

Expedia Training Data-set Analysis

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Exploring srch_id

Find the unique srch_id's and count how many of them are in the submission_sample data-set:

```
# Load Training data-set
tr <- read.csv("training_set_VU_DM.csv")
tr_row <- nrow(tr)
tr_col <- ncol(tr)

uniq_srch <- unique(tr$srch_id)
nrow(as.matrix(uniq_srch))

## [1] 199795

head(uniq_srch, 10)

## [1] 1 4 6 8 11 12 17 21 25 28

tail(uniq_srch, 10)

## [1] 332765 332768 332772 332774 332776 332777 332781 332782 332784 332785

last_srch_id <- tail(uniq_srch, n=1) # Highest srch_id in ascending list
perc_missing_srch <- (1 - (last_srch_id / tr_row)) * 100
```

Here we can see that a considerable amount of srch_id's are missing from the sequence (93%)

Exploring NULL values

```
#which(is.null(tr$srch_id)) # <-- this does not work!
#which(tr$srch_id == "NULL") # Not a real NULL value; it's only a string
#which(tr$comp3_rate == "NULL") # testing function on a column that contains "NULL"

# Create a new table for analyzing NULL stats
df_null_ratios <- data.frame(matrix(ncol = 3, nrow = 54))
names(df_null_ratios) <- list("Column_Name", "Number_of_Nulls", "Percentage_of_Rows")

i <- 0
for (i in 1:tr_col) {
  # Populate column names from the training set into new table
  df_null_ratios[i, 1] <- names(tr)[i]

  # Find the rows in the training set that contain string "NULL"
  null_rows <- which(tr[,i] == "NULL")
}
```

```

# Populate "NULL" counts into new table
num_nulls <- nrow(as.matrix(null_rows))
df_null_ratios[i, 2] <- num_nulls

# Populate percentage of counts into new table
null_perc <- (num_nulls / tr_row) * 100
df_null_ratios[i, 3] <- null_perc
}

df_null_ratios

```

##	Column_Name	Number_of_Nulls	Percentage_of_Rows
## 1	srch_id	0	0.00000
## 2	date_time	0	0.00000
## 3	site_id	0	0.00000
## 4	visitor_location_country_id	0	0.00000
## 5	visitor_hist_starrating	4706481	94.92036
## 6	visitor_hist_adr_usd	4705359	94.89774
## 7	prop_country_id	0	0.00000
## 8	prop_id	0	0.00000
## 9	prop_starrating	0	0.00000
## 10	prop_review_score	0	0.00000
## 11	prop_brand_bool	0	0.00000
## 12	prop_location_score1	0	0.00000
## 13	prop_location_score2	1090348	21.99015
## 14	prop_log_historical_price	0	0.00000
## 15	position	0	0.00000
## 16	price_usd	0	0.00000
## 17	promotion_flag	0	0.00000
## 18	srch_destination_id	0	0.00000
## 19	srch_length_of_stay	0	0.00000
## 20	srch_booking_window	0	0.00000
## 21	srch_adults_count	0	0.00000
## 22	srch_children_count	0	0.00000
## 23	srch_room_count	0	0.00000
## 24	srch_saturday_night_bool	0	0.00000
## 25	srch_query_affinity_score	4640941	93.59855
## 26	orig_destination_distance	1607782	32.42577
## 27	random_bool	0	0.00000
## 28	comp1_rate	4838417	97.58125
## 29	comp1_inv	4828788	97.38705
## 30	comp1_rate_percent_diff	4863908	98.09535
## 31	comp2_rate	2933675	59.16639
## 32	comp2_inv	2828078	57.03671
## 33	comp2_rate_percent_diff	4402109	88.78179
## 34	comp3_rate	3424059	69.05646
## 35	comp3_inv	3307357	66.70281
## 36	comp3_rate_percent_diff	4485550	90.46462
## 37	comp4_rate	4650969	93.80080
## 38	comp4_inv	4614684	93.06900
## 39	comp4_rate_percent_diff	4827261	97.35626
## 40	comp5_rate	2735974	55.17916
## 41	comp5_inv	2598327	52.40309

## 42	comp5_rate_percent_diff	4117248	83.03671
## 43	comp6_rate	4718190	95.15651
## 44	comp6_inv	4697371	94.73663
## 45	comp6_rate_percent_diff	4862173	98.06036
## 46	comp7_rate	4642999	93.64006
## 47	comp7_inv	4601925	92.81168
## 48	comp7_rate_percent_diff	4819832	97.20643
## 49	comp8_rate	3041693	61.34490
## 50	comp8_inv	2970844	59.91602
## 51	comp8_rate_percent_diff	4343617	87.60212
## 52	click_bool	0	0.00000
## 53	gross_bookings_usd	4819957	97.20895
## 54	booking_bool	0	0.00000