Conjoint Analysis

Neha 2/2/2017

Introduction

This is a traditional conjoint analysis R Markdown document. The attributes in this demonstartion are: mobile provider, startup and monthly costs, if the provider offers 4G services in the area, whether the retailer has a retail location nearby also whether the provider supports Apple, Samsung or Nexus phones.

```
print.digits <- 2 # set number of digits on print and spine chart
library(support.CEs) # package for survey construction
## Loading required package: DoE.base
## Warning: package 'DoE.base' was built under R version 3.2.5
## Loading required package: grid
## Loading required package: conf.design
```

Attaching package: 'DoE.base' PRO version Are you a developer? Try out the HTML to PDF API

```
## The following objects are masked from 'package:stats':
##
       aov, 1m
##
## The following object is masked from 'package:graphics':
##
       plot.design
##
## The following object is masked from 'package:base':
##
       lengths
##
## Loading required package: MASS
## Loading required package: simex
## Loading required package: RCurl
## Warning: package 'RCurl' was built under R version 3.2.4
## Loading required package: bitops
## Loading required package: XML
```

Warning: package 'XML' was built under R version 3.2.5

```
# generate a balanced set of product profiles for survey
provider.survey <- Lma.design(attribute.names =</pre>
  list(brand = c("AT&T", "T-Mobile", "US Cellular", "Verizon"),
  startup = c("$100", "$200", "$300", "$400"),
 monthly = c("$100", "$200", "$300", "$400"),
  service = c("4G NO", "4G YES"),
  retail = c("Retail NO", "Retail YES"),
  apple = c("Apple NO", "Apple YES"),
  samsung = c("Samsung NO", "Samsung YES"),
  google = c("Nexus NO", "Nexus YES")), nalternatives = 1, nblocks=1, seed=9999)
```

```
## The columns of the array have been used in order of appearance.
## For designs with relatively few columns,
## the properties can sometimes be substantially improved
## using option columns with min3 or even min34.
```

