

main_sap_rozi_jednorozni

Rozi jednorozni

29 05 2020

##Inicijalizacija podataka, ciscenje dataframe, obrada testova nad amenitiesima

```
#file odakle su funkcije i koje biblioteke se koriste  
source("functions.R")  
library(ggplot2)
```

```
#Ucitavanje i odabir zeljenih columna  
airbnb<-read.csv("projekt.csv")  
airbnb_main_frame<-as.data.frame(airbnb)
```

```
#maknut sam id jer je u izvornom stanju neupotrebljiv, par redova dolje je dodan kako ide redosljed  
airbnb_bitno <- subset(airbnb_main_frame, select = -c( id,bed_type, description, first_review, host_ha
```

```
#ciscenje svih redaka koji sadrže NA vrijednosti  
airbnb_bitno<- na.omit(airbnb_bitno)
```

```
#dodan id vector, treba nam za funkciju get_positions_of_empty_amenities  
airbnb_bitno=get_Dataframe_With_Id_Row(airbnb_bitno)
```

```
#vector koji sadrži sve pozicije gdje je amenities prazan tj {}  
vector_of_empty_amenities=get_positions_of_empty_amenities(airbnb_bitno$amenities, airbnb_bitno$id)
```

```
#funkcija koja vraća dataframe bez redova gdje su amenities prazni  
airbnb_bitno=clean_Empty_Amenities(airbnb_bitno, vector_of_empty_amenities)
```

```
#ponovno skidamo id jer kad maknemo prazne redove onda nastaju rupe  
airbnb_bitno <- subset(airbnb_bitno, select = -c(id))
```

```
#refresh i dodavanje id-a nazad tako da ide 1,2,3,4.. do kraja bez rupa  
airbnb_bitno=get_Dataframe_With_Id_Row(airbnb_bitno)
```

```
numOfRows=nrow(airbnb_bitno)
```

```
#####
```

```
#stvara listu koja sadrži vector odvojenih amenities za svaki red  
#u petlji se još stvara i vector koji govori koliko je amenities u određenom redu  
list=list()  
number_of_amenities=c()
```

```

vector_of_all_amenities=c()
counter=0
counter2=0
for(i in 1:numOfRows) {
  temp_vector_of_amenities=make_Vector_Of_Amentities(airbnb_bitno$amenities[i])
  list[[i]]=temp_vector_of_amenities
  number_of_amentites[i]=length(list[[i]))-1
  size_of_temp=length(temp_vector_of_amenities)
  counter2=counter2+1

  for(j in 1:size_of_temp) {
    counter=counter+1
    vector_of_all_amenities[counter]=temp_vector_of_amenities[j]
  }
}
#dodaje dodatni column koji sadrzi broj amenitiesa za taj red
airbnb_bitno <- cbind(airbnb_bitno , number_of_amentites)

#list_of_vectors_of_amenities=list()

vector_of_unique_amenities=unique(vector_of_all_amenities)

list_of_tests=list()
#za svaki jedinstveni amenitie prolazi kroz dataframe
#pravi vector koji sadrzi true ili false vrijednost
#ovisno o tome je li stan u tom redu sadrzi taj amenitie
#na kraju spaja taj vector sa pocetnim dataframeom
#ovaj dio dodaje svih 129 amenitiera u airbnb_bitno
#zato je dataframe tako velik
count129=0
len_list=length(list)
number_of_apartments_that__have_certain_amenitie_vector=c()
for(i in 1:129) {

  temp_vector_of_amenities=rep(F, 73218)
  count129=count129+1
  print(paste("Loaded ", count129, "out of 129 amenities"))

  current_counter=0
  for (j in 1:len_list) {
    if(vector_of_unique_amenities[i] %in% list[[j]]) {
      temp_vector_of_amenities[j]=T
      current_counter=current_counter+1
    }
  }
  number_of_apartments_that__have_certain_amenitie_vector[i]=current_counter
  airbnb_bitno <- cbind(airbnb_bitno , temp_vector_of_amenities)
  names(airbnb_bitno)[names(airbnb_bitno) == "temp_vector_of_amenities"] <- vector_of_unique_amenities[i]

  if(i<124) {
    test=t.test(airbnb_bitno$price~temp_vector_of_amenities, mu=0, alt="two.sided", conf=0.99)
  }
}

```

```

    test$data.name=paste("price and ", vector_of_unique_amenities[i])
    list_of_tests[[i]]=test
  }
}

```

```

## [1] "Loaded 1 out of 129 amenities"
## [1] "Loaded 2 out of 129 amenities"
## [1] "Loaded 3 out of 129 amenities"
## [1] "Loaded 4 out of 129 amenities"
## [1] "Loaded 5 out of 129 amenities"
## [1] "Loaded 6 out of 129 amenities"
## [1] "Loaded 7 out of 129 amenities"
## [1] "Loaded 8 out of 129 amenities"
## [1] "Loaded 9 out of 129 amenities"
## [1] "Loaded 10 out of 129 amenities"
## [1] "Loaded 11 out of 129 amenities"
## [1] "Loaded 12 out of 129 amenities"
## [1] "Loaded 13 out of 129 amenities"
## [1] "Loaded 14 out of 129 amenities"
## [1] "Loaded 15 out of 129 amenities"
## [1] "Loaded 16 out of 129 amenities"
## [1] "Loaded 17 out of 129 amenities"
## [1] "Loaded 18 out of 129 amenities"
## [1] "Loaded 19 out of 129 amenities"
## [1] "Loaded 20 out of 129 amenities"
## [1] "Loaded 21 out of 129 amenities"
## [1] "Loaded 22 out of 129 amenities"
## [1] "Loaded 23 out of 129 amenities"
## [1] "Loaded 24 out of 129 amenities"
## [1] "Loaded 25 out of 129 amenities"
## [1] "Loaded 26 out of 129 amenities"
## [1] "Loaded 27 out of 129 amenities"
## [1] "Loaded 28 out of 129 amenities"
## [1] "Loaded 29 out of 129 amenities"
## [1] "Loaded 30 out of 129 amenities"
## [1] "Loaded 31 out of 129 amenities"
## [1] "Loaded 32 out of 129 amenities"
## [1] "Loaded 33 out of 129 amenities"
## [1] "Loaded 34 out of 129 amenities"
## [1] "Loaded 35 out of 129 amenities"
## [1] "Loaded 36 out of 129 amenities"
## [1] "Loaded 37 out of 129 amenities"
## [1] "Loaded 38 out of 129 amenities"
## [1] "Loaded 39 out of 129 amenities"
## [1] "Loaded 40 out of 129 amenities"
## [1] "Loaded 41 out of 129 amenities"
## [1] "Loaded 42 out of 129 amenities"
## [1] "Loaded 43 out of 129 amenities"
## [1] "Loaded 44 out of 129 amenities"
## [1] "Loaded 45 out of 129 amenities"
## [1] "Loaded 46 out of 129 amenities"

```

[illegible]

```
## [1] "Loaded 101 out of 129 amenities"
## [1] "Loaded 102 out of 129 amenities"
## [1] "Loaded 103 out of 129 amenities"
## [1] "Loaded 104 out of 129 amenities"
## [1] "Loaded 105 out of 129 amenities"
## [1] "Loaded 106 out of 129 amenities"
## [1] "Loaded 107 out of 129 amenities"
## [1] "Loaded 108 out of 129 amenities"
## [1] "Loaded 109 out of 129 amenities"
## [1] "Loaded 110 out of 129 amenities"
## [1] "Loaded 111 out of 129 amenities"
## [1] "Loaded 112 out of 129 amenities"
## [1] "Loaded 113 out of 129 amenities"
## [1] "Loaded 114 out of 129 amenities"
## [1] "Loaded 115 out of 129 amenities"
## [1] "Loaded 116 out of 129 amenities"
## [1] "Loaded 117 out of 129 amenities"
## [1] "Loaded 118 out of 129 amenities"
## [1] "Loaded 119 out of 129 amenities"
## [1] "Loaded 120 out of 129 amenities"
## [1] "Loaded 121 out of 129 amenities"
## [1] "Loaded 122 out of 129 amenities"
## [1] "Loaded 123 out of 129 amenities"
## [1] "Loaded 124 out of 129 amenities"
## [1] "Loaded 125 out of 129 amenities"
## [1] "Loaded 126 out of 129 amenities"
## [1] "Loaded 127 out of 129 amenities"
## [1] "Loaded 128 out of 129 amenities"
## [1] "Loaded 129 out of 129 amenities"
```

#izbacivanje testova koji nista ne znace:

#translationmissingenhostingamenity49 i translationmissingenhostingamenity50

```
clean_list_of_tests=list()
number_of_tests=length(list_of_tests)
counter=0
for (i in 1:number_of_tests) {
  if(list_of_tests[[i]]$data.name!="price and translationmissingenhostingamenity50" &&
    list_of_tests[[i]]$data.name!="price and translationmissingenhostingamenity49") {
    counter=counter+1
    clean_list_of_tests[[counter]]=list_of_tests[[i]]
  }
}
```

#pregled koliko stanova ima koji amenitie

```
names(number_of_apartments_that__have_certain_amenitie_vector)=vector_of_unique_amenities
number_of_apartments_that__have_certain_amenitie_vector
```

##	WirelessInternet	Airconditioning
##	70968	54994
##	Kitchen	Heating
##	67257	66806
##	Familykidfriendly	Essentials
##	36939	63855
##	Hairdryer	Iron
##	43240	41593
##	translationmissingenhostingamenity50	Washer
##	25226	43010
##	Dryer	Smokedetector
##	42554	61576
##	Fireextinguisher	Shampoo
##	30656	49343
##	Hangers	TV
##	49054	52159
##	CableTV	Breakfast
##	24143	8284
##	Buzzerwirelessintercom	Carbonmonoxidedetector
##	16955	47072
##	Laptopfriendlyworkspace	Internet
##	43605	44464
##	Indoorfireplace	Firstaidkit
##	9275	27478
##	Elevatorinbuilding	Pool
##	6400	6265
##	Freeparkingonpremises	Gym
##	23564	7462
##	Hottub	Wheelchairaccessible
##	6318	4836
##	Doorman	Dogs
##	4376	5240
##	Cats	Otherpets
##	3569	375
##	Lockonbedroomdoor	Petsliveonthisproperty
##	17952	9685
##	Privateentrance	Hotwater
##	7264	4255
##	Bedlinens	Extrapillowsandblankets
##	4166	3019
##	Coffeemaker	Refrigerator
##	3545	4637
##	Dishesandsilverware	Gardenorbackyard

