

Appendix: Implementation Code

```
## install packages
# install.packages("augSIMEX")
# install.packages("scales")
# install.packages("microbenchmark")
# install.packages("ggplot2")
library(augSIMEX)
library(scales)
library(microbenchmark)
library(ggplot2)

## Seed Generator
set.seed(2018)
Seeds <- sample(100000,size=1000,replace=FALSE)
options(warn=-1)

source("DataGenerator.R")

#####
## Section 5.1 Bias ##
## This Section will run approximately [6 hours] without parellal computing

Results_Table1 <- function(i,rate){
  tryCatch({
    Data <- DataGenerator(Seeds[i],rate,"1D")
    start <- proc.time()

    ## The main function of correcting measurement error and misclassification
    Re <- augSIMEX(mainformula = Y~Xstar+W+Zstar,
      mismodel=pi|qi~1, family = binomial,
      data=Data$Main, validationdata=Data$Validation, subset=NULL,
      err.var="Xstar", mis.var="Zstar", err.true="X", mis.true="Z",
      err.mat = NULL,
      lambda=NULL, M=5, B=20, nBoot=2, extrapolation=c("both"))
    ## Results without accounting for measurement error and misclassification
    naive <- glm(Y~Xstar+W+Zstar,family = binomial,data=Data$Main)
    end <- proc.time()-start
```

```

coefficients <- Re$coefficients
se <- (Re$vcov)
return(list(coefs.naive=naive$coefficients,
            coefs.linear=coefficients[(length(coefficients)/2+1):length(coefficients)],
            coefs.quadra=coefficients[1:(length(coefficients)/2)],
            se.naive=sqrt(diag(vcov(naive))),
            se.linear=se[(length(coefficients)/2+1):length(coefficients)],
            se.quadra=se[1:(length(coefficients)/2)],
            time=end[1]))
}, error = function(e) return(NA))
}

```

```

Table1_1 <- lapply(1:1000,FUN=Results_Table1,rate=0.1)
Table1_3 <- lapply(1:1000,FUN=Results_Table1,rate=0.3)
Table1_5 <- lapply(1:1000,FUN=Results_Table1,rate=0.5)

```

```

save(Table1_1,file="Table1_1.RData")
save(Table1_3,file="Table1_3.RData")
save(Table1_5,file="Table1_5.RData")

```

```
#####
```

```
## Evaluation and rearrange to obtain Table 1
```

```
## Function of computing performance results from linear and quadratic extrepolation
```

```

get_naive_Table1 <- function(data,truebeta){
  pooled_form <- unlist(lapply(data,FUN = function(x){
    if (is.na(x)) return(NA) else return(x$coefs.naive)}))
  matrix_form <- matrix(pooled_form,nrow=1000,byrow = T)
  colnames(matrix_form) <- c("(Intercept)","X","W","Z")
  matrix_form <- matrix_form[,c("(Intercept)","X","Z","W")]
  mean <- colMeans(matrix_form,na.rm = T)
  bias <- abs(mean-truebeta)
  emperical_sd <- apply(matrix_form,MARGIN=2,FUN=sd, na.rm = T)
  output <- cbind(bias,emperical_sd)
  return(as.matrix(output))
}

```

```

get_linear_Table1 <- function(data,truebeta){
  pooled_form <- unlist(lapply(data,FUN = function(x){
    if (is.na(x)) return(NA) else return(x$coefs.linear)}))
  matrix_form <- matrix(pooled_form,nrow=1000,byrow = T)
  colnames(matrix_form) <- names(data[[1]]$coefs.linear)
  matrix_form <- matrix_form[,c("(Intercept)","X","Z","W")]
  mean <- colMeans(matrix_form,na.rm = T)
  bias <- abs(mean-truebeta)
  emperical_sd <- apply(matrix_form,MARGIN=2,FUN=sd, na.rm = T)
  output <- cbind(bias,emperical_sd)
  return(as.matrix(output))
}

get_quadra_Table1 <- function(data,truebeta){
  pooled_form <- unlist(lapply(data,FUN = function(x){
    if (is.na(x)) return(NA) else return(x$coefs.quadra)}))
  matrix_form <- matrix(pooled_form,nrow=1000,byrow = T)
  colnames(matrix_form) <- names(data[[1]]$coefs.quadra)
  matrix_form <- matrix_form[,c("(Intercept)","X","Z","W")]
  mean <- colMeans(matrix_form,na.rm = T)
  bias <- abs(mean-truebeta)
  emperical_sd <- apply(matrix_form,MARGIN=2,FUN=sd ,na.rm = T)
  output <- cbind(bias,emperical_sd)
  return(as.matrix(output))
}

True_beta_Table1 <- c(0.1,-1,0.7,0.5)

Table1_1_naive <- get_naive_Table1(Table1_1,True_beta_Table1)
Table1_1_linear <- get_linear_Table1(Table1_1,True_beta_Table1)
Table1_1_quadratic <- get_quadra_Table1(Table1_1,True_beta_Table1)
Table1_3_naive <- get_naive_Table1(Table1_3,True_beta_Table1)
Table1_3_linear <- get_linear_Table1(Table1_3,True_beta_Table1)
Table1_3_quadratic <- get_quadra_Table1(Table1_3,True_beta_Table1)
Table1_5_naive <- get_naive_Table1(Table1_5,True_beta_Table1)
Table1_5_linear <- get_linear_Table1(Table1_5,True_beta_Table1)
Table1_5_quadratic <- get_quadra_Table1(Table1_5,True_beta_Table1)

