

A Site Selection and Recruitment Strategy of JPTA Program Evaluation: Data Analysis

DS4SI Assignment 2

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```
# Import the JPTA dataset
jpta <- read.csv("jpta.csv", header = TRUE)
N <- dim(jpta)[1]
## Step 1:
# Create 4 subsets and group the dataset by regions
jpta_1 <- subset(jpta, jpta$region == 1) # Northeast
jpta_2 <- subset(jpta, jpta$region == 2) # North Central
jpta_3 <- subset(jpta, jpta$region == 3) # South
jpta_4 <- subset(jpta, jpta$region == 4) # West

# Subset by urban (1 for urban area and 0 for rural area)
# Sort the comfort variable in descending order
jpta_1_1 <- jpta_1 %>% # Northeast Urban
  subset(jpta_1$urban == 1) %>%
  arrange(desc(comfort))

jpta_1_0 <- jpta_1 %>% # Northeast Rural
  subset(jpta_1$urban == 0) %>%
  arrange(desc(comfort))

jpta_2_1 <- jpta_2 %>% # North Central Urban
  subset(jpta_2$urban == 1) %>%
  arrange(desc(comfort))

jpta_2_0 <- jpta_2 %>% # North Central Rural
  subset(jpta_2$urban == 0) %>%
  arrange(desc(comfort))

jpta_3_1 <- jpta_3 %>% # South Urban
  subset(jpta_3$urban == 1) %>%
  arrange(desc(comfort))

jpta_3_0 <- jpta_3 %>% # South Rural
  subset(jpta_3$urban == 0) %>%
  arrange(desc(comfort))

jpta_4_1 <- jpta_4 %>% # West Urban
  subset(jpta_4$urban == 1) %>%
  arrange(desc(comfort))

jpta_4_0 <- jpta_4 %>% # West Rural
  subset(jpta_4$urban == 0) %>%
```

```
arrange(desc(comfort))
```

```
## Step 2: Determine the weight of each subset
```

```
w_1_0 <- dim(jpta_1_0)[1]/N  
w_1_1 <- dim(jpta_1_1)[1]/N  
w_2_0 <- dim(jpta_2_0)[1]/N  
w_2_1 <- dim(jpta_2_1)[1]/N  
w_3_0 <- dim(jpta_3_0)[1]/N  
w_3_1 <- dim(jpta_3_1)[1]/N  
w_4_0 <- dim(jpta_4_0)[1]/N  
w_4_1 <- dim(jpta_4_1)[1]/N
```

```
## Step 3: Create variables with empty vectors to store results
```

```
total_comfort <- rep(0, times = N) # The sum of comfort  
per_comfort <- rep(0, times = N) # The average comfort  
total_cost <- rep(0, times = N) # The sum of cost
```

```
# Calculate the sum of comfort, the average comfort and the sum of cost when  
# n sites are selected (1 <= n <= 400)
```

```
for (n in 1:400) {  
  comfort_1_0 <- sum(jpta_1_0[1:round(n * w_1_0)], ]$comfort)  
  comfort_1_1 <- sum(jpta_1_1[1:round(n * w_1_1)], ]$comfort)  
  comfort_2_0 <- sum(jpta_2_0[1:round(n * w_2_0)], ]$comfort)  
  comfort_2_1 <- sum(jpta_2_1[1:round(n * w_2_1)], ]$comfort)  
  comfort_3_0 <- sum(jpta_3_0[1:round(n * w_3_0)], ]$comfort)  
  comfort_3_1 <- sum(jpta_3_1[1:round(n * w_3_1)], ]$comfort)  
  comfort_4_0 <- sum(jpta_4_0[1:round(n * w_4_0)], ]$comfort)  
  comfort_4_1 <- sum(jpta_4_1[1:round(n * w_4_1)], ]$comfort)  
  
  total_comfort[n] <- sum(comfort_1_0, comfort_1_1, comfort_2_0, comfort_2_1,  
    comfort_3_0, comfort_3_1, comfort_4_0, comfort_4_1)  
  