##	4244	1128
##	Petsallowed	Safetycard
##	10168	11492
##	24hourcheckin	SelfCheckIn
##	18973	11028
##	Lockbox	Suitableforevents
##	5733	4248
##	translationmissingenhostingamenity49	Elevator
##	20366	10773
##	Microwave	Cookingbasics
##	3901	3946
##	Oven	Stove
##	3933	4070
##	Smokingallowed	Outletcovers
##	3663	446
##	Bathtub	Dishwasher
##	3760	2310
##	Singlelevelhome	Luggagedropoffallowed
##	518	1635
##	Stepfreeaccess	Wideclearancetobed
##	827	404
##	Accessibleheightbed	Widedoorway
##	348	560
##	Accessibleheighttoilet	Wideentryway
##	260	301
##	Hotwaterkettle	Babysitterrecommendations
##	173	677
##	Pack&nPlaytravelcrib	Keypad
##	1124	2992
##	Freeparkingonstreet	Smartlock
##	77	1442
##	Widehallwayclearance	Flat
##	472	396
##	smoothpathwaytofrontdoor	Welllitpathtoentrance
##	396	769
##	Longtermstaysallowed	Privatelivingroom
##	1666	2516
##	Children&sbooksandtoys	Children&sdinnerware
##	1126	635
##	BBQgrill	Patioorbalcony
##	632	1218
##	Hostgreetssystem	Ethernetconnection
##	1195	690
##	Fixedgrabbarsforshowertoilet	Highchair
##	67	635
##	Roomdarkeningshades	Changingtable
##	1496	225
##	Stairgates	Windowguards
##	299	612
##	Tablecornerguards	Crib
##	88	442
##	Other	Privatebathroom
##	217	74
##	DoormanEntry	Babymonitor

##	452	133
##	Babybath	Fireplaceguards
##	305	219
##	Gameconsole	Firmmattress
##	478	115
##	Pocketwifi	Cleaningbeforecheckout
##	217	268
##	Handheldshowerhead	Waterfront
##	135	115
##	Groundflooraccess	EVcharger
##	25	48
##	Wideclearancetoshower toilet	Beachessentials
##	122	102
##	Beachfront	Bathtubwithshowerchair
##	33	34
##	Lakeaccess	WasherDryer
##	18	28
##	Disabledparkingspot	Rollinshowerwithchair
##	34	3
##	Airpurifier	SkiinSkiout
##	14	13
##	Pathtoentrancelit atnight	Firmmattress
##	27	15
##	Paidparkingoffpremises	Flatsmoothpathwaytofrontdoor
##	7	11
##	Grabrailsforshowerandtoilet	Bodysoap
##	2	1
##	Handsoap	Bathtowel
##	1	1
##	Handorpaper towel	Toiletpaper
##	1	1
##	Wideclearancetoshowerandtoilet	
##	1	

```

List_of_prices_per_city=sort_Numeric_Categorical(airbnb_bitno$price, airbnb_bitno$city)
List_of_price_outliers_per_city=list()
len=length(List_of_prices_per_city)

#prebrojavanje outliera po gradovima
number_of_outliers=list()
relative_number_of_outliers=list()

for(i in 1:len) {
  num_of_apartments_per_city=length(List_of_prices_per_city[[i]])

  temp_vector=c(List_of_prices_per_city[[i]])

  Q1=quantile(temp_vector, 0.25)
  Q1=unnamed(Q1)
  Q3=quantile(temp_vector, 0.75)
  Q3=unnamed(Q3)
  IQR=Q3-Q1

```



```

donja_granica=Q1 - 1.5*IQR
gornja_granica=Q3 + 1.5*IQR
print(paste("donja_granica",donja_granica))
print(paste("gornja_granica",gornja_granica))

temp_prices=c()
counter=0
for(j in 1:num_of_apartments_per_city) {

  if(temp_vector[j] >= gornja_granica || temp_vector[j] <= donja_granica){
    counter = counter + 1
    temp_prices[counter]=temp_vector[j]

  }

  number_of_outliers[[i]]=counter
  relative_number_of_outliers[[i]]=counter/num_of_apartments_per_city
  List_of_price_outliers_per_city[[i]]=temp_prices
}
}

```

```

## [1] "donja_granica -87.50000000000001"
## [1] "gornja_granica 332.5"
## [1] "donja_granica -117.5"
## [1] "gornja_granica 486.5"
## [1] "donja_granica -125"
## [1] "gornja_granica 434.9999999999999"
## [1] "donja_granica -79.99999999999997"
## [1] "gornja_granica 320"
## [1] "donja_granica -62.5"
## [1] "gornja_granica 277.5"
## [1] "donja_granica -99.99999999999999"
## [1] "gornja_granica 380"

```

```

names(number_of_outliers) <- unique(airbnb_bitno$city)
names(relative_number_of_outliers) <- unique(airbnb_bitno$city)
names(List_of_price_outliers_per_city) <- unique(airbnb_bitno$city)
print("iz ispisa vidimo da su samo gornje granice bitne")

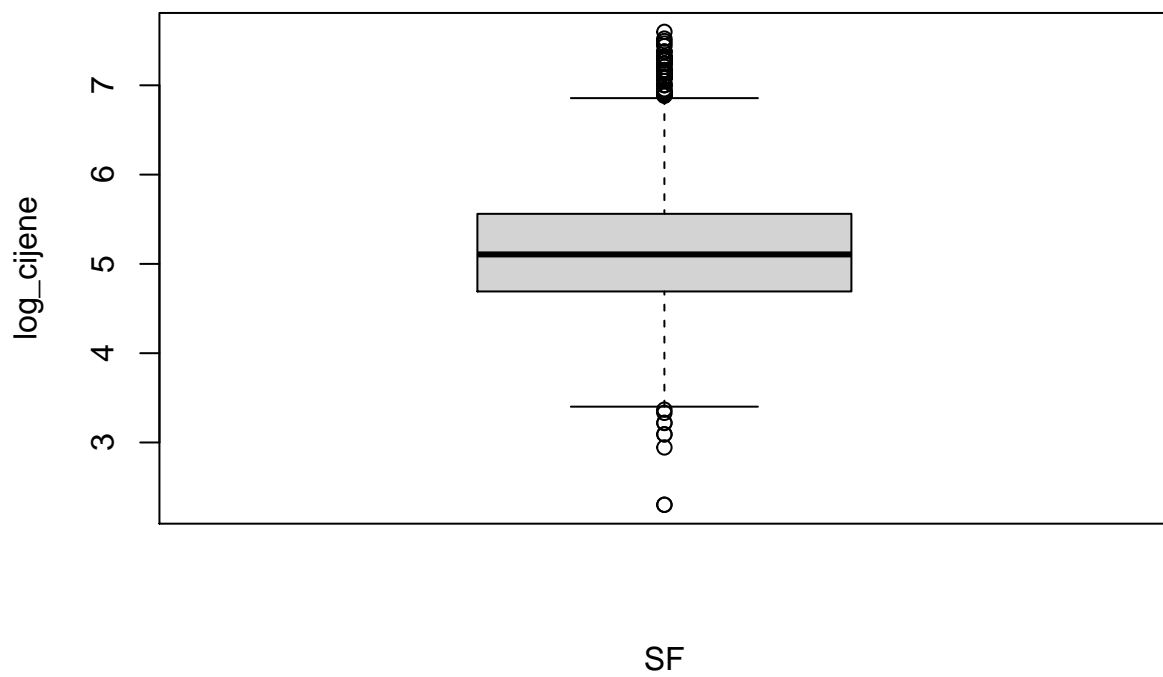
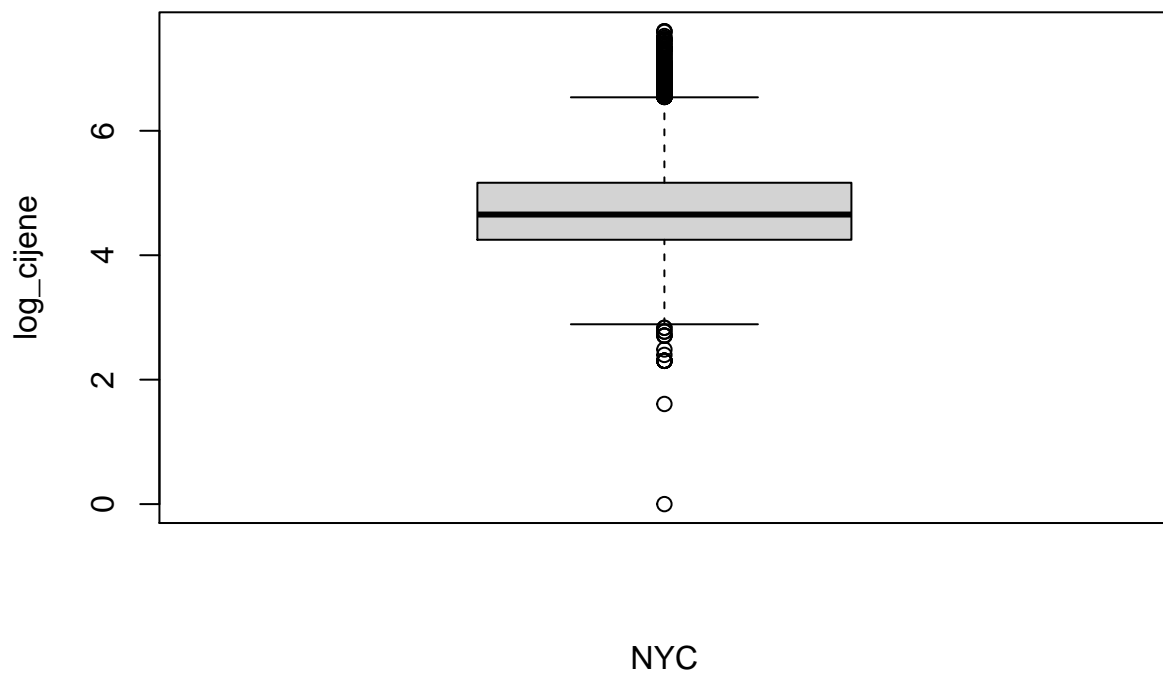
```

