Methods for Utility Evaluation #2

MATH 301 Data Confidentiality

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```
origdata<- read.csv("CEdata.csv")</pre>
head(origdata)
     UrbanRural Income Race Expenditure
## 1
              1 98600
                          1
                                5972.167
## 2
              1 24360
                                5854.500
              1 80200
## 3
                                5506.667
                           1
## 4
              1 150500
                                8968.891
## 5
              1 130000
                               10092.833
                           1
## 6
              1 32836
                                5520.267
```

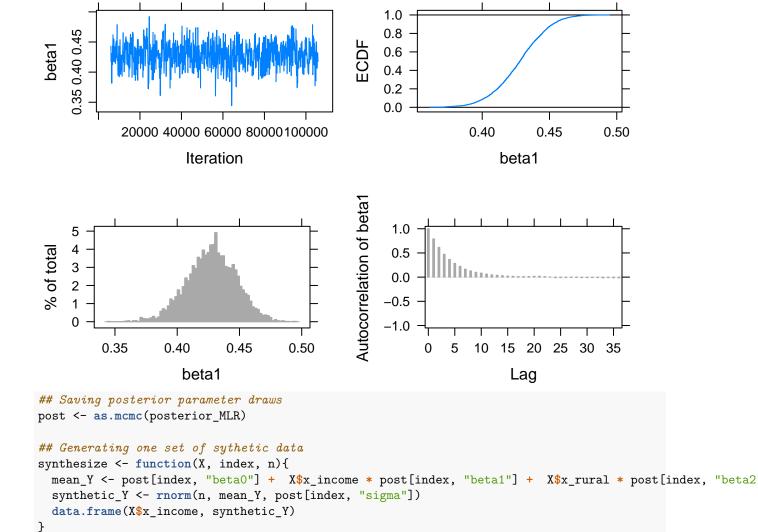
Read Drechsler (2001) Chapter 6-1, 7-1 in the References folder, and prepare the following results.

```
origdata$LogExp <- log(origdata$Expenditure)</pre>
origdata$LogIncome <- log(origdata$Income)</pre>
## create indicator variable for Rural (2)
origdata$Rural = fastDummies::dummy_cols(origdata$UrbanRural)[,names(fastDummies::dummy_cols(origdata$U
== ".data 1"]
## create indicator variables for Black (3), Native American (4),
## Asian (5), Pacific Islander (6), and Multi-race (7)
origdata$Race_Black = fastDummies::dummy_cols(origdata$Race)[,names(fastDummies::dummy_cols(origdata$Ra
origdata$Race_NA = fastDummies::dummy_cols(origdata$Race)[,names(fastDummies::dummy_cols(origdata$Race)
origdata$Race_Asian = fastDummies::dummy_cols(origdata$Race)[,names(fastDummies::dummy_cols(origdata$Race)
origdata$Race_PI = fastDummies::dummy_cols(origdata$Race)[,names(fastDummies::dummy_cols(origdata$Race)
origdata$Race_M = fastDummies::dummy_cols(origdata$Race)[,names(fastDummies::dummy_cols(origdata$Race))
## JAGS script
modelString <-"
model {
## sampling
for (i in 1:N){
y[i] ~ dnorm(beta0 + beta1*x_income[i] + beta2*x_rural[i] +
beta3*x_race_B[i] + beta4*x_race_N[i] +
beta5*x_race_A[i] + beta6*x_race_P[i] +
beta7*x_race_M[i], invsigma2)
}
## priors
beta0 ~ dnorm(mu0, g0)
beta1 ~ dnorm(mu1, g1)
beta2 ~ dnorm(mu2, g2)
beta3 ~ dnorm(mu3, g3)
beta4 ~ dnorm(mu4, g4)
beta5 ~ dnorm(mu5, g5)
```