  per_comfort[n] <- total_comfort[n]/n  
  
  cost_1_0 <- sum((jpta_1_0[1:round(n * w_1_0)], ]$distance) * 500)  
  cost_1_1 <- sum((jpta_1_1[1:round(n * w_1_1)], ]$distance) * 500)  
  cost_2_0 <- sum((jpta_2_0[1:round(n * w_2_0)], ]$distance) * 500)  
  cost_2_1 <- sum((jpta_2_1[1:round(n * w_2_1)], ]$distance) * 500)  
  cost_3_0 <- sum((jpta_3_0[1:round(n * w_3_0)], ]$distance) * 500)  
  cost_3_1 <- sum((jpta_3_1[1:round(n * w_3_1)], ]$distance) * 500)  
  cost_4_0 <- sum((jpta_4_0[1:round(n * w_4_0)], ]$distance) * 500)  
  cost_4_1 <- sum((jpta_4_1[1:round(n * w_4_1)], ]$distance) * 500)  
  
  total_cost[n] <- sum(cost_1_0, cost_1_1, cost_2_0, cost_2_1, cost_3_0, cost_3_1,  
    cost_4_0, cost_4_1) + 1000 * n  
}
```

```
## Step 4: Determine at least how many sites should we select so that the sum  
## of comfort can reach 30
```

```
N_30 <- 400 - sum(total_comfort >= 30) + 1  
# Determine the max of the average comfort from 41 to 400  
N_ave_max <- N_30 - 1 + which.max(per_comfort[41:400])
```

```
## Step 5: Creates plots to visualize the result Total comfort
```

```
plot(total_comfort, type = "l", lty = 1, lwd = 1.5, col = "black", ann = FALSE,
```

```

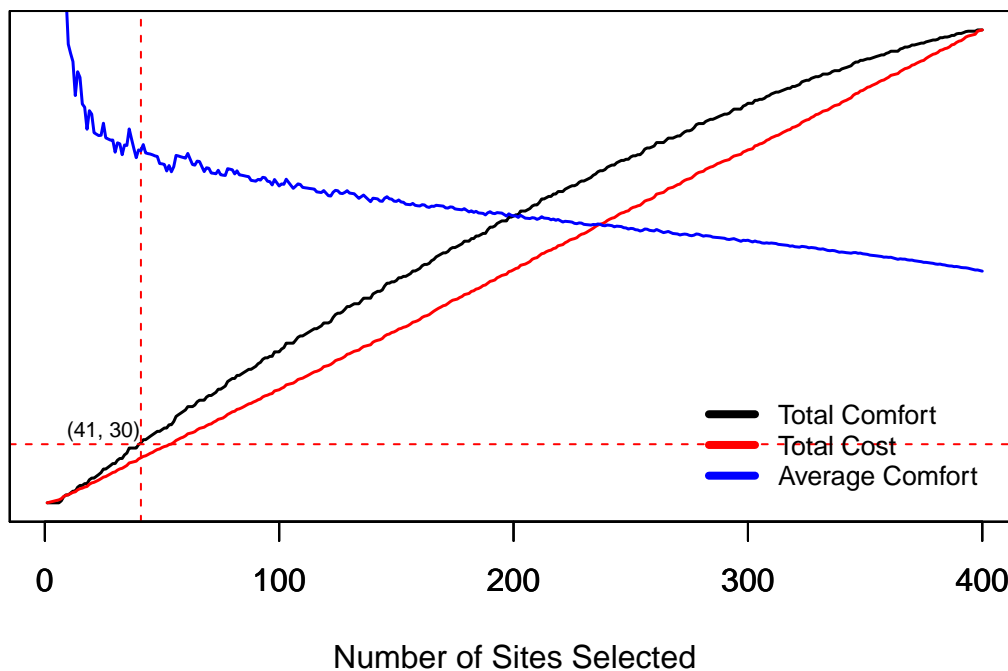
    yaxt = "n")
# Add line to show when the sum of comfort reaches 30
abline(h = 30, lty = 2, lwd = 1, col = 2)
# Add line to show when the number of sites selected is equal to 41
abline(v = 41, lty = 2, lwd = 1, col = 2)
text(x = 25, y = 35, "(41, 30)", cex = 0.7)

par(new = TRUE)
# Total cost
plot(total_cost, type = "l", lty = 1, lwd = 1.5, col = "red", ann = FALSE, yaxt = "n")