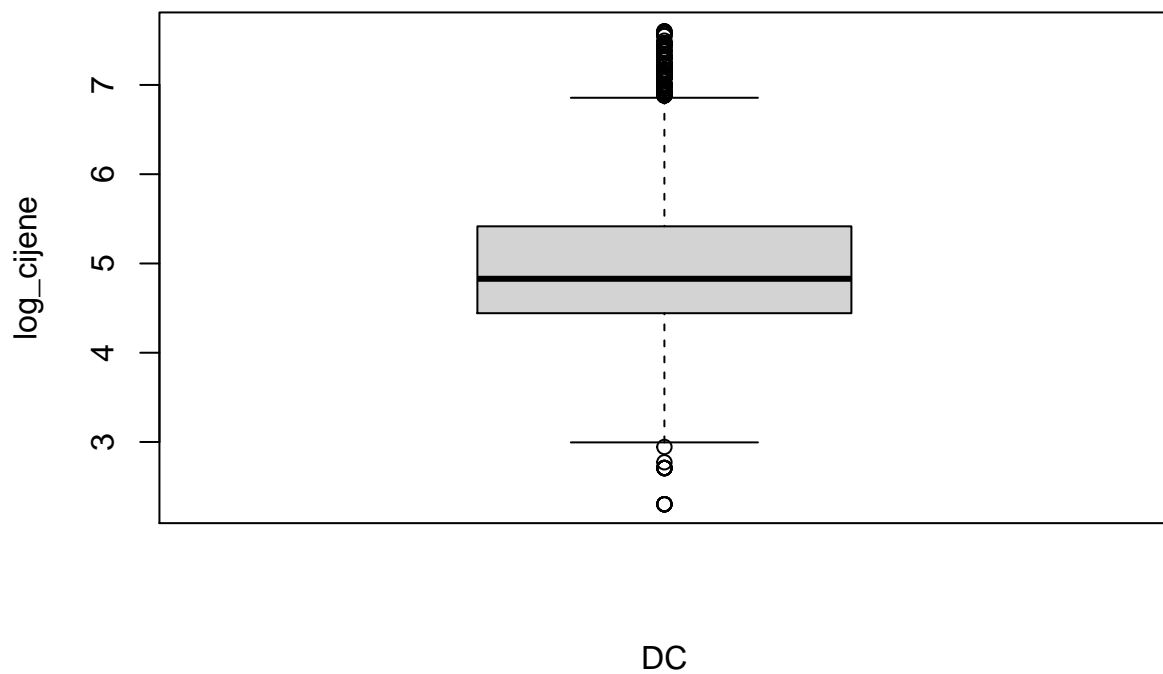
```
## [1] "iz ispisa vidimo da su samo gornje granice bitne"
```

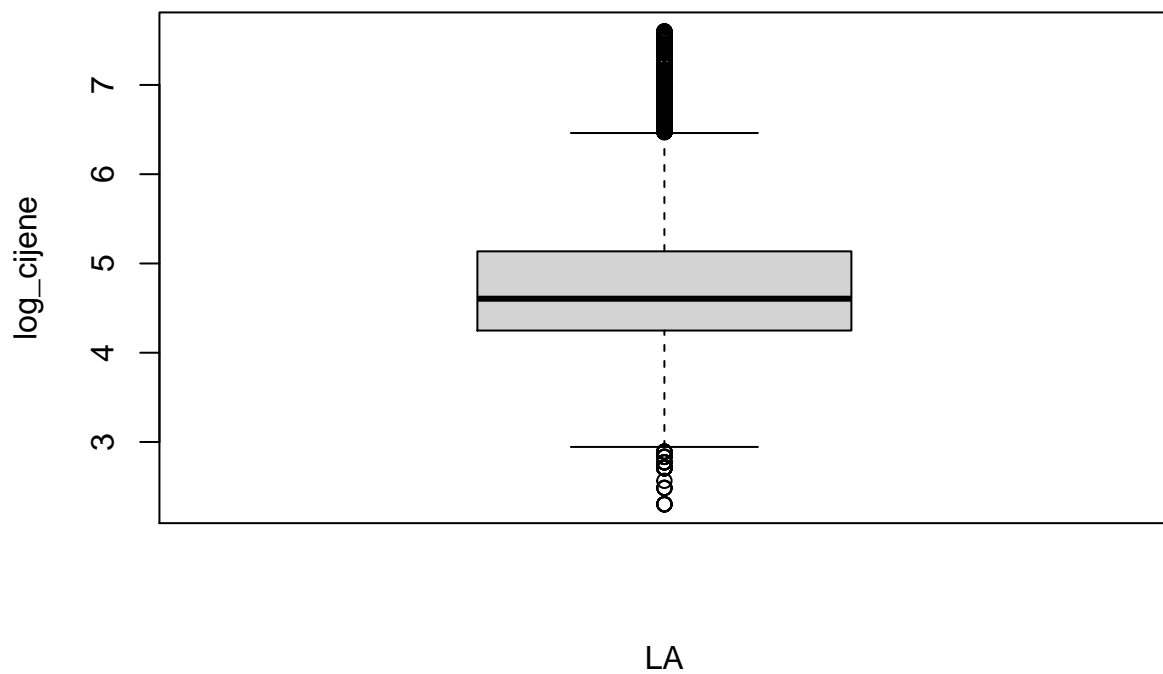
```
print("ocekivano, razdioba cijena je nagnuta u lijevo")
```

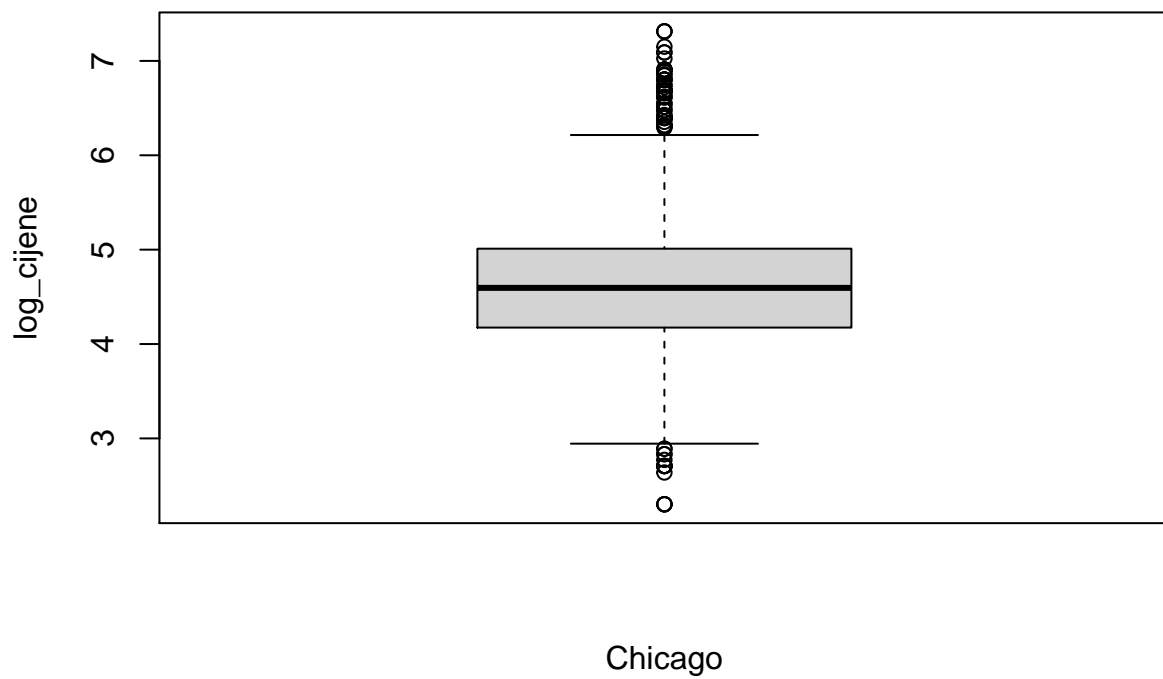
```
## [1] "ocekivano, razdioba cijena je nagnuta u lijevo"
```