```
beta6 ~ dnorm(mu6, g6)
beta7 ~ dnorm(mu7, g7)
invsigma2 ~ dgamma(a, b)
sigma <- sqrt(pow(invsigma2, -1))</pre>
}"
y = as.vector(origdata$LogExp)
x_income = as.vector(origdata$LogIncome)
x_rural = as.vector(origdata$Rural)
x_race_B = as.vector(origdata$Race_Black)
x_race_N = as.vector(origdata$Race_NA)
x_race_A = as.vector(origdata$Race_Asian)
x_race_P = as.vector(origdata$Race_PI)
x_race_M = as.vector(origdata$Race_M)
N = length(y) # Compute the number of observations
## Pass the data and hyperparameter values to JAGS
the_data <- list("y" = y, "x_income" = x_income,</pre>
"x_rural" = x_rural, "x_race_B" = x_race_B,
"x_race_N" = x_race_N, "x_race_A" = x_race_A,
"x_race_P" = x_race_P, "x_race_M" = x_race_M,
"N" = N,
"mu0" = 0, "g0" = 1, "mu1" = 0, "g1" = 1,
mu2" = 0, g2" = 1, mu3" = 0, g3" = 1,
"mu4" = 0, "g4" = 1, "mu5" = 0, "g5" = 1,
"mu6" = 0, "g6" = 1, "mu7" = 0, "g7" = 1,
a'' = 1, b'' = 1
initsfunction <- function(chain){</pre>
.RNG.seed \leftarrow c(1,2) [chain]
.RNG.name <- c("base::Super-Duper",
"base::Wichmann-Hill")[chain]
return(list(.RNG.seed=.RNG.seed,
.RNG.name=.RNG.name))
}
## Run the JAGS code for this model:
posterior MLR <- run.jags(modelString,</pre>
n.chains = 1,
data = the_data,
monitor = c("beta0", "beta1", "beta2",
"beta3", "beta4", "beta5",
"beta6", "beta7", "sigma"),
adapt = 1000,
burnin = 5000,
sample = 5000,
thin = 20,
inits = initsfunction)
## Loading required namespace: rjags
## Compiling rjags model...
## Calling the simulation using the rjags method...
## Note: the model did not require adaptation
```

```
## Burning in the model for 5000 iterations...
## Running the model for 100000 iterations...
## Simulation complete
## Calculating summary statistics...
## Warning: Convergence cannot be assessed with only 1 chain
## Finished running the simulation
## JAGS output
summary(posterior_MLR)
            Lower95
                          Median
                                     Upper95
                                                    Mean
                                                                 SD Mode
## beta0 3.58046793 4.00567714 4.46873420
                                              4.01026369 0.22739281
                                                                      NA
## beta1
         0.38869072
                     0.42806382
                                 0.46552491
                                              0.42748795 0.01979320
                                                                      NA
## beta2 0.06829052
                     0.26991798
                                 0.47936432
                                              0.27011927 0.10454790
                                                                      NA
## beta3 -0.33605648 -0.19443785 -0.04513397 -0.19471621 0.07412360
                                                                      NA
                                0.52291297
## beta4 -0.51146051
                     0.01343539
                                              0.01371237 0.26926773
                                                                      NA
## beta5 -0.06626353
                     0.15835010
                                 0.39460914
                                              0.15949581 0.11802871
                                                                      NA
## beta6 -0.44077361
                     0.08271013
                                 0.67487227
                                              0.08950545 0.28350013
                                                                      NA
## beta7 -0.31034009 0.04244541 0.37173213 0.04256299 0.17347832
                                                                      NA
## sigma 0.69006164 0.72142424 0.75342096
                                              0.72145995 0.01616366
                                                                      NA
##
               MCerr MC%ofSD SSeff
                                          AC.200 psrf
## beta0 0.0093617177
                          4.1
                               590
                                    0.095231969
## beta1 0.0007975010
                          4.0
                               616
                                    0.083523145
                                                   NA
## beta2 0.0020741456
                          2.0 2541
                                    0.005204065
## beta3 0.0010482660
                          1.4 5000
                                    0.009144817
                                                   NA
## beta4 0.0038080208
                          1.4
                              5000
                                    0.001185663
                                                   NA
## beta5 0.0016691781
                          1.4 5000 -0.026020203
                                                   NA
## beta6 0.0040092973
                          1.4 5000
                                    0.004342214
                                                   NA
## beta7 0.0024533540
                          1.4
                              5000
                                    0.004259465
                                                   NA
## sigma 0.0002285887
                          1.4 5000
                                    0.010254110
                                                   NA
plot(posterior_MLR, vars = "beta1")
```

Generating plots...



i. Generate m=20 synthetic datasets given your synthesis model for the CE sample. If you are using set.seed(), make sure that you do not generate the same synthetic data for each m=20.