```

```

Table1 <- cbind(rbind(Table1_1_naive,Table1_3_naive,Table1_5_naive),
               rbind(Table1_1_linear,Table1_3_linear,Table1_5_linear),
               rbind(Table1_1_quadratic,Table1_3_quadratic,Table1_5_quadratic))
Table1 <- round(Table1,3)
Table1

#####
## Section 5.2 variance estimation ##
## This Section will run approximately [6 hours] without parellal computing

Results_Table2_1D <- function(i){
  tryCatch({
    start <- proc.time()
    Data <- DataGenerator(Seeds[i],0.5,"1D")

    ## The main function of correcting measurement error and misclassification
    Re <- augSIMEX(mainformula = Y~Xstar+W+Zstar,
                  mismodel=pi|qi~1, family = binomial,
                  data=Data$Main,validationdata=Data$Validation, subset=NULL,
                  err.var="Xstar", mis.var="Zstar", err.true="X",
                  mis.true="Z", err.mat = NULL,
                  lambda=NULL, M=5, B=20, nBoot=50, extrapolation=c("both"))
    ## Results without accounting for measurement error and misclassification
    naive <- glm(Y~Xstar+W+Zstar,family = binomial,data=Data$Main)
    coefficients <- Re$coefficients
    se <- (Re$vcov)
  }, error = function(e) return(NA))
  end <- proc.time()-start

  return(list(coefs.naive=naive$coefficients,
             coefs.linear=coefficients[(length(coefficients)/2+1):length(coefficients)],
             coefs.quadra=coefficients[1:(length(coefficients)/2)],
             se.naive=sqrt(diag(vcov(naive))),
             se.linear=se[(length(coefficients)/2+1):length(coefficients)],
             se.quadra=se[1:(length(coefficients)/2)],
             time=end[1]))
}

```

```

Results_Table2_2D <- function(i){
  tryCatch({
    start <- proc.time()
    Data <- DataGenerator(Seeds[i],0.5,"2D")
    meformula <- Xstar.X1|Xstar.X2~-1+X.X1|-1+X.X2

    ## The main function of correcting measurement error and misclassification
    Re <- augSIMEX(mainformula = Y~Xstar.X1+Xstar.X2+Zstar+W.W1+W.W2,
      pimodel=pi|qi~1, family = binomial,
      data=Data$Main,meformula=meformula,
      validationdata=Data$Validation, subset=NULL,
      err.var=c("Xstar.X1","Xstar.X2"), mis.var="Zstar",
      err.true=c("X.X1","X.X2"), mis.true="Z", err.mat = NULL,
      lambda=NULL, M=5, B=20, nBoot=50, extrapolation=c("both"))
    ## Results without accounting for measurement error and misclassification
    naive <- glm(Y~Xstar.X1+Xstar.X2+Zstar+W.W1+W.W2,
      family = binomial,data=Data$Main)
    coefficients <- Re$coefficients
    se <- (Re$vcov)
  }, error = function(e) return(NA))
  end <- proc.time()-start

  return(list(coefs.naive=naive$coefficients,
    coefs.linear=coefficients[(length(coefficients)/2+1):length(coefficients)],
    coefs.quadra=coefficients[1:(length(coefficients)/2)],
    se.naive=sqrt(diag(vcov(naive))),
    se.linear=se[(length(coefficients)/2+1):length(coefficients)],
    se.quadra=se[1:(length(coefficients)/2)],
    time=end[1]))
}

```

```

Results_Table2_RM <- function(i){
  tryCatch({
    start <- proc.time()
    Data <- DataGenerator(Seeds[i],0.5,"RM")

    ## The main function of correcting measurement error and misclassification

```

```

Re <- augSIMEX(mainformula = Y~Xstar1+Zstar+W, family = binomial(link=logit),
  data=Data$Main,validationdata=Data$Validation, subset=NULL,
  err.var="Xstar1", mis.var="Zstar", err.true="X",
  mis.true="Z", err.mat = NULL,
  repeated = TRUE,repind=list(Xstar1=c("Xstar1","Xstar2")),
  lambda=NULL, M=5, B=20, nBoot=50, extrapolation=c("both"))

## Results without accounting for measurement error and misclassification
naive <- glm(Y~Xstar1+Zstar+W,family = binomial,data=Data$Main)
coefficients <- Re$coefficients
se <- (Re$vcov)
}, error = function(e) return(NA))
end <- proc.time()-start

return(list(coefs.naive=naive$coefficients,
  coefs.linear=coefficients[(length(coefficients)/2+1):length(coefficients)],
  coefs.quadra=coefficients[1:(length(coefficients)/2)],
  se.naive=sqrt(diag(vcov(naive))),
  se.linear=se[(length(coefficients)/2+1):length(coefficients)],
  se.quadra=se[1:(length(coefficients)/2)],
  time=end[1]))
}

Table2_1D <- lapply(1:1000,FUN=Results_Table2_1D)
Table2_2D <- lapply(1:1000,FUN=Results_Table2_2D)
Table2_RM <- lapply(1:1000,FUN=Results_Table2_RM)

#####
## Evaluation and rearrangement of obtaining Table 2
## Function of computing performance results from linear
## and quadratic extrepolation
data.var.order <- c("(Intercept)","X","W","Z")
calibration.var.order <- c("(Intercept)","X","Z","W")
coef.object <- "coefs.naive"
se.object <- "se.naive"

get_Table2_unit <- function(data,truebeta,coef.object,se.object,