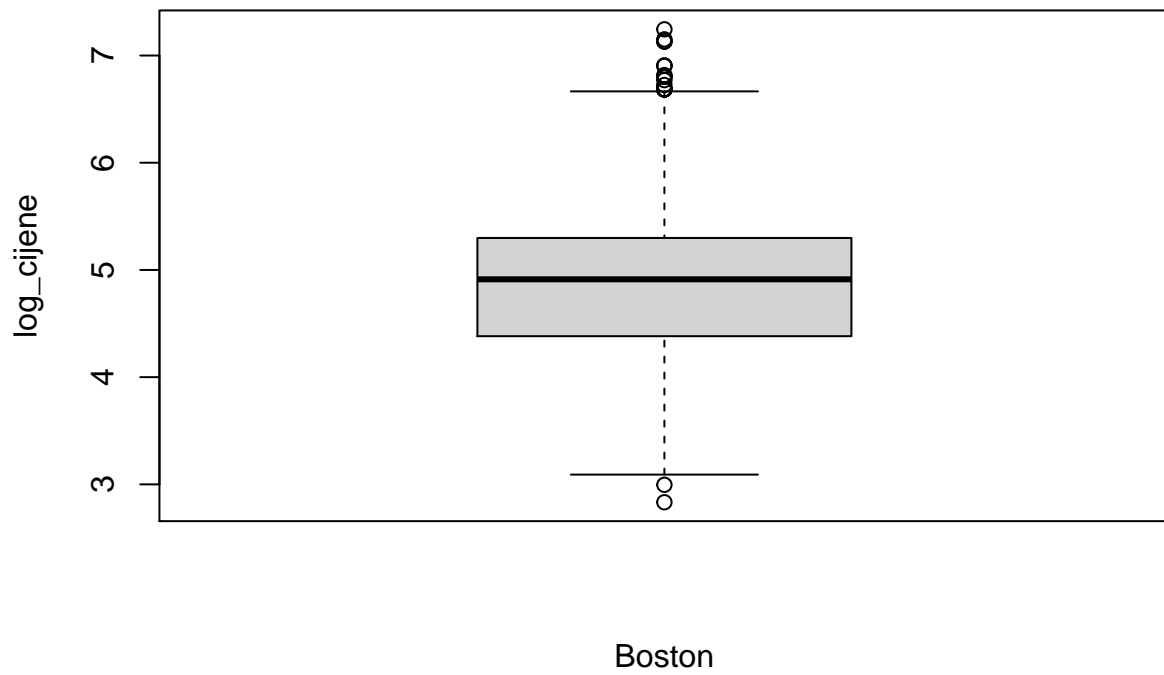
Deskriptivna statistika



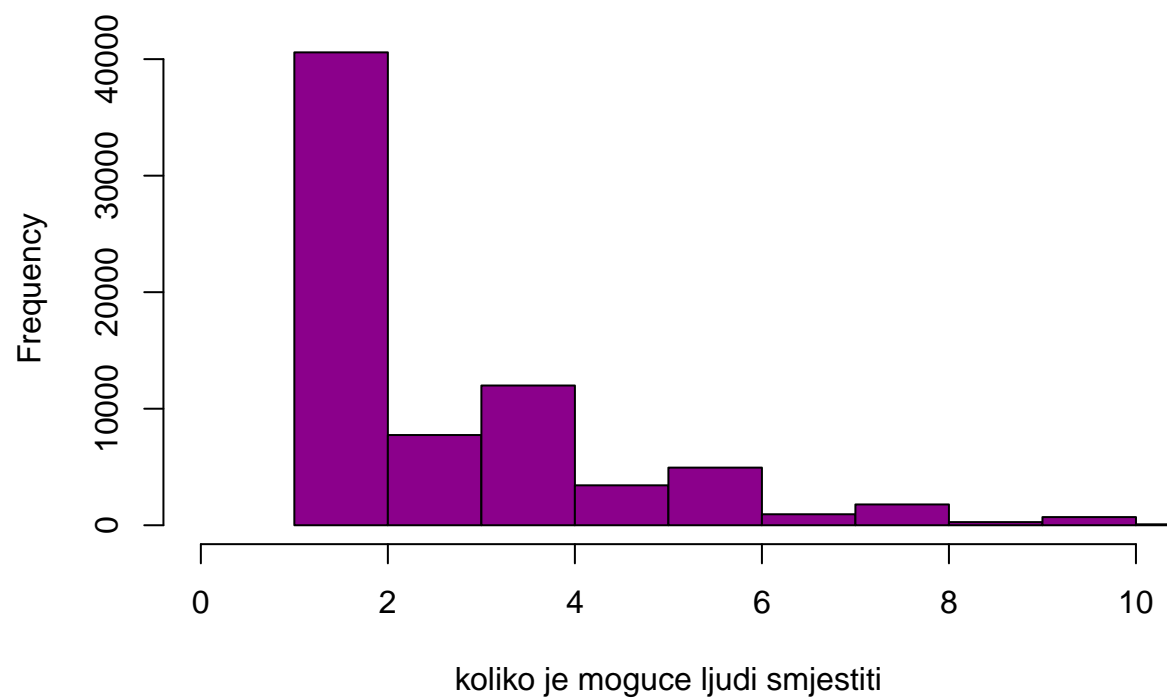




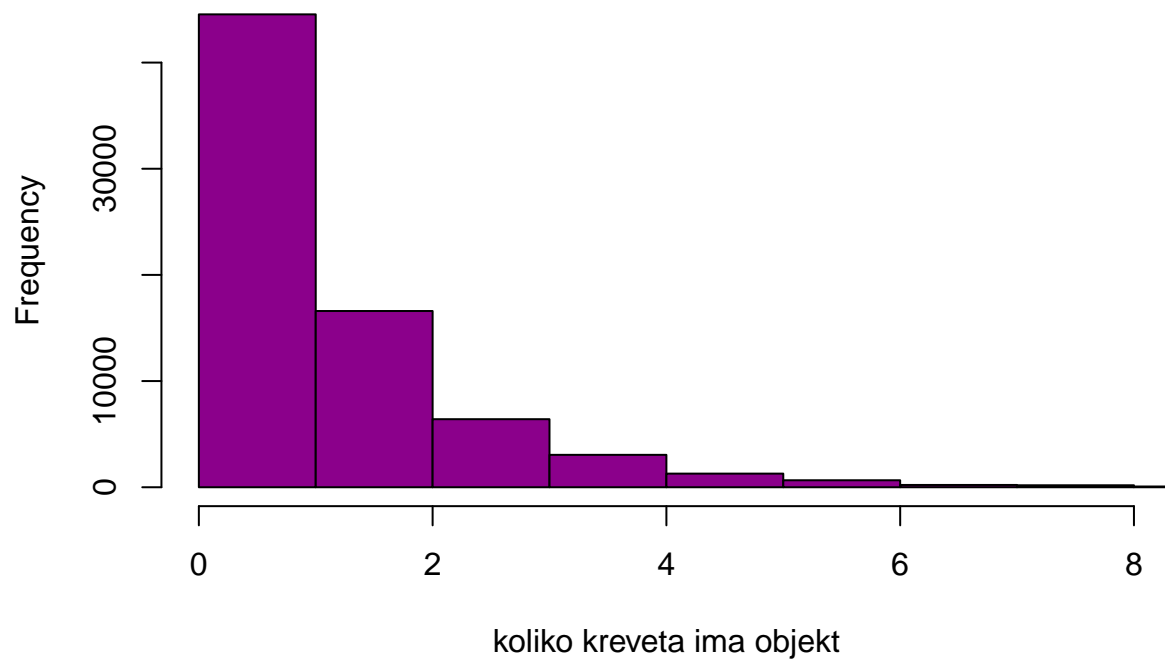


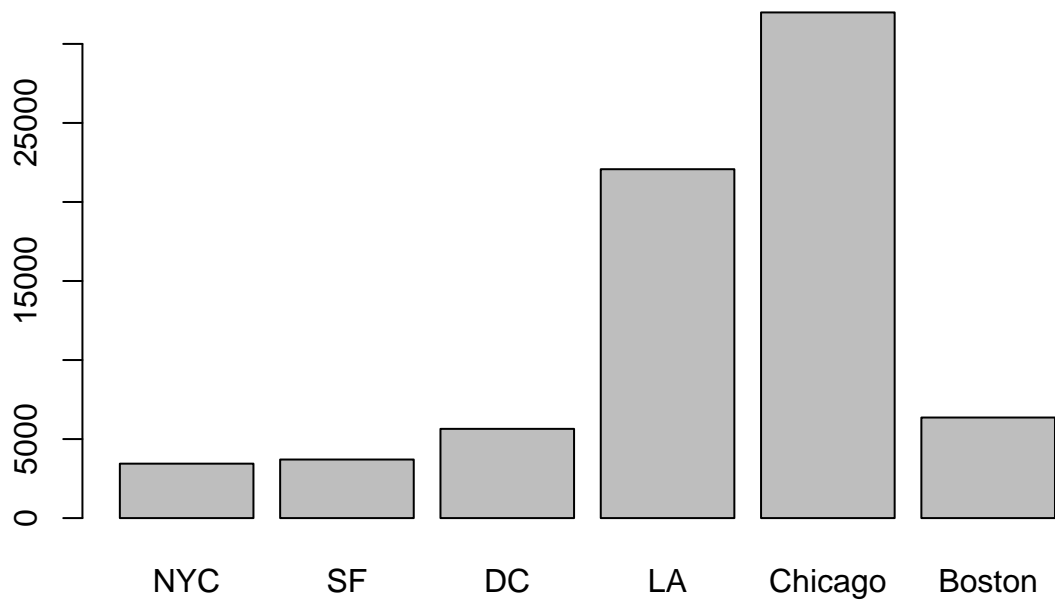


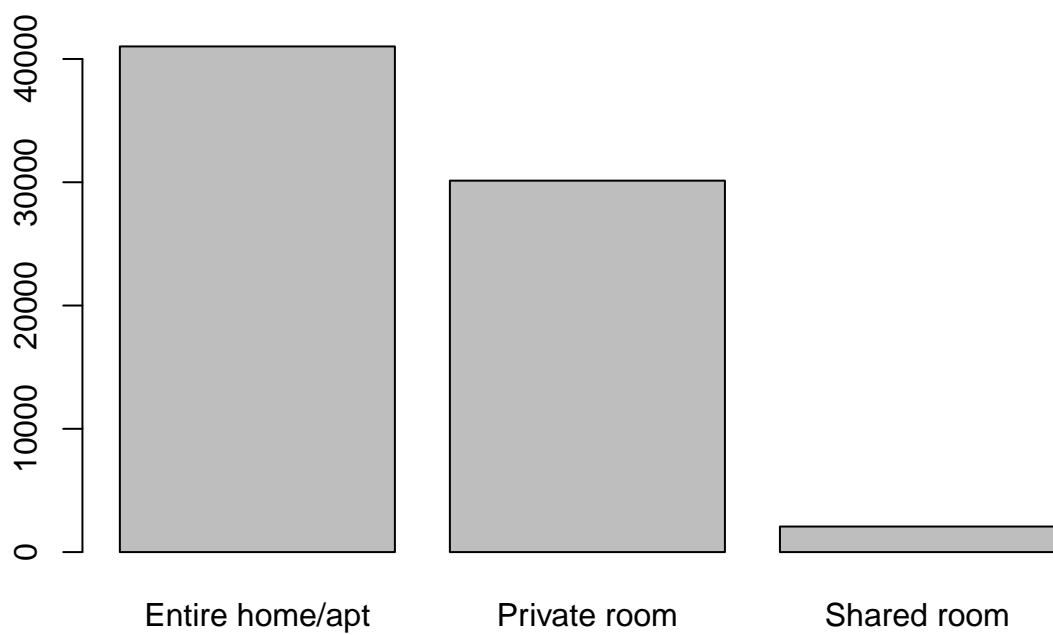
Histogram accomodates



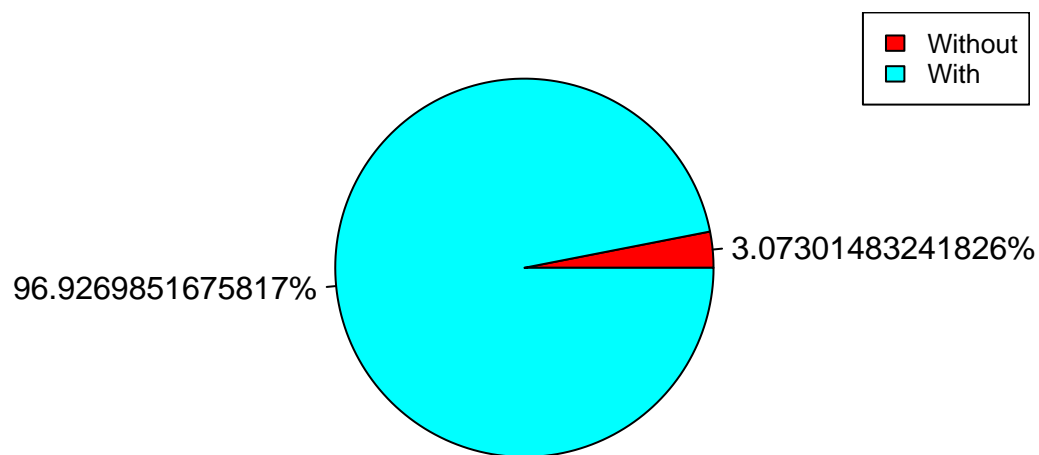
Histogram beds







Wireless



#pitanja:

#ima li neki grad znacajno veci broj outliera?
relative_number_of_outliers

```
## $NYC
## [1] 0.0549217
##
## $SF
## [1] 0.08299277
##
## $DC
## [1] 0.1174491
##
## $LA
## [1] 0.08349946
##
## $Chicago
## [1] 0.07638327
##
## $Boston
## [1] 0.05489399
```

#DC-ima najveći relativni udio outliera
#za nas skup podataka outlieri su samo visoke cijene

```

#jer je razdioba cijena nagnuta na lijevo
#u iducem pitanju se vidi da od svih numerickih varijabli iz
#pocetnog datasea accommodates najkoreliranija sa cijenom objekta
#to svojstvo cemo koristiti i ovdje te cemo t-testom
#pokazati da objekti u DC-u u prosjeku mogu primiti vise ljudi
#u odnosi na prosjek svih objekata u svim gradovima
mean_of_all_cities_accommodates=mean(airbnb_bitno$accommodates)
vector_of_positions=get_DC(airbnb_bitno$city, airbnb_bitno$id)
dc=airbnb_bitno[vector_of_positions,]
dc <- subset(dc, select = -c(id))

dc=get_Dataframe_With_Id_Row(dc)

test=t.test(dc$accommodates, mu=mean_of_all_cities_accommodates, alt="two.sided", conf=0.99)
print(paste("svi gradovi accommodates:", mean_of_all_cities_accommodates, " DC: ", test$estimate))

## [1] "svi gradovi accommodates: 3.16579256466989 DC: 3.54915854738707"

print(paste("mean je povecan za:", ((test$estimate-mean_of_all_cities_accommodates)
                                     /mean_of_all_cities_accommodates), "posto" ))

## [1] "mean je povecan za: 0.121096368408823 posto"

print(test)

##
## One Sample t-test
##
## data: dc$accommodates
## t = 13.178, df = 5644, p-value < 2.2e-16
## alternative hypothesis: true mean is not equal to 3.165793
## 99 percent confidence interval:
## 3.474199 3.624118
## sample estimates:
## mean of x
## 3.549159

#u DC-u 357 outliera su objekti sa bar 5 accommodatesa
#ukupno outliera je 663
#prema gornja_granica=Q3+1.5IQR
outlier_price=435

dc_price_per_accommodates=sort_Numeric_Numeric(dc$price, dc$accommodates)
list_numbers=list(5,6,7,8,9,10,11,12,13,14,15,16)
counter=0
big_apartments=length(list_numbers)
for (i in 1:big_apartments) {
  current=list_numbers[[i]]
  l=length(dc_price_per_accommodates[[current]])
  ll=dc_price_per_accommodates[[current]]

```

```

for (i in 1:l) {
  if (ll[i]>outlier_price) {
    counter=counter+1
  }
}

}

print(paste("broj outliera sa preko 5 accomodates: ",counter))

```

```
## [1] "broj outliera sa preko 5 accomodates: 357"
```

```

#Koja od varijabli je najvise korelirana sa cijenom ugostiteljskog objekta? Ima li to smisla?
airbnb_korelacija <- subset(airbnb_bitno, select = -c(property_type,room_type,cancellation_policy,clean_
airbnb_korelacija[is.na(airbnb_korelacija)] <- 0
airbnb_korelacija <- lapply(airbnb_korelacija, as.numeric)

```

```
## Warning in lapply(airbnb_korelacija, as.numeric): NAs introduced by coercion
```

```
## Warning in lapply(airbnb_korelacija, as.numeric): NAs introduced by coercion
```

```
with(airbnb_korelacija, cor(price, accomodates))
```

```
## [1] 0.5215426
```

```
with(airbnb_korelacija, cor(price, bathrooms))
```

```
## [1] 0.4612756
```

```
with(airbnb_korelacija, cor(price, number_of_reviews))
```

```
## [1] -0.07082064
```

```
with(airbnb_korelacija, cor(price, bedrooms))
```

```
## [1] 0.496533
```

```
with(airbnb_korelacija, cor(price, beds))
```

```
## [1] 0.4352527
```

```

#od numerickih varijabli najkoreliranija je accomodates
#sada ide provjera za amenitese
#uzimamo samo testove koji imaju preko 10000 stupnjeva slobode
list_of_strings_correlation=list()
list_of_tests_enough_degrees_of_freedom=list()
number_of_valid_tests=length(clean_list_of_tests)
counter=0
for (i in 1:number_of_valid_tests) {

```

```

    if(clean_list_of_tests[[i]]$parameter>=10000) {
      temp_string=paste(clean_list_of_tests[[i]]$data.name, clean_list_of_tests[[i]]$conf.int[1],
" ", clean_list_of_tests[[i]]$conf.int[2],": p-value ", clean_list_of_tests[[i]]$p.value )
      counter=counter+1
      list_of_tests_enough_degrees_of_freedom[[counter]]=clean_list_of_tests[[i]]
      list_of_strings_correlation[[counter]]=temp_string
      print(temp_string)
      print("\n")
    }
  }
}

```

```

## [1] "price and Airconditioning -13.8858076274485 -6.63073891384078 : p-value 3.30015436643326e-1
## [1] "\n"
## [1] "price and Familykidfriendly -73.9130410536827 -67.6782811842195 : p-value 0"
## [1] "\n"
## [1] "price and Essentials -14.1823640534319 -3.97287234644289 : p-value 4.66916192237339e-06"
## [1] "\n"
## [1] "price and Hairdryer -24.8664551047106 -18.3594779955358 : p-value 1.69657958458808e-65"
## [1] "\n"
## [1] "price and Iron -26.7452051527984 -20.3407430851092 : p-value 8.78878226108424e-80"
## [1] "\n"
## [1] "price and Washer -55.1002124451845 -49.1620938301594 : p-value 0"
## [1] "\n"
## [1] "price and Dryer -55.9082610873432 -49.9488287751579 : p-value 0"
## [1] "\n"
## [1] "price and Smokedetector -13.2629159723203 -4.13783204733681 : p-value 9.08334895062772e-07"
## [1] "\n"
## [1] "price and Fireextinguisher -27.7113010167365 -21.059820700186 : p-value 2.48050789337543e-7
## [1] "\n"
## [1] "price and Shampoo -18.5745074375982 -11.6556690614951 : p-value 2.39470685770189e-29"
## [1] "\n"
## [1] "price and Hangers -8.81313195529614 -1.84715340653733 : p-value 8.09096642070118e-05"
## [1] "\n"
## [1] "price and TV -69.1696470959746 -63.2910948146106 : p-value 0"
## [1] "\n"
## [1] "price and CableTV -65.7645170753627 -58.4176533407661 : p-value 0"
## [1] "\n"
## [1] "price and Buzzerwirelessintercom -13.114871551265 -5.66530493724311 : p-value 8.49309862608
## [1] "\n"
## [1] "price and Carbonmonoxidedetector -17.2975757133634 -10.6013854156219 : p-value 7.6456180419
## [1] "\n"
## [1] "price and Laptopfriendlyworkspace -22.9085230573131 -16.4419416990809 : p-value 2.836025183
## [1] "\n"
## [1] "price and Internet -20.5973186179269 -14.0670173770801 : p-value 1.68564251671432e-42"
## [1] "\n"
## [1] "price and Indoorfireplace -101.690046449858 -87.4047733983284 : p-value 2.10522556619178e-2
## [1] "\n"
## [1] "price and Firstaidkit -6.82570469799039 -0.0944205949140079 : p-value 0.00809485214894145"
## [1] "\n"
## [1] "price and Freeparkingonpremises -31.2376078829187 -23.5702265616192 : p-value 2.30533030013
## [1] "\n"
## [1] "price and Lockonbedroomdoor 24.8993513829261 32.2280592965084 : p-value 4.07032375992866e-8
## [1] "\n"

