = 100)
```