print(questionnaire(provider.survey)) # print survey design for review

```
##
## Block 1
##
## Question 1
##
           alt.1
## brand "AT&T"
## startup "$100"
## monthly "$100"
```

```
## service "4G NO"
## retail "Retail NO"
## apple "Apple NO"
## samsung "Samsung NO"
## google "Nexus NO"
##
## Question 2
##
          alt.1
## brand "Verizon"
## startup "$300"
## monthly "$100"
## service "4G NO"
## retail "Retail YES"
## apple "Apple YES"
## samsung "Samsung YES"
## google "Nexus NO"
##
## Question 3
##
          alt.1
## brand "US Cellular"
## startup "$400"
## monthly "$200"
## service "4G NO"
## retail "Retail NO"
## apple "Apple NO"
## samsung "Samsung YES"
## google "Nexus NO"
##
## Question 4
          alt.1
##
## brand "Verizon"
```

```
## startup "$400"
## monthly "$400"
## service "4G YES"
## retail "Retail YES"
## apple "Apple NO"
## samsung "Samsung NO"
## google "Nexus NO"
##
## Question 5
          alt.1
##
         "Verizon"
## brand
## startup "$200"
## monthly "$300"
## service "4G NO"
## retail "Retail NO"
## apple
         "Apple NO"
## samsung "Samsung YES"
## google "Nexus YES"
##
## Question 6
          alt.1
##
## brand "Verizon"
## startup "$100"
## monthly "$200"
## service "4G YES"
## retail "Retail NO"
## apple "Apple YES"
## samsung "Samsung NO"
## google "Nexus YES"
##
## Question 7
```

```
##
          alt.1
## brand "US Cellular"
## startup "$300"
## monthly "$300"
## service "4G YES"
## retail "Retail NO"
## apple "Apple YES"
## samsung "Samsung NO"
## google "Nexus NO"
##
## Question 8
##
           alt.1
## brand
         "AT&T"
## startup "$400"
## monthly "$300"
## service "4G NO"
## retail "Retail YES"
## apple "Apple YES"
## samsung "Samsung NO"
## google "Nexus YES"
##
## Question 9
##
           alt.1
## brand "AT&T"
## startup "$200"
## monthly "$400"
## service "4G YES"
## retail "Retail NO"
## apple "Apple YES"
## samsung "Samsung YES"
## google "Nexus NO"
```

```
##
## Question 10
##
          alt.1
## brand "T-Mobile"
## startup "$400"
## monthly "$100"
## service "4G YES"
## retail "Retail NO"
## apple "Apple YES"
## samsung "Samsung YES"
## google "Nexus YES"
##
## Question 11
          alt.1
##
## brand "US Cellular"
## startup "$100"
## monthly "$400"
## service "4G NO"
## retail "Retail YES"
## apple "Apple YES"
## samsung "Samsung YES"
## google "Nexus YES"
##
## Question 12
          alt.1
##
## brand "T-Mobile"
## startup "$200"
## monthly "$200"
## service "4G NO"
## retail "Retail YES"
## apple "Apple YES"
```

```
## samsung "Samsung NO"
## google "Nexus NO"
##
## Question 13
##
           alt.1
## brand "T-Mobile"
## startup "$100"
## monthly "$300"
## service "4G YES"
## retail "Retail YES"
## apple "Apple NO"
## samsung "Samsung YES"
## google "Nexus NO"
##
## Question 14
           alt.1
##
## brand "US Cellular"
## startup "$200"
## monthly "$100"
## service "4G YES"
## retail "Retail YES"
## apple "Apple NO"
## samsung "Samsung NO"
## google "Nexus YES"
##
## Question 15
##
           alt.1
## brand "T-Mobile"
## startup "$300"
## monthly "$400"
## service "4G NO"
```

```
## retail "Retail NO"
## apple "Apple NO"
## samsung "Samsung NO"
## google "Nexus YES"
##
## Question 16
##
          alt.1
## brand "AT&T"
## startup "$300"
## monthly "$200"
## service "4G YES"
## retail "Retail YES"
## apple "Apple NO"
## samsung "Samsung YES"
## google "Nexus YES"
##
## NULL
```

```
#sink("questions_for_survey.txt") # send survey to external text file
questionnaire(provider.survey)
```

```
##
## Block 1
##
## Question 1
##
           alt.1
## brand "AT&T"
## startup "$100"
## monthly "$100"
## service "4G NO"
```

```
## retail "Retail NO"
## apple "Apple NO"
## samsung "Samsung NO"
## google "Nexus NO"
##
## Question 2
          alt.1
##
## brand "Verizon"
## startup "$300"
## monthly "$100"
## service "4G NO"
## retail "Retail YES"
## apple "Apple YES"
## samsung "Samsung YES"
## google "Nexus NO"
##
## Question 3
##
          alt.1
## brand "US Cellular"
## startup "$400"
## monthly "$200"
## service "4G NO"
## retail "Retail NO"
## apple "Apple NO"
## samsung "Samsung YES"
## google "Nexus NO"
##
## Question 4
##
          alt.1
## brand "Verizon"
## startup "$400"
```

```
## monthly "$400"
## service "4G YES"
## retail "Retail YES"
## apple "Apple NO"
## samsung "Samsung NO"
## google "Nexus NO"
##
## Question 5
          alt.1
##
## brand "Verizon"
## startup "$200"
## monthly "$300"
## service "4G NO"
## retail "Retail NO"
## apple "Apple NO"
## samsung "Samsung YES"
## google "Nexus YES"
##
## Question 6
##
          alt.1
## brand "Verizon"
## startup "$100"
## monthly "$200"
## service "4G YES"
## retail "Retail NO"
## apple "Apple YES"
## samsung "Samsung NO"
## google "Nexus YES"
##
## Question 7
           alt.1
##
```

```
## brand "US Cellular"
## startup "$300"
## monthly "$300"
## service "4G YES"
## retail "Retail NO"
## apple "Apple YES"
## samsung "Samsung NO"
## google "Nexus NO"
##
## Question 8
##
           alt.1
## brand
         "AT&T"
## startup "$400"
## monthly "$300"
## service "4G NO"
## retail "Retail YES"
## apple "Apple YES"
## samsung "Samsung NO"
## google "Nexus YES"
##
## Question 9
##
          alt.1
## brand
         "AT&T"
## startup "$200"
## monthly "$400"
## service "4G YES"
## retail "Retail NO"
## apple "Apple YES"
## samsung "Samsung YES"
## google "Nexus NO"
##
```

```
## Question 10
          alt.1
##
## brand "T-Mobile"
## startup "$400"
## monthly "$100"
## service "4G YES"
## retail "Retail NO"
## apple "Apple YES"
## samsung "Samsung YES"
## google "Nexus YES"
##
## Question 11
##
          alt.1
## brand "US Cellular"
## startup "$100"
## monthly "$400"
## service "4G NO"
## retail "Retail YES"
## apple "Apple YES"
## samsung "Samsung YES"
## google "Nexus YES"
##
## Question 12
          alt.1
##
## brand "T-Mobile"
## startup "$200"
## monthly "$200"
## service "4G NO"
## retail "Retail YES"
## apple "Apple YES"
## samsung "Samsung NO"
```

```
## google "Nexus NO"
##
## Question 13
##
           alt.1
## brand "T-Mobile"
## startup "$100"
## monthly "$300"
## service "4G YES"
## retail "Retail YES"
## apple "Apple NO"
## samsung "Samsung YES"
## google "Nexus NO"
##
## Question 14
##
           alt.1
## brand "US Cellular"
## startup "$200"
## monthly "$100"
## service "4G YES"
## retail "Retail YES"
## apple "Apple NO"
## samsung "Samsung NO"
## google "Nexus YES"
##
## Question 15
##
          alt.1
## brand "T-Mobile"
## startup "$300"
## monthly "$400"
## service "4G NO"
## retail "Retail NO"
```

```
"Apple NO"
## apple
## samsung "Samsung NO"
## google "Nexus YES"
##
## Question 16
##
           alt.1
           "AT&T"
## brand
## startup "$300"
## monthly "$200"
## service "4G YES"
## retail "Retail YES"
## apple
          "Apple NO"
## samsung "Samsung YES"
## google "Nexus YES"
```

```
#sink() # send output back to the screen
```

```
# user-defined function for plotting descriptive attribute names
effect.name.map <- function(effect.name) {
   if(effect.name=="brand") return("Mobile Service Provider")
   if(effect.name=="startup") return("Start-up Cost")
   if(effect.name=="monthly") return("Monthly Cost")
   if(effect.name=="service") return("Offers 4G Service")
   if(effect.name=="retail") return("Has Nearby Retail Store")
   if(effect.name=="apple") return("Sells Apple Products")
   if(effect.name=="samsung") return("Sells Samsung Products")
   if(effect.name=="google") return("Sells Google/Nexus Products")
}</pre>
```