```
set.seed(123)
m <- 20
n <- dim(origdata)[1]
synthetic_m <- vector("list",m)
new <- data.frame(x_income, x_rural, x_race_B, x_race_N, x_race_A, x_race_P, x_race_M)
for (l in 1:m){
    synthetic_one <- synthesize(new, 4980+1, n)
    names(synthetic_one) <- c("OrigLogIncome", "SynLogIncome")
    synthetic_m[[1]] <- synthetic_one
}</pre>
```

ii. Estimate a few analysis-specific utility measures, e.g. the mean and median of a continuous synthetic variable, the regression analysis coefficients, for each synthetic dataset.

```
## Estimates the mean, median, mode, variance, and range of synthetic log Income, as well as regression mean \leftarrow c()
```

```
median \leftarrow c()
mode <- c()
variance <- c()</pre>
range <- c()
for (1 in 1:m){
  mean[1] = mean(synthetic_m[[1]]$SynLogIncome)
 median[1] = median(synthetic m[[1]]$SynLogIncome)
 mode[1] = mode(synthetic m[[1]]$SynLogIncome)
  variance[l] = var(synthetic_m[[l]]$SynLogIncome)
  range[1] = range(synthetic_m[[1]]$SynLogIncome)
  print(lm(origdata$LogExp ~ synthetic_m[[1]]$SynLogIncome))
}
## Warning in range[1] <- range(synthetic_m[[1]]$SynLogIncome): number of
## items to replace is not a multiple of replacement length
##
## Call:
## lm(formula = origdata$LogExp ~ synthetic_m[[1]]$SynLogIncome)
## Coefficients:
                      (Intercept) synthetic_m[[1]]$SynLogIncome
##
                          5.9110
##
                                                           0.3257
## Warning in range[1] <- range(synthetic_m[[1]]$SynLogIncome): number of
## items to replace is not a multiple of replacement length
##
## Call:
## lm(formula = origdata$LogExp ~ synthetic_m[[1]]$SynLogIncome)
## Coefficients:
##
                      (Intercept) synthetic_m[[1]]$SynLogIncome
                           6.0121
##
                                                           0.3145
## Warning in range[1] <- range(synthetic_m[[1]]$SynLogIncome): number of</pre>
## items to replace is not a multiple of replacement length
##
## Call:
## lm(formula = origdata$LogExp ~ synthetic_m[[1]]$SynLogIncome)
## Coefficients:
##
                                  synthetic_m[[1]]$SynLogIncome
                      (Intercept)
##
                           6.1379
                                                           0.3024
## Warning in range[1] <- range(synthetic m[[1]]$SynLogIncome): number of
## items to replace is not a multiple of replacement length
##
## lm(formula = origdata$LogExp ~ synthetic_m[[1]]$SynLogIncome)
## Coefficients:
##
                      (Intercept)
                                   synthetic_m[[1]]$SynLogIncome
##
                           5.9313
                                                           0.3243
```