```

```
## [1] "price and Petsliveonthisproperty 31.9519711140054 40.002294577991 : p-value 2.8377784372756"
## [1] "\n"
## [1] "price and Petsallowed -38.6679913879877 -27.8749084186549 : p-value 2.97078775305985e-56"
## [1] "\n"
## [1] "price and Safetycard -21.59301936911 -12.2957806666272 : p-value 6.81575550049459e-21"
## [1] "\n"
## [1] "price and 24hourcheckin -26.3254991206186 -19.100352270327 : p-value 9.02956505360116e-59"
## [1] "\n"
## [1] "price and SelfCheckIn -20.1063834301533 -12.1862616651451 : p-value 9.96275625231173e-26"
## [1] "\n"
## [1] "price and Elevator -31.2147272632188 -22.4294628822546 : p-value 2.57928700977233e-55"
## [1] "\n"
```

```
print(paste("broj amenitiesa koji imaju više od 10000 stupnjeva slobode", counter))
```

```
## [1] "broj amenitiesa koji imaju više od 10000 stupnjeva slobode 27"
```

```
print("p-value 0 indicates extremly high correlation")
```

```
## [1] "p-value 0 indicates extremly high correlation"
```

```
print("nas izbor 5 preporuka iznajmljivacima je imati:")
```

```
## [1] "nas izbor 5 preporuka iznajmljivacima je imati:"
```

```
print("TV, CableTV, Freeparkingonpremises, Familykidfriendly, Fireextinguisher")
```

```
## [1] "TV, CableTV, Freeparkingonpremises, Familykidfriendly, Fireextinguisher"
```

```
# ima smisla da je broj osoba koje aparatman moze primiti uskokoreliran sa brojem osoba koje ce
# rezervirati apartman pa tako i sa cijenom
# od amenitiesa je uvijek pozeljno imati tv i kablsku, besplatan parking na posjedu
# kao i biti pirpremljen na dolazak obitelji sa djecom
# vatrogasni aparat je vazan element sigurnosti i indicira na dobro opremljen apartman
```

```
#varijable koje smo odabrali: TV, CableTV, Freeparkingonpremises, Familykidfriendly, Fireextinguisher
#kombinacije varijabli koje predviđaju cijenu:
#kombinacija1:TV, CableTV,Freeparkingonpremises
#kombinacija2:TV, CableTV,Familykidfriendly
#kombinacija3:TV, CableTV,Fireextinguisher
#kombinacija4:TV, Freeparkingonpremises, Familykidfriendly
#kombinacija5:TV, Freeparkingonpremises, Fireextinguisher
#kombinacija6:TV, Familykidfriendly, Fireextinguisher
```

```

#kombinacija7:CableTV,Freeparkingonpremises, Familykidfriendly
#kombinacija8:CableTV,Freeparkingonpremises, Fireextinguisher
#kombinacija9:CableTV,Familykidfriendly, Fireextinguisher
#kombinacija10: Freeparkingonpremises, Familykidfriendly, Fireextinguisher

tv_vector=airbnb_bitno["TV"]
cabletv_vector=airbnb_bitno["CableTV"]
freeparkingonpremises_vector=airbnb_bitno["Freeparkingonpremises"]
familykidfriendly_vector=airbnb_bitno["Familykidfriendly"]
fireextinguisher_vector=airbnb_bitno["Fireextinguisher"]

comb1=c(tv_vector&cabletv_vector&freeparkingonpremises_vector)
comb2=c(tv_vector&cabletv_vector&familykidfriendly_vector)
comb3=c(tv_vector&cabletv_vector&fireextinguisher_vector)

comb4=c(tv_vector&freeparkingonpremises_vector&familykidfriendly_vector)
comb5=c(tv_vector&freeparkingonpremises_vector&fireextinguisher_vector)

comb6=c(tv_vector&familykidfriendly_vector&fireextinguisher_vector)

comb7=c(cabletv_vector&freeparkingonpremises_vector&familykidfriendly_vector)
comb8=c(cabletv_vector&freeparkingonpremises_vector&fireextinguisher_vector)
comb9=c(cabletv_vector&familykidfriendly_vector&fireextinguisher_vector)

comb10=c(freeparkingonpremises_vector&familykidfriendly_vector&fireextinguisher_vector)

airbnb_bitno <- cbind(airbnb_bitno , comb1, comb2, comb3, comb4, comb5,
                      comb6, comb7, comb8, comb9, comb10)

list_of_combinations=list(comb1, comb2, comb3, comb4, comb5,
                          comb6, comb7, comb8, comb9, comb10)
list_of_combinations_tests=list()

for(i in 1:10) {
  test=t.test(airbnb_bitno$price~list_of_combinations[[i]], mu=0, alt="two.sided", conf=0.99)
  print(paste("comb",i, test$conf.int[1], test$conf.int[2], "p-value:",test$p.value ))
}

```

```

## [1] "comb 1 -73.5857519457325 -60.64939203991 p-value: 3.53671415775073e-152"
## [1] "comb 2 -94.4398999507401 -84.5925313939519 p-value: 0"
## [1] "comb 3 -66.6324258782835 -56.2782805915725 p-value: 5.17406591765725e-199"
## [1] "comb 4 -79.5155669601984 -68.4067384660601 p-value: 3.85088183886776e-248"
## [1] "comb 5 -58.7793161462146 -47.0641724947939 p-value: 3.15188692827505e-117"
## [1] "comb 6 -82.5500433130342 -72.7644322068185 p-value: 0"
## [1] "comb 7 -101.059367616042 -84.9641981406729 p-value: 2.16520470263699e-183"
## [1] "comb 8 -81.3691967716243 -64.6780031139225 p-value: 4.99004306303371e-108"
## [1] "comb 9 -95.6496660051461 -82.2084353494337 p-value: 3.32601447147636e-239"
## [1] "comb 10 -76.0820013337378 -62.4131291415186 p-value: 7.45437086010524e-145"

```

```

#odabiremo kombinaciju 2 i 6 jer imaju najmanju p vrijednost
#medutim, kako je u kombinaciji 2 Tv i CableTV koji su izuzetno korelirani
#kombinacija koja nabolje predviđa cijenu po nama je
#kombinacija6:TV, Familykidfriendly, Fireextinguisher

```

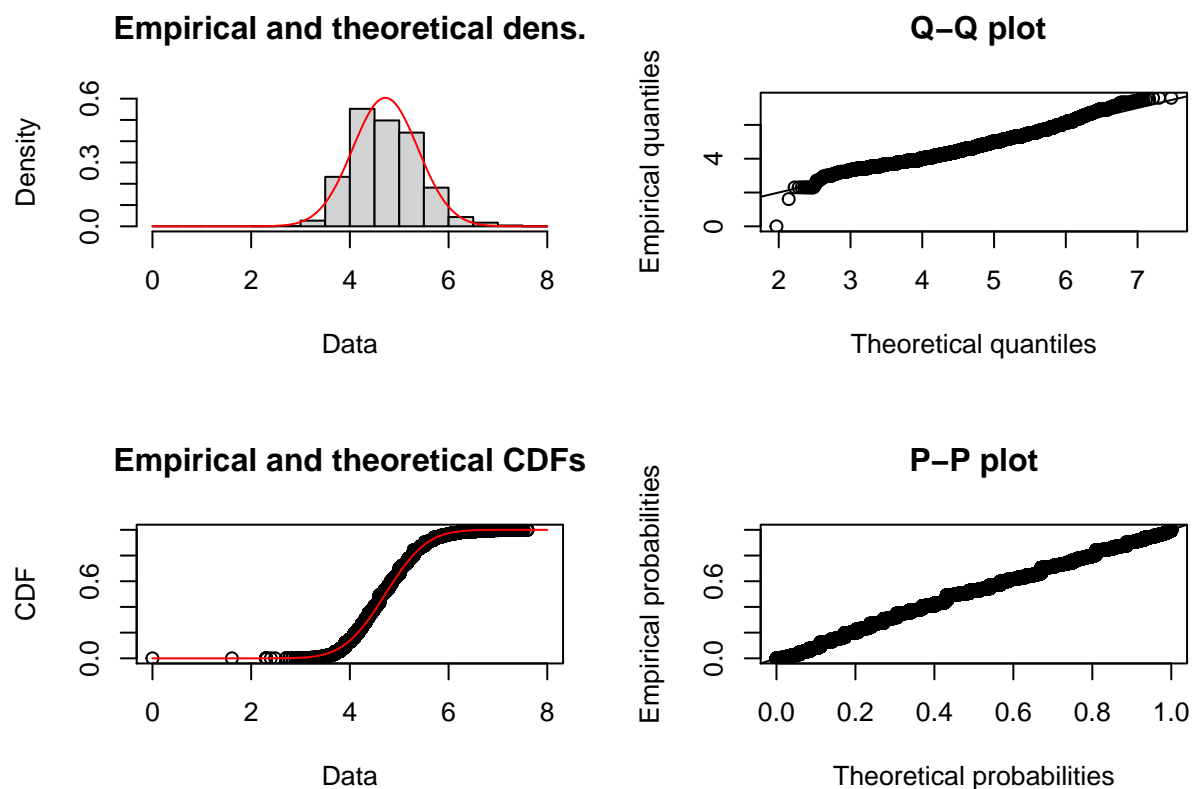


```
#Mozemo li pretpostaviti nekakvu distribuciju nad cijenama objekata za pojedine gradove?
#mozemo, noramlnu distrubuciju nad logaritmiranim cijenama, evo i grafova po gradovima
library(fitdistrplus)
```

```
## Loading required package: MASS
```

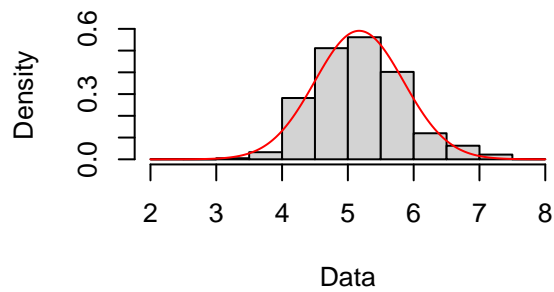
```
## Loading required package: survival
```

```
#NYC
normal_dist <- fitdistr(log(List_of_prices_per_city[[1]]), "norm")
plot(normal_dist)
```

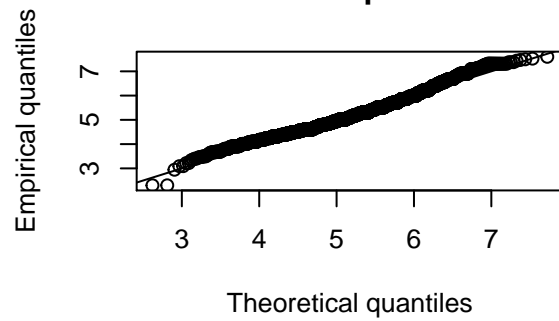


```
#SF
normal_dist<- fitdistr(log(List_of_prices_per_city[[2]]), "norm")
plot(normal_dist)
```

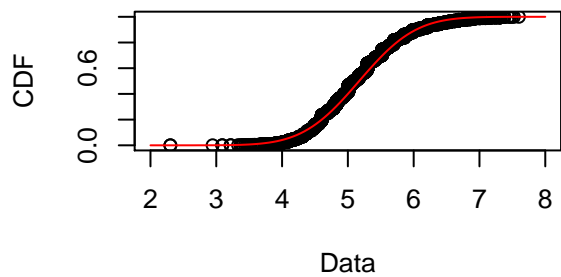
Empirical and theoretical dens.