conjoint.data.frame <- read.csv("/Users/neha/Documents/marketing/MDS_Chapter_1/mobile_services_
ranking.csv")</pre>

```
# set up sum contrasts for effects coding as needed for conjoint analysis
options(contrasts=c("contr.sum","contr.poly"))

# main effects model specification
main.effects.model <- {ranking ~ brand + startup + monthly + service +
    retail + apple + samsung + google}

# fit linear regression model using main effects only (no interaction terms)
main.effects.model.fit <- lm(main.effects.model, data=conjoint.data.frame)
print(summary(main.effects.model.fit))</pre>
```

```
##
## Call:
## lm.default(formula = main.effects.model, data = conjoint.data.frame)
##
## Residuals:
##
              2
                     3
                           4
                                  5
                                         6
                                                                   10
## -0.125  0.125  0.125  -0.125  -0.125  0.125  0.125  0.125  0.125
##
      11
             12
                    13
                           14
                                 15
                                        16
## -0.125 -0.125 0.125 0.125 0.125 -0.125
##
## Coefficients:
##
                Estimate Std. Error t value Pr(>|t|)
## (Intercept) 8.500e+00 1.250e-01 68.000
                                           0.00936 **
## brand1 1.374e-16 2.165e-01 0.000
                                           1.00000
## brand2
              -2.500e-01 2.165e-01 -1.155 0.45437
## brand3
              -1.202e-16 2.165e-01 0.000 1.00000
```

```
## startup1
              7.500e-01 2.165e-01 3.464
                                          0.17891
             8.240e-16 2.165e-01 0.000
## startup2
                                          1.00000
## startup3
              -2.794e-16 2.165e-01 0.000
                                          1.00000
## monthly1
              5.000e+00 2.165e-01 23.094 0.02755 *
## monthly2
             2.000e+00 2.165e-01 9.238 0.06865 .
## monthly3
              -1.250e+00 2.165e-01 -5.774 0.10918
## service1
              -1.750e+00 1.250e-01 -14.000
                                          0.04540 *
## retail1
              2.500e-01 1.250e-01 2.000
                                          0.29517
## apple1
            2.500e-01 1.250e-01 2.000
                                          0.29517
## samsung1
              -1.125e+00 1.250e-01 -9.000
                                          0.07045 .
## google1
              -7.500e-01 1.250e-01 -6.000 0.10514
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.5 on 1 degrees of freedom
## Multiple R-squared: 0.9993, Adjusted R-squared: 0.989
## F-statistic: 97.07 on 14 and 1 DF, p-value: 0.0794
```

```
# save key list elements of the fitted model as needed for conjoint measures
conjoint.results <-
    main.effects.model.fit[c("contrasts","xlevels","coefficients")]

conjoint.results$attributes <- names(conjoint.results$contrasts)

# compute and store part-worths in the conjoint.results list structure
part.worths <- conjoint.results$xlevels # list of same structure as xlevels
end.index.for.coefficient <- 1 # intitialize skipping the intercept
part.worth.vector <- NULL # used for accumulation of part worths
for(index.for.attribute in seq(along=conjoint.results$contrasts)) {
    nlevels <- length(unlist(conjoint.results$xlevels[index.for.attribute]))
    begin.index.for.coefficient <- end.index.for.coefficient + 1</pre>
```

```
end.index.for.coefficient <- begin.index.for.coefficient + nlevels -2
last.part.worth <- -sum(conjoint.results$coefficients[
   begin.index.for.coefficient:end.index.for.coefficient])
part.worths[index.for.attribute] <-
   list(as.numeric(c(conjoint.results$coefficients[
        begin.index.for.coefficient:end.index.for.coefficient],
        last.part.worth)))
part.worth.vector <-
   c(part.worth.vector,unlist(part.worths[index.for.attribute]))
}
conjoint.results$part.worths <- part.worths</pre>
```