par(new = TRUE)
# Average comfort
plot(per_comfort, type = "l", lty = 1, lwd = 1.5, col = "blue", ann = FALSE,
     yaxt = "n", ylim = c(0, 1))
title(main = "The total comfort, average comfort and total cost when selecting N sites",
      xlab = "Number of Sites Selected", cex.main = 0.9)
legend("bottomright", inset = 0.03, legend = c("Total Comfort", "Total Cost",
      "Average Comfort"), col = c("black", "red", "blue"), lty = c(1, 1, 1), lwd = c(4,
      4, 4), cex = 0.8, box.lty = 0)

```

The total comfort, average comfort and total cost when selecting N sites



```

## Step 6: Determine which 41 sites should we select
N <- N_30
sample_1_0 <- jpta_1_0[1:round(N * w_1_0), ]
sample_1_1 <- jpta_1_1[1:round(N * w_1_1), ]
sample_2_0 <- jpta_2_0[1:round(N * w_2_0), ]
sample_2_1 <- jpta_2_1[1:round(N * w_2_1), ]
sample_3_0 <- jpta_3_0[1:round(N * w_3_0), ]
sample_3_1 <- jpta_3_1[1:round(N * w_3_1), ]

```

```

sample_4_0 <- jpta_4_0[1:round(N * w_4_0), ]
sample_4_1 <- jpta_4_1[1:round(N * w_4_1), ]

sample <- rbind(sample_1_0, sample_1_1, sample_2_0, sample_2_1, sample_3_0,
  sample_3_1, sample_4_0, sample_4_1) %>% arrange(site_id)

# Calculate the total cost
Cost <- 1000 * N + sum(sample$distance) * 500

# The total number of sites is:
dim(sample)[1]

## [1] 41

# The total cost of selecting 41 sites is:
paste("$", round(Cost, digits = 2))

## [1] "$ 84337.63"

# The list of sample sites:
data.frame(sample$site_id, sample$region, sample$distance, sample$comfort)

##   sample.site_id sample.region sample.distance sample.comfort
## 1           1085             4       2.5534772       0.6793509
## 2           1374             2       2.2719130       0.7871178
## 3           1802             3       2.7949002       0.7906509
## 4           1895             2       2.7248187       0.6823765
## 5           2082             4       2.0700336       0.6944425
## 6           2096             4       0.9828134       0.7505317
## 7           2708             1       0.8369433       0.7144757
## 8           2816             1       1.3842986       0.7654412
## 9           2901             3       2.7871634       0.7672680
## 10          2951             1       1.8131748       0.7123354
## 11          3075             2       0.6239017       0.6793425
## 12          3232             1       2.1004726       0.9328601
## 13          3396             4       1.0863968       0.7427873
## 14          3860             4       1.8536565       0.7121777
## 15          4118             1       1.9800302       0.6688466
## 16          4429             3       2.8258375       0.8211312
## 17          4540             1       0.8029432       0.7601828
## 18          4807             4       1.9432895       0.7542857
## 19          5091             4       2.7301329       0.6775037
## 20          5095             4       2.3858727       0.7045969
## 21          5263             1       1.5362527       0.7261004
## 22          5336             4       1.9907189       0.7801155
## 23          5454             3       2.3391063       0.7361615
## 24          5514             2       2.2454205       0.8088736
## 25          5545             2       2.2046690       0.6606413
## 26          5767             1       1.9574461       0.6942057
## 27          5907             2       1.4495354       0.6498019
## 28          5948             3       2.8606709       0.7119592
## 29          5977             1       2.3926313       0.7056639
## 30          6533             3       3.7435490       0.8374992
## 31          6759             1       1.9093724       0.7440476
## 32          7154             4       1.9727270       0.9197648
## 33          7440             3       3.3886186       0.8198539

```

## 34	8164	3	3.1234164	0.9402033
## 35	8648	1	2.4795256	0.6859688
## 36	8744	4	2.0059626	0.7154524
## 37	9446	2	1.7756025	0.7450564
## 38	9582	3	1.8655949	0.7176159
## 39	9670	3	2.9613396	0.7220163
## 40	9681	2	2.0548630	0.6714709
## 41	9714	4	1.8661634	0.7314373