```

```

                                data.var.order,calibration.var.order){
#### Rearrange the data of naive study
## Bias
beta_pooled <- unlist(lapply(data,FUN = function(x){
  if (is.na(x)) return(NA) else return(x[[coef.object]]))}))
beta_matrix <- matrix(beta_pooled,nrow=1000,byrow = T)
colnames(beta_matrix) <- data.var.order
beta_matrix <- beta_matrix[,calibration.var.order]
mean <- colMeans(beta_matrix,na.rm = T)
bias <- abs(mean-truebeta)

## Standard Error
se_pooled <- unlist(lapply(data,FUN = function(x){
  if (is.na(x)) return(NA) else return(x[[se.object]]))}))
se_matrix <- matrix(se_pooled,nrow=1000,byrow = T)
colnames(se_matrix) <- data.var.order
se_matrix <- se_matrix[,calibration.var.order]
se <- colMeans(se_matrix,na.rm = T)

## Coverage Rate
coverage <- lapply(1:1000,FUN = function(i){
  if (is.na(data[[i]])) return(NA) else{
    LB <- data[[i]][[coef.object]]-1.96*data[[i]][[se.object]]
    UB <- data[[i]][[coef.object]]+1.96*data[[i]][[se.object]]
    names(LB) <- names(UB) <- data.var.order
    LB <- LB[calibration.var.order] # reorder upper bound
    UB <- UB[calibration.var.order] # reorder lower bound
    return(truebeta>=LB & truebeta<=UB)
  }
})
CR <- colMeans(matrix(unlist(coverage),nrow=1000,byrow=T),na.rm=T)

output <- data.frame(bias=round(bias,3),se=round(se,3),CR=percent(CR))
return(as.matrix(output))
}

True_beta_1D <- c(0.1,-1,0.7,0.5)
Table2_1D_naive <- get_Table2_unit(Table2_1D, True_beta_1D,

```

```

      "coefs.naive", "se.naive",
      data.var.order = c("(Intercept)", "X", "W", "Z"),
      calibration.var.order = c("(Intercept)", "X", "Z", "W"))
Table2_1D_linear <- get_Table2_unit(Table2_1D, True_beta_1D,
      "coefs.linear", "se.linear",
      data.var.order = c("X", "(Intercept)", "W", "Z"),
      calibration.var.order = c("(Intercept)", "X", "Z", "W"))
Table2_1D_quadra <- get_Table2_unit(Table2_1D, True_beta_1D,
      "coefs.quadra", "se.quadra",
      data.var.order = c("X", "(Intercept)", "W", "Z"),
      calibration.var.order = c("(Intercept)", "X", "Z", "W"))

True_beta_2D <- c(0.1, -1, 0.6, 0.7, -0.5, 1)
Table2_2D_naive <- get_Table2_unit(Table2_2D, True_beta_2D,
      "coefs.naive", "se.naive",
      data.var.order = c("(Intercept)", "X1", "X2", "Z", "W1", "W2"),
      calibration.var.order = c("(Intercept)", "X1", "X2", "Z", "W1", "W2"))
Table2_2D_linear <- get_Table2_unit(Table2_2D, True_beta_2D,
      "coefs.linear", "se.linear",
      data.var.order = c("X1", "X2", "(Intercept)", "W1", "W2", "Z"),
      calibration.var.order = c("(Intercept)", "X1", "X2", "Z", "W1", "W2"))
Table2_2D_quadra <- get_Table2_unit(Table2_2D, True_beta_2D,
      "coefs.quadra", "se.quadra",
      data.var.order = c("X1", "X2", "(Intercept)", "W1", "W2", "Z"),
      calibration.var.order = c("(Intercept)", "X1", "X2", "Z", "W1", "W2"))

True_beta_1D <- c(0.1, -1, 0.7, 0.5)
Table2_RM_naive <- get_Table2_unit(Table2_RM, True_beta_1D,
      "coefs.naive", "se.naive",
      data.var.order = c("(Intercept)", "X", "W", "Z"),
      calibration.var.order = c("(Intercept)", "X", "Z", "W"))
Table2_RM_linear <- get_Table2_unit(Table2_RM, True_beta_1D,
      "coefs.linear", "se.linear",
      data.var.order = c("X", "(Intercept)", "W", "Z"),
      calibration.var.order = c("(Intercept)", "X", "Z", "W"))
Table2_RM_quadra <- get_Table2_unit(Table2_RM, True_beta_1D,
      "coefs.quadra", "se.quadra",
      data.var.order = c("X", "(Intercept)", "W", "Z"),
      calibration.var.order = c("(Intercept)", "X", "Z", "W"))

```

```

Table2 <- cbind(rbind(Table2_1D_naive,Table2_2D_naive,Table2_RM_naive),
               rbind(Table2_1D_linear,Table2_2D_linear,Table2_RM_linear),
               rbind(Table2_1D_quadra,Table2_2D_quadra,Table2_RM_quadra))
Table2

#####
## Section 5.3 variance estimation ##
## This Section will run approximately [6 hours] without parallel computing
Results_Table3_probit_parallel<-function(i){
  tryCatch({
    library(augSIMEX)
    source("DataGenerator.R")

    set.seed(2018)
    Seeds<-sample(100000,size=1000,replace=FALSE)

    start<-proc.time()
    Data <- DataGenerator2(Seeds[i],0.2,"1D",nm=1000,nv=500,link="probit")

    # truemodel <- glm(Y~X1+Z+W1,family = binomial(link=probit))
    ## The main function of correcting measurement error and misclassification
    Re <- augSIMEX(mainformula = Y~Xstar+W+Zstar, mismodel=pi|qi~1,
                  family = binomial(link=probit),
                  data=Data$Main,validationdata=Data$Validation, subset=NULL,
                  err.var="Xstar", mis.var="Zstar", err.true="X", mis.true="Z",
                  err.mat = NULL,
                  lambda=NULL, M=5, B=20, nBoot=50, extrapolation=c("both"),
                  solvefun = "nleqslv")

    ## Results without accounting for measurement error and misclassification
    naive<-glm(Y~Xstar+W+Zstar,family = binomial(link=probit),data=Data$Main)
    end<-proc.time()-start

    coefficients <- Re$coefficients
    se <- Re$vcov
    return(list(coefs.naive=naive$coefficients,