```
## Warning in range[1] <- range(synthetic_m[[1]]$SynLogIncome): number of
## items to replace is not a multiple of replacement length
##
## Call:
## lm(formula = origdata$LogExp ~ synthetic_m[[1]]$SynLogIncome)
## Coefficients:
##
                     (Intercept) synthetic_m[[1]]$SynLogIncome
##
                          5.6249
## Warning in range[1] <- range(synthetic_m[[1]]$SynLogIncome): number of
## items to replace is not a multiple of replacement length
##
## Call:
## lm(formula = origdata$LogExp ~ synthetic_m[[1]]$SynLogIncome)
## Coefficients:
##
                     (Intercept) synthetic m[[1]]$SynLogIncome
                          5.8679
                                                          0.3327
##
## Warning in range[1] <- range(synthetic_m[[1]]$SynLogIncome): number of
## items to replace is not a multiple of replacement length
##
## Call:
## lm(formula = origdata$LogExp ~ synthetic_m[[1]]$SynLogIncome)
##
## Coefficients:
##
                                  synthetic_m[[1]]$SynLogIncome
                     (Intercept)
##
                           5.803
                                                           0.341
## Warning in range[1] <- range(synthetic_m[[1]]$SynLogIncome): number of</pre>
## items to replace is not a multiple of replacement length
##
## Call:
## lm(formula = origdata$LogExp ~ synthetic m[[1]]$SynLogIncome)
##
## Coefficients:
##
                                  synthetic_m[[1]]$SynLogIncome
                     (Intercept)
##
                          5.5808
## Warning in range[1] <- range(synthetic m[[1]]$SynLogIncome): number of
## items to replace is not a multiple of replacement length
##
## Call:
## lm(formula = origdata$LogExp ~ synthetic_m[[1]]$SynLogIncome)
## Coefficients:
                                  synthetic_m[[1]]$SynLogIncome
##
                     (Intercept)
##
                          5.9365
                                                          0.3238
## Warning in range[1] <- range(synthetic_m[[1]]$SynLogIncome): number of
## items to replace is not a multiple of replacement length
##
```

```
## Call:
## lm(formula = origdata$LogExp ~ synthetic_m[[1]]$SynLogIncome)
## Coefficients:
                      (Intercept) synthetic_m[[1]]$SynLogIncome
##
##
                           6.2384
## Warning in range[1] <- range(synthetic_m[[1]]$SynLogIncome): number of</pre>
## items to replace is not a multiple of replacement length
##
## Call:
## lm(formula = origdata$LogExp ~ synthetic_m[[1]]$SynLogIncome)
## Coefficients:
##
                      (Intercept)
                                   synthetic_m[[1]]$SynLogIncome
##
                           5.8536
                                                           0.3319
## Warning in range[1] <- range(synthetic_m[[1]]$SynLogIncome): number of</pre>
## items to replace is not a multiple of replacement length
##
## Call:
## lm(formula = origdata$LogExp ~ synthetic_m[[1]]$SynLogIncome)
##
## Coefficients:
##
                                   synthetic_m[[1]]$SynLogIncome
                      (Intercept)
##
                           6.2642
                                                           0.2872
## Warning in range[1] <- range(synthetic_m[[1]]$SynLogIncome): number of</pre>
## items to replace is not a multiple of replacement length
##
## Call:
## lm(formula = origdata$LogExp ~ synthetic_m[[1]]$SynLogIncome)
## Coefficients:
                                   synthetic_m[[1]]$SynLogIncome
##
                      (Intercept)
##
                           5.7997
## Warning in range[1] <- range(synthetic_m[[1]]$SynLogIncome): number of
## items to replace is not a multiple of replacement length
##
## Call:
## lm(formula = origdata$LogExp ~ synthetic_m[[1]]$SynLogIncome)
## Coefficients:
                                   synthetic_m[[1]]$SynLogIncome
                      (Intercept)
##
                           5.9573
                                                           0.3209
##
## Warning in range[1] <- range(synthetic_m[[1]]$SynLogIncome): number of</pre>
## items to replace is not a multiple of replacement length
##
## Call:
## lm(formula = origdata$LogExp ~ synthetic_m[[1]]$SynLogIncome)
## Coefficients:
```

