

Marketing Models - Assignment 1

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Load libraries

```
library(stargazer)
```

```
##
```

```
## Please cite as:
```

```
## Hlavac, Marek (2018). stargazer: Well-Formatted Regression and Summary Statistics Tables.
```

```
## R package version 5.2.2. https://CRAN.R-project.org/package=stargazer
```

```
library(tidyverse)
```

```
## -- Attaching packages ----- tidyverse 1.3.0 --
```

```
## v ggplot2 3.3.2    v purrr  0.3.4
```

```
## v tibble  3.0.3    v dplyr  1.0.2
```

```
## v tidyr   1.1.2    v stringr 1.4.0
```

```
## v readr   1.3.1    v forcats 0.5.0
```

```
## -- Conflicts ----- tidyverse_conflicts() --
```

```
## x dplyr::filter() masks stats::filter()
```

```
## x dplyr::lag()    masks stats::lag()
```

Load the data

```
data <- load("Assignment_1.RData")
```

Assignment 1

Inspect the data

```
head(Conjoint_data)
```

```
##   respondent_id profile      form  noapply disinfect bio    price choice
## 1             1      1 Concentrate 200 times      Yes  No 35 cents      1
## 2             1      2      Powder 200 times      Yes  No 35 cents      0
## 3             1      3      Premix 100 times      Yes Yes 49 cents      1
## 4             1      4      Powder 200 times      Yes Yes 49 cents      0
## 5             1      5      Powder  50 times      Yes  No 79 cents      0
## 6             1      6 Concentrate 200 times      No  Yes 79 cents      0
```

Estimate the model

```
# Set baseline for variables in regression
Conjoint_data$form <- relevel(Conjoint_data$form, ref = "Powder")
Conjoint_data$noapply <- relevel(Conjoint_data$noapply, ref = "200 times")
Conjoint_data$disinfect <- relevel(Conjoint_data$disinfect, ref = "No")
Conjoint_data$price <- relevel(Conjoint_data$price, ref = "35 cents")

# Estimate model
model <-
  glm(choice ~ form + noapply + disinfect + bio + price,
       family = "binomial",
       data = Conjoint_data)

# Show results
results <- summary(model)
results
```

```
##
## Call:
## glm(formula = choice ~ form + noapply + disinfect + bio + price,
##      family = "binomial", data = Conjoint_data)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -2.1680  -0.7913   0.3276   0.8173   2.1374
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)      1.1357     0.4662   2.436 0.014859 *
## formConcentrate    0.6482     0.3293   1.968 0.049035 *
## formPremix       -0.2519     0.3347  -0.753 0.451638
## noapply100 times -0.2335     0.3457  -0.675 0.499416
## noapply50 times  -0.5904     0.3601  -1.639 0.101111
## disinfectYes      1.1143     0.2998   3.717 0.000202 ***
## bioYes            0.2762     0.2862   0.965 0.334477
## price49 cents    -1.3828     0.3609  -3.831 0.000128 ***
## price79 cents    -3.0606     0.3810  -8.034 9.46e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
##
## (Dispersion parameter for binomial family taken to be 1)
##
## Null deviance: 449.25 on 329 degrees of freedom
## Residual deviance: 348.95 on 321 degrees of freedom
## AIC: 366.95
##
## Number of Fisher Scoring iterations: 4
```

```
# Tabulate results for report
stargazer(model, type = "text", title = "Logistic Regression Model",
           intercept.bottom = FALSE, no.space = TRUE)
```

```
##
## Logistic Regression Model
## =====
##                               Dependent variable:
##                               -----
##                               choice
## -----
## Constant                      1.136**
##                               (0.466)
## formConcentrate                0.648**
##                               (0.329)
## formPremix                    -0.252
##                               (0.335)
## noapply100 times              -0.234
##                               (0.346)
## noapply50 times               -0.590
##                               (0.360)
## disinfectYes                  1.114***
##                               (0.300)
## bioYes                        0.276
##                               (0.286)
## price49 cents                 -1.383***
##                               (0.361)
## price79 cents                 -3.061***
##                               (0.381)
## -----
## Observations                   330
## Log Likelihood                -174.477
## Akaike Inf. Crit.             366.955
## =====
## Note:                         *p<0.1; **p<0.05; ***p<0.01
```

Create empty part-worth

```
# Initialize empty part-worth table (type list)
partworth <-
  list("form" = rep(0, 3), "noapply" = rep(0, 3), "disinfect" = rep(0, 2),
       "bio" = rep(0, 2), "price" = rep(0, 3))
```

```

# Set levels part-worth
names(partworth$form) <- levels(Conjoint_data$form)
names(partworth$noapply) <- levels(Conjoint_data$noapply)
names(partworth$disinfect) <- levels(Conjoint_data$disinfect)
names(partworth$bio) <- levels(Conjoint_data$bio)
names(partworth$price) <- levels(Conjoint_data$price)

# Set insignificant coefficients to zero
coeffs <- results$coefficients[,1] * (results$coefficients[,4] < .05)
coeffs[1]

```

```

## (Intercept)
##      1.135682

```

```

# Fill the part-worth table
partworth$form[2:3] <- coeffs[2:3]
partworth$noapply[2:3] <- coeffs[4:5]
partworth$disinfect[2] <- coeffs[6]
partworth$bio[2] <- coeffs[7]
partworth$price[2:3] <- coeffs[8:9]

```

```

# Show results
partworth

```

```

## $form
##      Powder Concentrate      Premix
##  0.0000000  0.6481925  0.0000000
##
## $noapply
## 200 times 100 times  50 times
##         0         0         0
##
## $disinfect
##      No      Yes
## 0.000000 1.114257
##
## $bio
##  No Yes
##   0   0
##
## $price
## 35 cents 49 cents 79 cents
## 0.000000 -1.382753 -3.060619