```

```

        coefs.linear=coefficients[(length(coefficients)/2+1):length(coefficients)],
        coefs.quadra=coefficients[1:(length(coefficients)/2)],
        se.naive=sqrt(diag(vcov(naive))),
        se.linear=se[(length(coefficients)/2+1):length(coefficients)],
        se.quadra=se[1:(length(coefficients)/2)],
        time=end[1]))
  }, error = function(e) return(NA))
}

```

```

Results_Table3_poisson_parallel<-function(i){
  tryCatch({
    library(augSIMEX)
    source("DataGenerator.R")

    set.seed(2018)
    Seeds<-sample(100000,size=1000,replace=FALSE)

    start<-proc.time()
    Data <- DataGenerator2(Seeds[i],0.2,"1D",nm=1000,nv=500,link="log")
    true <- glm(Y~X1+W1+Z,family = poisson(link=log),data=Data$TrueData)

    ## The main function of correcting measurement error and misclassification
    Re <- augSIMEX(mainformula = Y~Xstar+W+Zstar, mismodel=pi|qi~1,
      family = poisson(link=log),
      data=Data$Main,validationdata=Data$Validation, subset=NULL,
      err.var="Xstar", mis.var="Zstar", err.true="X", mis.true="Z",
      err.mat = NULL,
      lambda=NULL, M=15, B=20, nBoot=50, extrapolation=c("both"),
      solvefun = "nleqslv")
    ## Results without accounting for measurement error and misclassification
    naive<-glm(Y~Xstar+W+Zstar,family = poisson(link=log),data=Data$Main)
    end<-proc.time()-start

    coefficients <- Re$coefficients
    se <- Re$vcov
    return(list(coefs.naive=naive$coefficients,
      coefs.linear=coefficients[(length(coefficients)/2+1):length(coefficients)],

```

```

        coefs.quadra=coefficients[1:(length(coefficients)/2)],
        se.naive=sqrt(diag(vcov(naive))),
        se.linear=se[(length(coefficients)/2+1):length(coefficients)],
        se.quadra=se[1:(length(coefficients)/2)],
        time=end[1]))
    }, error = function(e) return(NA))
}

Results_Table3_probit_repeat<-function(i){
  tryCatch({
    library(augSIMEX)
    source("DataGenerator.R")

    set.seed(2018)
    Seeds<-sample(100000,size=1000,replace=FALSE)

    start<-proc.time()
    Data <- DataGenerator2(Seeds[i],0.2,"RM",nm=1000,nv=500,link="probit")

    # truemodel <- glm(Y~X1+Z+W1,family = binomial(link=probit))
    ## The main function of correcting measurement error and misclassification
    Re <- augSIMEX(mainformula = Y~Xstar1+Zstar+W, family = binomial(link=probit),
      data=Data$Main,validationdata=Data$Validation, subset=NULL,
      err.var="Xstar1", mis.var="Zstar", err.true="X", mis.true="Z",
      err.mat = NULL,
      repeated = TRUE,repind=list(Xstar1=c("Xstar1","Xstar2")),
      lambda=NULL, M=5, B=20, nBoot=50, extrapolation=c("both"),
      solvefun ="nleqslv")
    ## Results without accounting for measurement error and misclassification
    naive<-glm(Y~Xstar1+W+Zstar,family = binomial(link=probit),data=Data$Main)
    end<-proc.time()-start

    coefficients <- Re$coefficients
    se <- Re$vcov
    return(list(coefs.naive=naive$coefficients,
      coefs.linear=coefficients[(length(coefficients)/2+1):length(coefficients)],
      coefs.quadra=coefficients[1:(length(coefficients)/2)],
      se.naive=sqrt(diag(vcov(naive))),

```

```

        se.linear=se[(length(coefficients)/2+1):length(coefficients)],
        se.quadra=se[1:(length(coefficients)/2)],
        time=end[1]))
}, error = function(e) return(NA))
}

Results_Table3_poisson_repeat<-function(i){
  tryCatch({
    library(augSIMEX)
    source("DataGenerator.R")

    set.seed(2018)
    Seeds<-sample(100000,size=1000,replace=FALSE)

    start<-proc.time()
    Data <- DataGenerator2(Seeds[i],0.2,"RM",nm=1000,nv=500,link="log")
    true <- glm(Y~X1+W1+Z,family = poisson(link=log),data=Data$TrueData)

    ## The main function of correcting measurement error and misclassification
    Re <- augSIMEX(mainformula = Y~Xstar1+Zstar+W, family = poisson(link=log),
                  data=Data$Main,validationdata=Data$Validation, subset=NULL,
                  err.var="Xstar1", mis.var="Zstar", err.true="X", mis.true="Z",
                  err.mat = NULL,
                  repeated = TRUE,repind=list(Xstar1=c("Xstar1","Xstar2")),
                  lambda=NULL, M=5, B=20, nBoot=50, extrapolation=c("both"),
                  solvefun ="nleqslv")

    ## Results without accounting for measurement error and misclassification
    naive<-glm(Y~Xstar1+W+Zstar,family = poisson(link=log),data=Data$Main)
    end<-proc.time()-start

    coefficients <- Re$coefficients
    se <- Re$vcov
    return(list(coefs.naive=naive$coefficients,
              coefs.linear=coefficients[(length(coefficients)/2+1):length(coefficients)],
              coefs.quadra=coefficients[1:(length(coefficients)/2)],
              se.naive=sqrt(diag(vcov(naive))),
              se.linear=se[(length(coefficients)/2+1):length(coefficients)],
              se.quadra=se[1:(length(coefficients)/2)],
              time=end[1]))
  }, error = function(e) return(NA))
}