Parth-worth, standardized parth worth and attribute importance

```
# compute standardized part-worths
standardize <- function(x) {(x - mean(x)) / sd(x)}
conjoint.results$standardized.part.worths <-
    lapply(conjoint.results$part.worths, standardize)

# compute and store part-worth ranges for each attribute
part.worth.ranges <- conjoint.results$contrasts
for(index.for.attribute in seq(along=conjoint.results$contrasts))
    part.worth.ranges[index.for.attribute] <-
    dist(range(conjoint.results$part.worths[index.for.attribute]))
conjoint.results$part.worth.ranges <- part.worth.ranges
sum.part.worth.ranges <- sum(as.numeric(conjoint.results$part.worth.ranges))</pre>
```

```
attribute.importance <- conjoint.results$contrasts</pre>
for(index.for.attribute in seq(along=conjoint.results$contrasts))
  attribute.importance[index.for.attribute] <-</pre>
  (dist(range(conjoint.results$part.worths[index.for.attribute]))/
  sum.part.worth.ranges) * 100
conjoint.results$attribute.importance <- attribute.importance</pre>
# data frame for ordering attribute names
attribute.name <- names(conjoint.results$contrasts)</pre>
attribute.importance <- as.numeric(attribute.importance)</pre>
temp.frame <- data.frame(attribute.name,attribute.importance)</pre>
conjoint.results$ordered.attributes <-</pre>
  as.character(temp.frame[sort.list(
  temp.frame$attribute.importance, decreasing = TRUE), "attribute.name"])
# respondent internal consistency added to list structure
conjoint.results$internal.consistency <- summary(main.effects.model.fit)$r.squared</pre>
# user-defined function for printing conjoint measures
```

compute and store importance values for each attribute

```
# user-defined function for printing conjoint measures
if (print.digits == 2)
  pretty.print <- function(x) {sprintf("%1.2f", round(x, digits = 2))}
if (print.digits == 3)
  pretty.print <- function(x) {sprintf("%1.3f", round(x, digits = 3))}</pre>
```