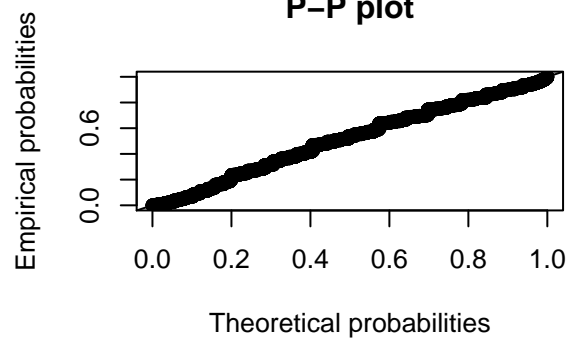
Q-Q plot



Empirical and theoretical CDFs

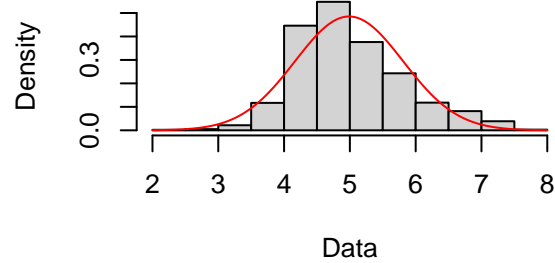


P-P plot

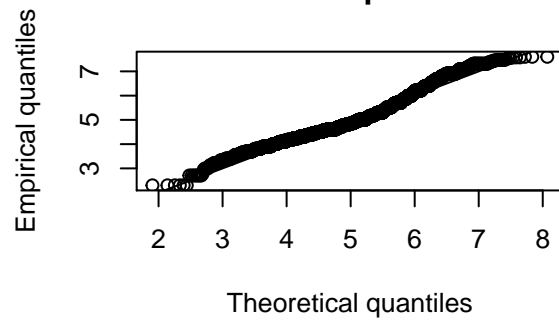


```
#DC
normal_dist <- fitdist(log(List_of_prices_per_city[[3]]), "norm")
plot(normal_dist)
```

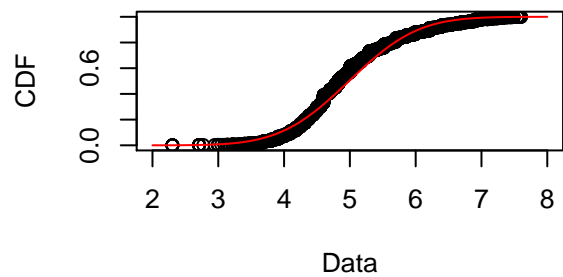
Empirical and theoretical dens.



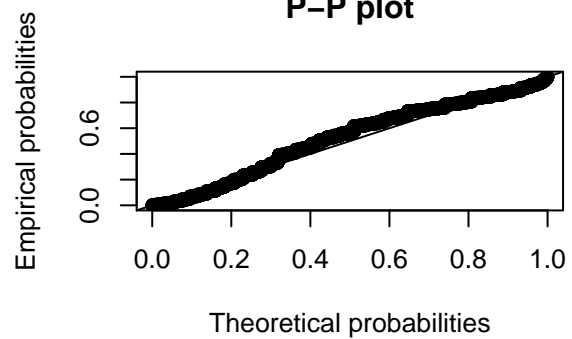
Q-Q plot



Empirical and theoretical CDFs

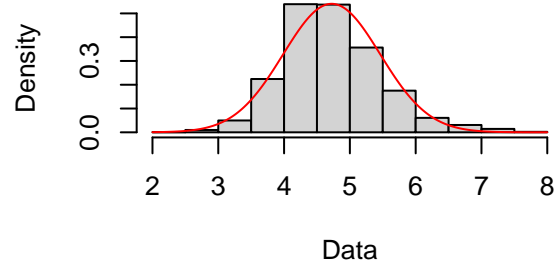


P-P plot

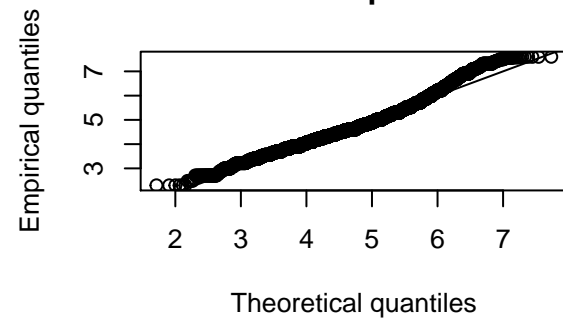


```
#LA
normal_dist <- fitdist(log(List_of_prices_per_city[[4]]), "norm")
plot(normal_dist)
```

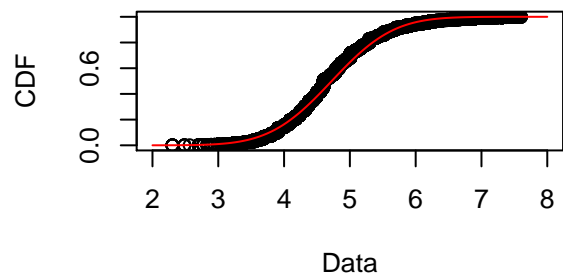
Empirical and theoretical dens.



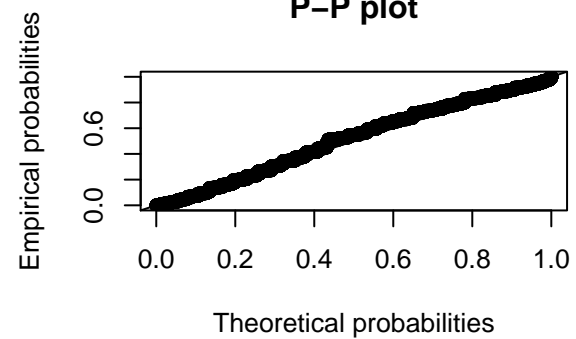
Q-Q plot



Empirical and theoretical CDFs

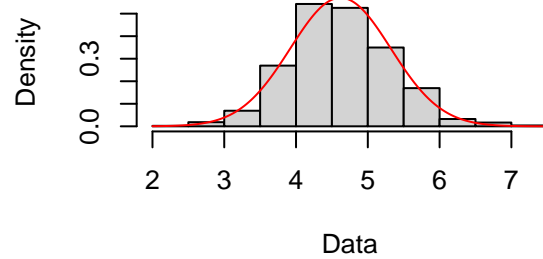


P-P plot

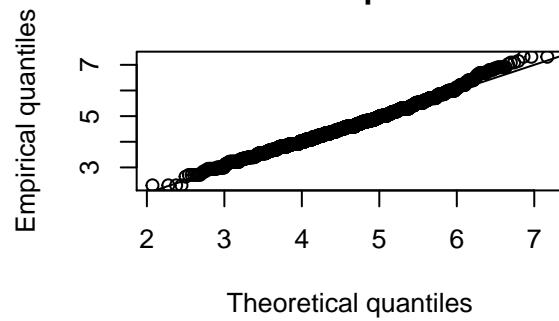


```
#Chicago  
normal_dist <- fitdist(log(List_of_prices_per_city[[5]]), "norm")  
plot(normal_dist)
```

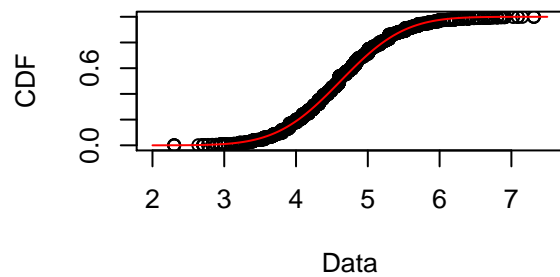
Empirical and theoretical dens.



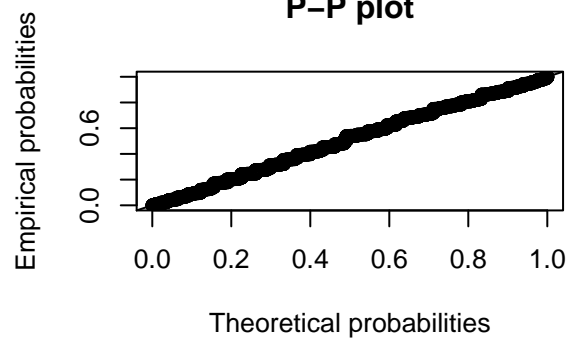
Q-Q plot



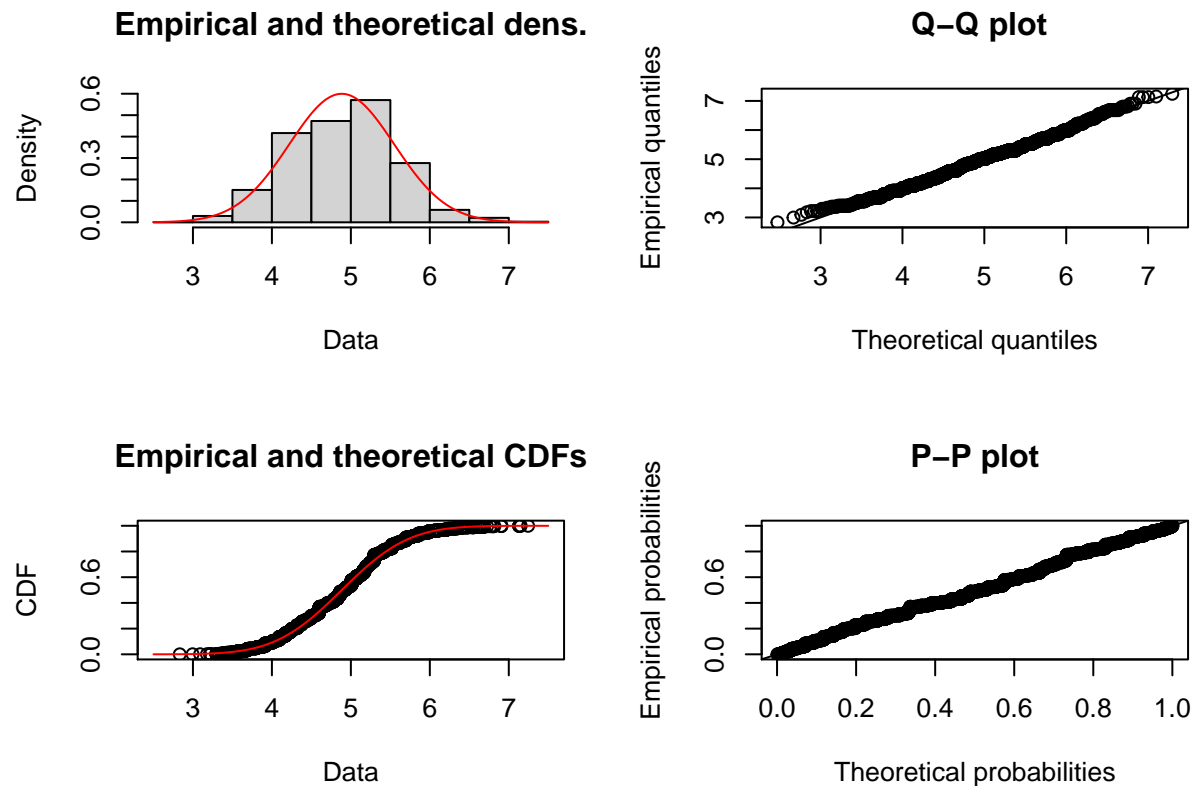
Empirical and theoretical CDFs



P-P plot



```
#Boston
normal_dist <- fitdist(log(List_of_prices_per_city[[6]]), "norm")
plot(normal_dist)
```