```

```

    }, error = function(e) return(NA))
}

```

```

Table3_pr <- parLapply(cl, 1:1000, Results_Table3_probit_parallel)
save(Table3_pr,file="Table3_pr_nle2.RData")
Table3_po <- parLapply(cl, 1:1000, Results_Table3_poisson_parallel)
save(Table3_po,file="Table3_po_nle2.RData")
Table3rep_pr <- parLapply(cl, 1:1000, Results_Table3_probit_repeat)
save(Table3rep_pr,file="Table3_pr_rep.RData")
Table3rep_po <- parLapply(cl, 1:1000, Results_Table3_poisson_repeat)
save(Table3rep_po,file="Table3_po_rep.RData")

```

```
#####
```

```

## Evaluation and rearrangement of obtaining Table 3
## Function of computing performance results from linear and quadratic extrepolation
data.var.order <- c("(Intercept)","X","W","Z")
calibration.var.order <- c("(Intercept)","X","Z","W")
coef.object <- "coefs.naive"
se.object <- "se.naive"

```

```

get_Table3_unit <- function(data,truebeta,coef.object,se.object,
                             data.var.order,calibration.var.order){
  #### Rearrange the data of naive study
  ## Bias
  beta_pooled <- unlist(lapply(data,FUN = function(x){
    if (is.na(x)) return(rep(NA,length(data.var.order))) else return(x[[coef.object]]))})
  beta_matrix <- matrix(beta_pooled,nrow=1000,byrow = T)
  colnames(beta_matrix) <- data.var.order
  beta_matrix <- beta_matrix[,calibration.var.order]
  mean <- colMeans(beta_matrix,na.rm = T)
  bias <- abs(mean-truebeta)

  ## Standard Error
  se_pooled <- unlist(lapply(data,FUN = function(x){

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    if (is.na(x)) return(rep(NA,length(data.var.order))) else return(x[[se.object]]))}
se_matrix <- matrix(se_pooled,nrow=1000,byrow = T)
colnames(se_matrix) <- data.var.order
se_matrix <- se_matrix[,calibration.var.order]
se <- colMeans(se_matrix,na.rm = T)

## Coverage Rate
coverage <- lapply(1:1000,FUN = function(i){
  if (is.na(data[[i]])) return(rep(NA,length(data.var.order))) else{
    LB <- data[[i]][[coef.object]]-1.96*data[[i]][[se.object]]
    UB <- data[[i]][[coef.object]]+1.96*data[[i]][[se.object]]
    names(LB) <- names(UB) <- data.var.order
    LB <- LB[calibration.var.order] # reorder upper bound
    UB <- UB[calibration.var.order] # reorder lower bound
    return(truebeta>=LB & truebeta<=UB)
  }
})
CR <- colMeans(matrix(unlist(coverage),nrow=1000,byrow=T),na.rm=T)

output <- data.frame(bias=round(bias,3),se=round(se,3),CR=percent(CR))
return(as.matrix(output))
}

True_beta <- c(0.1,-1,0.7,0.5)
Table3_ga_naive <- get_Table3_unit(Table3_ga, True_beta,"coefs.naive","se.naive",
                                   data.var.order = c("(Intercept)","X","W","Z"),
                                   calibration.var.order = c("(Intercept)","X","Z","W"))
Table3_ga_linear <- get_Table3_unit(Table3_ga, True_beta, "coefs.linear","se.linear",
                                   data.var.order = c("X","(Intercept)","W","Z"),
                                   calibration.var.order = c("(Intercept)","X","Z","W"))
Table3_ga_quadra <- get_Table3_unit(Table3_ga, True_beta, "coefs.quadra","se.quadra",
                                   data.var.order = c("X","(Intercept)","W","Z"),
                                   calibration.var.order = c("(Intercept)","X","Z","W"))

Table3_pr_naive <- get_Table3_unit(Table3_pr, True_beta,"coefs.naive","se.naive",
                                   data.var.order = c("(Intercept)","X","W","Z"),
                                   calibration.var.order = c("(Intercept)","X","Z","W"))
Table3_pr_linear <- get_Table3_unit(Table3_pr, True_beta, "coefs.linear","se.linear",