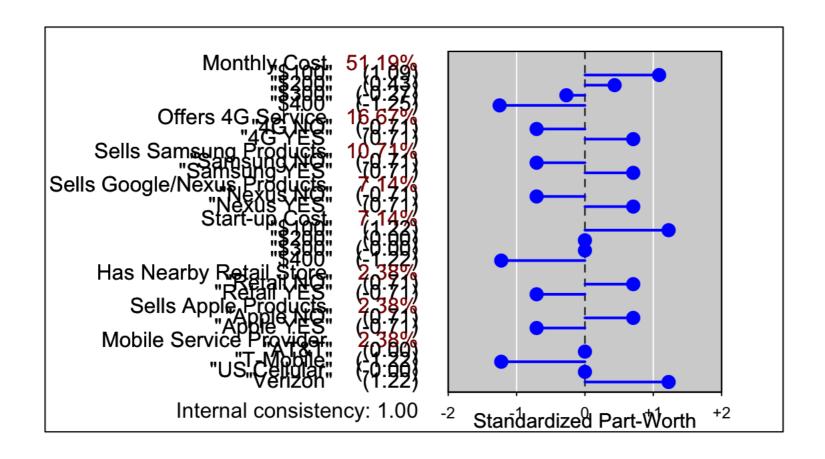
Results

```
# report conjoint measures to console
# use pretty.print to provide nicely formated output
```

```
for(k in seg(along=conjoint.results$ordered.attributes)) {
 cat("\n", "\n")
 cat(conjoint.results$ordered.attributes[k], "Levels: ",
 unlist(conjoint.results$xlevels[conjoint.results$ordered.attributes[k]]))
 cat("\n"," Part-Worths: ")
 cat(pretty.print(unlist(conjoint.results$part.worths
    [conjoint.results$ordered.attributes[k]]))
 cat("\n"," Standardized Part-Worths: ")
 cat(pretty.print(unlist(conjoint.results$standardized.part.worths
    [conjoint.results$ordered.attributes[k]]))
 cat("\n"," Attribute Importance: ")
 cat(pretty.print(unlist(conjoint.results$attribute.importance
    [conjoint.results$ordered.attributes[k]]))
```

```
##
##
## monthly Levels: "$100" "$200" "$300" "$400"
     Part-Worths: 5.00 2.00 -1.25 -5.75
##
##
     Standardized Part-Worths: 1.09 0.43 -0.27 -1.25
##
    Attribute Importance: 51.19
##
## service Levels: "4G NO" "4G YES"
     Part-Worths: -1.75 1.75
##
##
     Standardized Part-Worths: -0.71 0.71
##
     Attribute Importance: 16.67
##
## samsung Levels: "Samsung NO" "Samsung YES"
```

```
##
     Part-Worths: -1.12 1.12
    Standardized Part-Worths: -0.71 0.71
##
    Attribute Importance: 10.71
##
##
## google Levels: "Nexus NO" "Nexus YES"
     Part-Worths: -0.75 0.75
##
    Standardized Part-Worths: -0.71 0.71
##
    Attribute Importance: 7.14
##
##
## startup Levels: "$100" "$200" "$300" "$400"
    Part-Worths: 0.75 0.00 -0.00 -0.75
##
    Standardized Part-Worths: 1.22 0.00 -0.00 -1.22
##
    Attribute Importance: 7.14
##
##
## retail Levels: "Retail NO" "Retail YES"
##
     Part-Worths: 0.25 -0.25
    Standardized Part-Worths: 0.71 -0.71
##
    Attribute Importance: 2.38
##
##
## apple Levels: "Apple NO" "Apple YES"
    Part-Worths: 0.25 -0.25
##
    Standardized Part-Worths: 0.71 -0.71
##
##
    Attribute Importance: 2.38
##
## brand Levels: "AT&T" "T-Mobile" "US Cellular" "Verizon"
    Part-Worths: 0.00 -0.25 -0.00 0.25
##
    Standardized Part-Worths: 0.00 -1.22 -0.00 1.22
##
##
    Attribute Importance: 2.38
```



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