```
##
                     (Intercept) synthetic_m[[1]]$SynLogIncome
                          5.8145
##
                                                          0.3377
## Warning in range[1] <- range(synthetic_m[[1]]$SynLogIncome): number of
## items to replace is not a multiple of replacement length
##
## Call:
## lm(formula = origdata$LogExp ~ synthetic_m[[1]]$SynLogIncome)
## Coefficients:
##
                     (Intercept)
                                   synthetic_m[[1]]$SynLogIncome
                          6.0456
##
                                                          0.3119
## Warning in range[1] <- range(synthetic_m[[1]]$SynLogIncome): number of</pre>
## items to replace is not a multiple of replacement length
##
## Call:
## lm(formula = origdata$LogExp ~ synthetic_m[[1]]$SynLogIncome)
## Coefficients:
##
                     (Intercept) synthetic_m[[1]]$SynLogIncome
##
                          5.8341
## Warning in range[1] <- range(synthetic m[[1]]$SynLogIncome): number of
## items to replace is not a multiple of replacement length
##
## Call:
## lm(formula = origdata$LogExp ~ synthetic_m[[1]]$SynLogIncome)
## Coefficients:
                     (Intercept) synthetic_m[[1]]$SynLogIncome
##
##
                           5,666
                                                           0.356
## Warning in range[1] <- range(synthetic_m[[1]]$SynLogIncome): number of
## items to replace is not a multiple of replacement length
##
## Call:
## lm(formula = origdata$LogExp ~ synthetic_m[[1]]$SynLogIncome)
## Coefficients:
##
                     (Intercept)
                                  synthetic_m[[1]]$SynLogIncome
##
                          5.7171
                                                          0.3501
## Warning in range[1] <- range(synthetic_m[[1]]$SynLogIncome): number of
## items to replace is not a multiple of replacement length
##
## lm(formula = origdata$LogExp ~ synthetic_m[[1]]$SynLogIncome)
## Coefficients:
##
                                   synthetic_m[[1]]$SynLogIncome
                     (Intercept)
                          5.9466
                                                          0.3232
##
```

```
syndata <- synthetic_m[[1]]</pre>
```

Step 1: calculate key quantities

```
known.vars <- c("Rural", "Race", "Expenditure")</pre>
syn.vars <- c("LogIncome")</pre>
CEdata <- data.frame(origdata$Rural, origdata$Race, origdata$Expenditure, origdata$LogIncome)
CEdatasyn <- data.frame(syndata$Rural, syndata$Race, syndata$Expenditure, syndata$LogIncome)
n <- dim(origdata)[1]</pre>
KeyQuantities1 <- CalculateKeyQuantities(CEdata, CEdatasyn, known.vars, syn.vars, n)
## Step 2: calculate 3 summary measures
IdentificationRisk <- function(c_vector, T_vector, K_vector, F_vector, s, N){</pre>
  nonzero_c_index <- which(c_vector > 0)
  exp_match_risk <- sum(1/c_vector[nonzero_c_index]*T_vector[nonzero_c_index])
  true_match_rate <- sum(na.omit(K_vector))/N</pre>
  false_match_rate <- sum(na.omit(F_vector))/s</pre>
  res_r <- list(exp_match_risk = exp_match_risk,</pre>
                 true_match_rate = true_match_rate,
                 false_match_rate = false_match_rate
  )
  return(res_r)
}
## each record is a target, therefore N = n
c_vector <- KeyQuantities1[["c_vector"]]</pre>
T_vector <- KeyQuantities1[["T_vector"]]</pre>
K_vector <- KeyQuantities1[["K_vector"]]</pre>
F_vector <- KeyQuantities1[["F_vector"]]</pre>
s <- KeyQuantities1[["s"]]</pre>
N <- n
ThreeSummaries <- IdentificationRisk(c_vector, T_vector, K_vector, F_vector, s, N)
```

Summaries:

```
## Expected match risk
ThreeSummaries[["exp_match_risk"]]

## [1] 0

## True match rate
ThreeSummaries[["true_match_rate"]]

## [1] 0

## False match rate
ThreeSummaries[["false_match_rate"]]

## [1] NaN
```

Results and Discussion

I could not get the code to work for the identification disclosure risk for the continuous variable.