```

```

data.var.order = c("X","(Intercept)","W","Z"),
calibration.var.order = c("(Intercept)","X","Z","W"))
Table3_pr_quadra <- get_Table3_unit(Table3_pr, True_beta, "coefs.quadra","se.quadra",
data.var.order = c("X","(Intercept)","W","Z"),
calibration.var.order = c("(Intercept)","X","Z","W"))

Table3rep_pr_naive <- get_Table3_unit(Table3rep_pr, True_beta,"coefs.naive","se.naive",
data.var.order = c("(Intercept)","Xstar1","W","Zstar"),
calibration.var.order = c("(Intercept)","Xstar1","Zstar","W"))
Table3rep_pr_linear <- get_Table3_unit(Table3rep_pr, True_beta, "coefs.linear","se.linear",
data.var.order = c("X","(Intercept)","W","Z"),
calibration.var.order = c("(Intercept)","X","Z","W"))
Table3rep_pr_quadra <- get_Table3_unit(Table3rep_pr, True_beta, "coefs.quadra","se.quadra",
data.var.order = c("X","(Intercept)","W","Z"),
calibration.var.order = c("(Intercept)","X","Z","W"))

Table3_po_naive <- get_Table3_unit(Table3_po, True_beta,"coefs.naive","se.naive",
data.var.order = c("(Intercept)","X","W","Z"),
calibration.var.order = c("(Intercept)","X","Z","W"))
Table3_po_linear <- get_Table3_unit(Table3_po, True_beta, "coefs.linear","se.linear",
data.var.order = c("X","(Intercept)","W","Z"),
calibration.var.order = c("(Intercept)","X","Z","W"))
Table3_po_quadra <- get_Table3_unit(Table3_po, True_beta, "coefs.quadra","se.quadra",
data.var.order = c("X","(Intercept)","W","Z"),
calibration.var.order = c("(Intercept)","X","Z","W"))

Table3rep_po_naive <- get_Table3_unit(Table3rep_po, True_beta,"coefs.naive","se.naive",
data.var.order = c("(Intercept)","Xstar1","W","Zstar"),
calibration.var.order = c("(Intercept)","Xstar1","Zstar","W"))
Table3rep_po_linear <- get_Table3_unit(Table3rep_po, True_beta, "coefs.linear","se.linear",
data.var.order = c("X","(Intercept)","W","Z"),
calibration.var.order = c("(Intercept)","X","Z","W"))
Table3rep_po_quadra <- get_Table3_unit(Table3rep_po, True_beta, "coefs.quadra","se.quadra",
data.var.order = c("X","(Intercept)","W","Z"),
calibration.var.order = c("(Intercept)","X","Z","W"))

Table3_po_quadra_new <- ifelse(abs(Table3_po_quadra)<8,Table3_po_quadra,NA)
mean <- colMeans(Table3_po_quadra_new,na.rm = T)

```

```
Table3 <- cbind(rbind(Table3_pr_naive,Table3rep_pr_naive,Table3_po_naive,Table3rep_po_naive),
  rbind(Table3_pr_linear,Table3rep_pr_linear,Table3_po_linear,Table3rep_po_linear),
  rbind(Table3_pr_quadra,Table3rep_pr_quadra,Table3_po_quadra,Table3rep_po_quadra))
```

```
#####
```

```
## Section 6 Example ##
```

```
data(GeneUni)
```

```
example1_naive <- lm(rs38916331 ~ Tibia_M + Batch_0 + Weight, data = GeneUni$Main)
summary(example1_naive)
```

```
# Call:
```

```
# lm(formula = rs38916331 ~ Tibia_M + Batch_0 + Weight, data = GeneUni$Main)
```

```
#
```

```
# Residuals:
```

```
#   Min       1Q   Median       3Q      Max
# -0.8754 -0.8588  0.1337  0.1423  1.1460
```

```
#
```

```
# Coefficients:
```

```
#   Estimate Std. Error t value Pr(>|t|)
# (Intercept)  0.8994158  0.8581227   1.048   0.295
# Tibia_M      0.0002329  0.0489492   0.005   0.996
# Batch_0      0.0009394  0.0681910   0.014   0.989
# Weight      -0.0015747  0.0093007  -0.169   0.866
```

```
#
```

```
# Residual standard error: 0.6859 on 750 degrees of freedom
```

```
# Multiple R-squared:  4.314e-05, Adjusted R-squared:  -0.003957
```

```
# F-statistic: 0.01079 on 3 and 750 DF, p-value: 0.9985
```

```
set.seed(2018)
```

```
example1<-augSIMEX(mainformula = rs38916331 ~ Tibia_M + Batch_0 + Weight,
  family = gaussian(link = "identity"),
  mismodel = pilqi ~ 1, meformula = Tibia_M ~ Tibia_T,
  data = GeneUni$Main, validationdata = GeneUni$Validation, subset = NULL,
  err.var = "Tibia_M", mis.var = "Batch_0", err.true = "Tibia_T",
```

```

        mis.true = "Batch_T", err.mat = NULL, solvefun = "nleqslv",
        lambda = NULL, M = 10, B = 20, nBoot = 50, extrapolation="quadratic",
        initial = c(0,0,0,0,0.01))

summary(example1)

# Call:
#   augSIMEX(mainformula = rs38916331 ~ Tibia_M + Batch_0 + Weight,
#             mismodel = pi | qi ~ 1, meformula = Tibia_M ~ Tibia_T,
#             family = gaussian(link = "identity"),
#             data = GeneUni$Main, validationdata = GeneUni$Validation,
#             err.var = "Tibia_M", mis.var = "Batch_0", err.true = "Tibia_T",
#             mis.true = "Batch_T", err.mat = NULL, subset = NULL, lambda = NULL,
#             M = 10, B = 20, nBoot = 50, extrapolation = "quadratic",
#             solvefun = "nleqslv", initial = c(0, 0, 0, 0, 0.01))
#
#
# The error-prone variable: Tibia_M
# The corresponding true variable: Tibia_T
# Number of simulations: 20
# Number of iterations in bootstrap: 50
#
#
# Coefficients:
#   Estimate Std. Error t value Pr(>|t|)
# (Intercept)  0.988156   1.041682   0.949   0.343
# Tibia_T      -0.003697   0.061068  -0.061   0.952
# Weight       -0.001831   0.012590  -0.145   0.884
# Batch_T       0.002227   0.143170   0.016   0.988
#
#
# (Dispersion parameter for gaussian family taken to be NULL)
#
# Null deviance: 352.89  on 753  degrees of freedom
# Residual deviance: 352.96  on 750  degrees of freedom
# AIC: 1577.4

plot(example1,ylim = c(-0.01,0.01) )