#Ima li neki grad statisticki znacajno vece cijene ugostiteljskih objekata nego neki drugi grad?

```
#uporedba cijena apartmana po svakom gradu sa meanom cijena svih apartmana u dataframeu
mean_of_all_cities=mean(airbnb_bitno$price)
print(paste("mean cijena apartmana u svim stanovima je", mean_of_all_cities))
```

```
## [1] "mean cijena apartmana u svim stanovima je 160.386721844355"
```

```
city_names=unique(airbnb_bitno$city)

for (i in 1:6) {
  city_prices_tests=list()
  test=t.test(List_of_prices_per_city[[i]], mu=mean_of_all_cities, alt="two.sided", conf=0.99)
  test$data.name=paste(city_names[i], "mean price")
  print(test)
}
```

```
##
## One Sample t-test
##
## data: NYC mean price
## t = -24.369, df = 31990, p-value < 2.2e-16
```

```

## alternative hypothesis: true mean is not equal to 160.3867
## 99 percent confidence interval:
## 140.8967 144.6233
## sample estimates:
## mean of x
## 142.76
##
##
## One Sample t-test
##
## data: SF mean price
## t = 26.132, df = 6361, p-value < 2.2e-16
## alternative hypothesis: true mean is not equal to 160.3867
## 99 percent confidence interval:
## 220.6494 233.8329
## sample estimates:
## mean of x
## 227.2411
##
##
## One Sample t-test
##
## data: DC mean price
## t = 16.876, df = 5644, p-value < 2.2e-16
## alternative hypothesis: true mean is not equal to 160.3867
## 99 percent confidence interval:
## 209.4978 227.1973
## sample estimates:
## mean of x
## 218.3476
##
##
## One Sample t-test
##
## data: LA mean price
## t = -3.8639, df = 22071, p-value = 0.0001119
## alternative hypothesis: true mean is not equal to 160.3867
## 99 percent confidence interval:
## 152.5778 158.8251
## sample estimates:
## mean of x
## 155.7015
##
##
## One Sample t-test
##
## data: Chicago mean price
## t = -13.658, df = 3704, p-value < 2.2e-16
## alternative hypothesis: true mean is not equal to 160.3867
## 99 percent confidence interval:
## 127.1133 137.6770
## sample estimates:
## mean of x
## 132.3951

```

```
##
##
## One Sample t-test
##
## data: Boston mean price
## t = 2.4413, df = 3442, p-value = 0.01468
## alternative hypothesis: true mean is not equal to 160.3867
## 99 percent confidence interval:
## 160.0876 171.4332
## sample estimates:
## mean of x
## 165.7604
```

```
print("iz navedenih cijena vidimo da San Francisco ima znacajno veće cijene od prosjeka")
```

```
## [1] "iz navedenih cijena vidimo da San Francisco ima znacajno veće cijene od prosjeka"
```

```
#na interentu smo pronasli podatak da San Francisco je drugi najgusce naseljeni grad u SAD-u
#sto znacajno utjece na cijenu apartmana, kao i podatak da je izuzetno turisticki posjecen
#sto takoder dize cijenu apartmana
```

```
#Utjece li znacajno naknada za ciscenje ugostiteljskog objekta na prosjecnu cijenu objekta?
cleaning_fee_test=t.test(airbnb_bitno$price~airbnb_bitno$cleaning_fee, mu=0, alt="two.sided", conf=0.99)
print(cleaning_fee_test)
```

```
##
## Welch Two Sample t-test
##
## data: airbnb_bitno$price by airbnb_bitno$cleaning_fee
## t = -7.2775, df = 28682, p-value = 3.489e-13
## alternative hypothesis: true difference in means is not equal to 0
## 99 percent confidence interval:
## -15.326967 -7.313089
## sample estimates:
## mean in group False mean in group True
## 152.0291 163.3491
```

```
#utjece, ali ne u tolikoj mjeri kao neke druge "vaznije varijable" navedene gore
#takoder iako je p-vrijednost mala, nije mala kao u vec prije navedenim varijablama
```

```
#Koje biste mjere preporucili iznajmljivacima kuca u NYC kako bi potencijalno mogli
#povecati cijenu nocenja (pod pretpostavkom da svi iznajmljivaci imaju jednaku popunjenost)?
#ocekujemo isti zakljucak kao i kada smo odabrali 5 varijabli
#za koje smo vidjeli veoma malu p-vrijednost
#a to su: TV, CableTV, Freeparkingonpremises, Familykidfriendly, Fireextinguisher
#kod je isti kao i za sve stanove
```



```

#samo trazimo da test ima bar 5000 stupnjeva slobode umjesto dosadasnjih 10000

vector_of_positions=get_positions_of_nyc_houses(airbnb_bitno$property_type, airbnb_bitno$id)
nyc=airbnb_bitno[vector_of_positions,]
nyc <- subset(nyc, select = -c(id))

nyc=get_Dataframe_With_Id_Row(nyc)

#iz nekog razloga petlja puca za i= 116, 118, 120 i 121 pa smo skratili petlju do 115
#u vektoru number_of_apartments_that_have_certain_amenitie_vector[115:129]
#se vidi da ti amenitisi gotovo pa uopce nisu bitni
list_of_nyc_tests=list()
position_of_first_amenitie=15
for (i in 1:115) {

  test=t.test(nyc$price~nyc[,position_of_first_amenitie+i], mu=0, alt="two.sided", conf=0.99)
  test$data.name=paste("nyc price and ", vector_of_unique_amenities[i])
  list_of_nyc_tests[[i]]=test

}
clean_list_of_nyc_tests=list()
number_of_tests=length(list_of_nyc_tests)
counter=0
for (i in 1:number_of_tests) {
  if(list_of_nyc_tests[[i]]$data.name!="nyc price and translationmissingenhostingamenity50" &&
    list_of_nyc_tests[[i]]$data.name!="nyc price and translationmissingenhostingamenity49") {
    counter=counter+1
    clean_list_of_nyc_tests[[counter]]=list_of_nyc_tests[[i]]
  }
}

list_of_strings_correlation_nyc=list()
list_of_tests_enough_degrees_of_freedom_nyc=list()
number_of_valid_tests_nyc=length(clean_list_of_nyc_tests)
counter=0
for (i in 1:number_of_valid_tests_nyc) {
  if(clean_list_of_nyc_tests[[i]]$parameter>=5000) {
    temp_string=paste(clean_list_of_nyc_tests[[i]]$data.name, clean_list_of_nyc_tests[[i]]$conf.int[1],
    counter=counter+1
    list_of_tests_enough_degrees_of_freedom_nyc[[counter]]=clean_list_of_nyc_tests[[i]]
    list_of_strings_correlation_nyc[[counter]]=temp_string
    print(temp_string)
    print("\n")
  }
}

```

```

## [1] "nyc price and Airconditioning -26.5783790281828 -6.63073891384078 : p-value 2.34240317858064e-
## [1] "\n"
## [1] "nyc price and Kitchen -107.199314144537 -37.3192072675765 : p-value 1.89777975836664e-193"
## [1] "\n"

```

```

## [1] "nyc price and Familykidfriendly -125.727630005561 -67.6782811842195 : p-value 7.55451649881953
## [1] "\n"
## [1] "nyc price and Hairdryer -35.4352715910092 -18.3594779955358 : p-value 9.11770631474748e-12"
## [1] "\n"
## [1] "nyc price and Iron -44.2550342557583 -20.3407430851092 : p-value 3.30185535208452e-21"
## [1] "\n"
## [1] "nyc price and Washer -114.601209923993 -49.1620938301594 : p-value 9.03193333156248e-258"
## [1] "\n"
## [1] "nyc price and Dryer -114.535388846736 -49.9488287751579 : p-value 2.34003507517581e-252"
## [1] "\n"
## [1] "nyc price and Fireextinguisher -33.5220480268326 -21.059820700186 : p-value 4.84231999880826e-
## [1] "\n"
## [1] "nyc price and Shampoo -19.1343366840287 -11.6556690614951 : p-value 0.0373133076457561"
## [1] "\n"
## [1] "nyc price and Hangers -13.830018380501 -1.84715340653733 : p-value 0.428556142291734"
## [1] "\n"
## [1] "nyc price and TV -114.355849776002 -63.2910948146106 : p-value 3.17872222937546e-230"
## [1] "\n"
## [1] "nyc price and CableTV -102.224960701425 -58.4176533407661 : p-value 9.33319632320794e-119"
## [1] "\n"
## [1] "nyc price and Carbonmonoxidedetector -27.3685475631264 -10.6013854156219 : p-value 1.361484935
## [1] "\n"
## [1] "nyc price and Laptopfriendlyworkspace -43.6462049687999 -16.4419416990809 : p-value 1.02973175
## [1] "\n"
## [1] "nyc price and Internet -46.6532493634283 -14.0670173770801 : p-value 1.20970011214251e-22"
## [1] "\n"
## [1] "nyc price and Indoorfireplace -145.555467941945 -87.4047733983284 : p-value 7.87854515821582e-
## [1] "\n"
## [1] "nyc price and Firstaidkit -7.03530703383842 -0.0944205949140079 : p-value 0.517313633485835"
## [1] "\n"
## [1] "nyc price and Freeparkingonpremises -76.2823777754751 -23.5702265616192 : p-value 8.4658512409
## [1] "\n"
## [1] "nyc price and Lockonbedroomdoor 48.8058365397684 32.2280592965084 : p-value 1.51093282848303e-5
## [1] "\n"
## [1] "nyc price and Petsliveonthisproperty 52.5753570099268 40.002294577991 : p-value 3.348748633761
## [1] "\n"
## [1] "nyc price and 24hourcheckin -41.5979315192266 -19.100352270327 : p-value 1.49684861834662e-15"
## [1] "\n"

```

*#zaključak je da: TV, CableTV, Freeparkingonpremises, Familykidfriendly, Fireextinguisher
#su stvari koje ce iznajmljivacima kuca u nyc povecati najam*

““