```

Compute probability

```

prob <-
  exp(coeffs[[1]] + max(partworth$form) + max(partworth$noapply) +
        max(partworth$disinfect) + max(partworth$bio) + max(partworth$price)) /
  (1 + exp(coeffs[[1]] + max(partworth$form) + max(partworth$noapply) +

```

```

max(partworth$disinfect) + max(partworth$bio) +
max(partworth$price)))
prob

```

```
## [1] 0.947754
```

Question 2

Question 2a

```

lik_bgnbd <- function(para) {
  # This is the likelihood function with CLV_data
  # Change CLV_data to the names your datasets; the variables should be the same.

  # unpack 4 parameters:
  # r and alpha are the parameters of the gamma distribution of lambda
  # a and b are the parameters of the beta distribution of the "death" rate
  r <- para[1]
  alpha <- para[2]
  a <- para[3]
  b <- para[4]

  # unpack three variables:
  # x - # of transactions (frequency)
  # t_x - the most recent time of transaction (recency)
  # T - the end time for the Poisson process
  x <- CLV_data$x
  t_x <- CLV_data$t_x
  T <- CLV_data$T

  # A1 - A4 corresponds to the log of 4 terms in the likelihood function
  # See p.280 of Fader et al. (2005)
  # They are the log of A1-A4 for the log-likelihood
  A1 <- lgamma(r+x) + r*log(alpha) - lgamma(r)
  A2 <- lgamma(a+b) + lgamma(b+x) - lgamma(b) - lgamma(a+b+x)
  A3 <- -(r+x)*log(alpha+T)

  A4 <- rep(0,length(x))
  idx <- x>0
  if (sum(idx)>0) {
    A4[idx] <- log(a/(b+x[idx]-1)) - (r+x[idx])*log(alpha+t_x[idx])
  }

  lh <- sum(A1+A2+log(exp(A3)+idx*exp(A4)))

  # To return the minus of likelihood for minimization
  # maximize likelihood = minimize negative likelihood
  return(-lh)
}

```

```

pred_Y <- function(para, H, t) {
  # Given estimated parameters, for a customer with history {x,t_x,T}, to
  # predict her # of transactions in t time.
  # Please see Equation (10) of Fader et al. (2005), p.279;
  # This function uses a package "hypergeo" for function evaluation;
  # As evaluation of 2F1(.) is somewhat costly, this function is to accommodate
  # only 1 customer;
  # You can extend the function to calculate for many customers by iterations or
  # use lapply(.);

  # The function takes three input:
  # para - a vector of the 4 parameters;
  # H - a data.frame with only 1 row (for 1 customer) and three variables
  # {x,t_x,T};
  # t - how long into the future?

  # unpack 4 parameters:
  # r and alpha are the parameters of the gamma distribution of lambda
  # a and b are the parameters of the beta distribution of the "death" rate
  r <- para[1]
  alpha <- para[2]
  a <- para[3]
  b <- para[4]

  # unpack three variables:
  # x - # of transactions (frequency)
  # t_x - the most recent time of transaction (recency)
  # T - the end time for the Poisson process
  x <- H$x
  t_x <- H$t_x
  T <- H$T

  # the term of the Gaussian hypergeometric function 2F1(.)
  hg <- hypergeo::hypergeo(r+x,b+x,a+b+x-1,t/(alpha+T+t))
  hg <- Re(hg) # hg is a real number in the complex format Re(.) makes it real.

  # the numerator term, given hg
  Y1 <- (a+b+x-1)/(a-1)*((1-(((alpha+T)/(alpha+T+t))^(r+x)))*hg)
  Y2 <- 1
  if (x>0) {
    Y2 <- Y2 + a/(b+x-1)*(((alpha+T)/(alpha+t_x))^(r+x))
  }

  return(Y1/Y2)
}

```

Estimating the Parameters of the BG/NBD model

```

# to specify the constraint matrix
ui <- diag(4)
ci <- rep(0,4)

```

```

# para: the starting values; a vector with 4 elements corresponding to
# (r,alpha,a,b).
# you can change this to other values or test different values.
para <- rep(.01,4)

# to run the constrained optimization
results <- constrOptim(para, lik_bgnbd, NULL, ui, ci)

# to get the estimated parameters in the order of (r,alpha,a,b)
para <- results$par
names(para) <- c("r","alpha","a","b")

stargazer(para, summary=FALSE, title="Parameters GB/NBD model", type = "text")

```

```

##
## Parameters GB/NBD model
## =====
## r      alpha  a      b
## -----
## 0.232  8.957  1.194  3.191
## -----

```

Predict the No. of Transactions

```

# suppose we want to predict next year and set t = 52
t <- 30

preds <- c()
for(i in 1:5) {
  # let's predict the first customer in the data
  H <- CLV_data[i,]

  # to obtain the prediction
  Yhat <- pred_Y(para,H,t)
  preds <- append(preds, as.numeric(Yhat))
}
preds <- round(preds, 3)

res <- tibble("gamer ID" = c(rep(1:5)), "number of pruchase" = preds)

stargazer(res, summary=FALSE, type="text",
           title="Predict # of purchases in the next month (t=30)")

```

```

##
## Predict # of purchases in the next month (t=30)
## =====
##   gamer ID number of pruchase
## -----
## 1      1          0.178
## 2      2          0.243
## 3      3          0.087

```

##	4	4	0.753
##	5	5	0.089
##	-----		