```

```

#### Example 2: Repeated Measurements
set.seed(2018)
data(GeneRepeat)
example2<-augSIMEX(mainformula = rs223979909 ~ Totdist + Batch_0 + Weight,
                    family = gaussian(link = "identity"),
                    mismodel = pi|qi ~ 1,
                    data = GeneRepeat$Main, validationdata = GeneRepeat$Validation,
                    err.var = "Totdist", mis.var = "Batch_0", err.true = "Tibia_T",
                    mis.true = "Batch_T", err.mat = NULL, repeated = TRUE,
                    repind= list(Totdist=c("Totdist5" , "Totdist10" , "Totdist15" ,
                                           "Totdist20", "Totdist25" , "Totdist30")),
                    lambda = NULL, M = 10, B = 20, nBoot = 50, extrapolation="quadratic",
                    solvefun = "nleqslv", initial = c(0,0,0,0,0.01))
summary(example2)

# Call:
# augSIMEX(mainformula = rs223979909 ~ Totdist + Batch_0 + Weight,
#           mismodel = pi | qi ~ 1, family = gaussian(link = "identity"),
#           data = GeneRepeat$Main, validationdata = GeneRepeat$Validation,
#           err.var = "Totdist", mis.var = "Batch_0", err.true = "Tibia_T",
#           mis.true = "Batch_T", err.mat = NULL, repeated = TRUE,
#           repind = list(Totdist = c("Totdist5",
#                                     "Totdist10", "Totdist15", "Totdist20", "Totdist25", "Totdist30")),
#           lambda = NULL, M = 10, B = 20, nBoot = 50, extrapolation = "quadratic",
#           solvefun = "nleqslv", initial = c(0, 0, 0, 0, 0.01))
#
#
# The error-prone variable: Totdist
# The corresponding true variable: Tibia_T
# Number of simulations: 20
# Number of iterations in bootstrap: 50
#
#
# Coefficients:
# Estimate Std. Error t value Pr(>|t|)
# (Intercept) 7.479e-01 2.717e-01 2.752 0.00592 **
# Tibia_T     -4.265e-05 2.803e-05 -1.521 0.12819

```

```

# Weight      -5.465e-03  9.047e-03  -0.604  0.54577
# Batch_T     1.227e-01  9.853e-02   1.245  0.21304
# ---
#   Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
#
#
# (Dispersion parameter for gaussian family taken to be NULL)
#
# Null deviance: 255.04  on 671  degrees of freedom
# Residual deviance: 270.87  on 668  degrees of freedom
# AIC: 1306.5

#####
## Section 7 Discussion          ##
Data <- DataGenerator(Seeds[1],0.5,"1D")

## The main function of correcting measurement error and misclassification
Re_cpp <- augSIMEX(mainformula = Y~Xstar+W+Zstar, mismodel=pi|qi~1, family = binomial,
                  data=Data$Main,validationdata=Data$Validation, subset=NULL,
                  err.var="Xstar", mis.var="Zstar", err.true="X", mis.true="Z", err.mat = NULL,
                  lambda=NULL, M=3, B=3, nBoot=3, extrapolation=c("both"))
Re_R <- augSIMEX(mainformula = Y~Xstar+W+Zstar, mismodel=pi|qi~1, family = binomial,
                 data=Data$Main,validationdata=Data$Validation, subset=NULL,
                 err.var="Xstar", mis.var="Zstar", err.true="X", mis.true="Z", err.mat = NULL,
                 lambda=NULL, M=3, B=3, nBoot=3, extrapolation=c("both"),cppmethod = F)

speedcompare<-microbenchmark(
  cpp=cpp <- augSIMEX(mainformula = Y~Xstar+W+Zstar,
                     mismodel=pi|qi~1, family = binomial,
                     data=Data$Main,validationdata=Data$Validation, subset=NULL,
                     err.var="Xstar", mis.var="Zstar", err.true="X", mis.true="Z",
                     err.mat = NULL, lambda=NULL, M=2, B=2, nBoot=2,
                     extrapolation=c("both")),
  R=R <- augSIMEX(mainformula = Y~Xstar+W+Zstar, mismodel=pi|qi~1, family = binomial,
                 data=Data$Main,validationdata=Data$Validation, subset=NULL,
                 err.var="Xstar", mis.var="Zstar", err.true="X", mis.true="Z",
                 err.mat = NULL,

```

```

        lambda=NULL, M=2, B=2, nBoot=2, extrapolation=c("both"),
        cppmethod = F), times = 50)
if (require("ggplot2")) {
  autoplot(speedcompare)
}

## figure 2: speed and the simulation parameter
# c("M","B","nBoot","type","parameter","time")

timerecord<-NULL
Data <- DataGenerator(Seeds[1],0.5,"1D")

for (N in 1:200){
  for (i in c(5,10,15,20,25)){
    start <- proc.time()
    R <- augSIMEX(mainformula = Y~Xstar+W+Zstar, mismodel=pi|qi~1, family = binomial,
                  data=Data$Main,validationdata=Data$Validation, subset=NULL,
                  err.var="Xstar", mis.var="Zstar", err.true="X", mis.true="Z",
                  err.mat = NULL,
                  lambda=NULL, M=i, B=5, nBoot=5, extrapolation=c("both"))
    end <- proc.time() - start
    timerecord<-rbind(timerecord,c(i,5,5,"1D","M",end[1]))
    start <- proc.time()
    R <- augSIMEX(mainformula = Y~Xstar+W+Zstar, mismodel=pi|qi~1, family = binomial,
                  data=Data$Main,validationdata=Data$Validation, subset=NULL,
                  err.var="Xstar", mis.var="Zstar", err.true="X", mis.true="Z",
                  err.mat = NULL,
                  lambda=NULL, M=5, B=i, nBoot=5, extrapolation=c("both"))
    end <- proc.time() - start
    timerecord<-rbind(timerecord,c(5,i,5,"1D","B",end[1]))
    start <- proc.time()
    R <- augSIMEX(mainformula = Y~Xstar+W+Zstar, mismodel=pi|qi~1, family = binomial,
                  data=Data$Main,validationdata=Data$Validation, subset=NULL,
                  err.var="Xstar", mis.var="Zstar", err.true="X", mis.true="Z",
                  err.mat = NULL,
                  lambda=NULL, M=5, B=5, nBoot=i, extrapolation=c("both"))
    end <- proc.time() - start
    timerecord<-rbind(timerecord,c(5,5,i,"1D","nBoot",end[1]))
  }
}

```

```
}
```

```
Data <- DataGenerator(Seeds[1],0.5,"2D")
meformula <- Xstar.X1|Xstar.X2~-1+X.X1|-1+X.X2
for (N in 1:200){
  for (i in c(5,10,15,20,25)){
    start <- proc.time()
    Re<-augSIMEX(mainformula = Y~Xstar.X1+Xstar.X2+Zstar+W.W1+W.W2,
                  pimodel=pi~X.X1+X.X2+W.W1+W.W2,
                  qimodel=qi~X.X1+X.X2+W.W1+W.W2, family = binomial,
                  data=Data$Main,meformula=meformula,validationdata=Data$Validation,
                  subset=NULL,
                  err.var=c("Xstar.X1","Xstar.X2"), mis.var="Zstar",
                  err.true=c("X.X1","X.X2"), mis.true="Z", err.mat = NULL,
                  lambda=NULL, M=i, B=5, nBoot=5, extrapolation=c("both"))
    end <- proc.time() - start
    timerecord<-rbind(timerecord,c(i,5,5,"2D","M",end[1]))
    start <- proc.time()
    Re<-augSIMEX(mainformula = Y~Xstar.X1+Xstar.X2+Zstar+W.W1+W.W2,
                  pimodel=pi~X.X1+X.X2+W.W1+W.W2,
                  qimodel=qi~X.X1+X.X2+W.W1+W.W2, family = binomial,
                  data=Data$Main,meformula=meformula,validationdata=Data$Validation,
                  subset=NULL,
                  err.var=c("Xstar.X1","Xstar.X2"), mis.var="Zstar",
                  err.true=c("X.X1","X.X2"), mis.true="Z", err.mat = NULL,
                  lambda=NULL, M=5, B=i, nBoot=5, extrapolation=c("both"))
    end <- proc.time() - start
    timerecord<-rbind(timerecord,c(5,i,5,"2D","B",end[1]))
    start <- proc.time()
    Re<-augSIMEX(mainformula = Y~Xstar.X1+Xstar.X2+Zstar+W.W1+W.W2,
                  pimodel=pi~X.X1+X.X2+W.W1+W.W2,
                  qimodel=qi~X.X1+X.X2+W.W1+W.W2, family = binomial,
                  data=Data$Main,meformula=meformula,validationdata=Data$Validation,
                  subset=NULL,
                  err.var=c("Xstar.X1","Xstar.X2"), mis.var="Zstar",
                  err.true=c("X.X1","X.X2"), mis.true="Z", err.mat = NULL,
                  lambda=NULL, M=5, B=5, nBoot=i, extrapolation=c("both"))
    end <- proc.time() - start
    timerecord<-rbind(timerecord,c(5,5,i,"2D","nBoot",end[1]))
```

```

    }
  }

Data <- DataGenerator(Seeds[1],0.5,"RM")
for (N in 1:200){
  for (i in c(5,10,15,20,25)){
    start <- proc.time()
    Re <- augSIMEX(mainformula = Y~Xstar1+Zstar+W, family = binomial(link=logit),
                  data=Data$Main,validationdata=Data$Validation, subset=NULL,
                  err.var="Xstar1", mis.var="Zstar", err.true="X", mis.true="Z",
                  err.mat = NULL,
                  repeated = TRUE,repind=list(Xstar1=c("Xstar1","Xstar2")),
                  lambda=NULL, M=i, B=5, nBoot=5, extrapolation=c("both"))
    end <- proc.time() - start
    timerecord<-rbind(timerecord,c(i,5,5,"RM","M",end[1]))
    start <- proc.time()
    Re <- augSIMEX(mainformula = Y~Xstar1+Zstar+W, family = binomial(link=logit),
                  data=Data$Main,validationdata=Data$Validation, subset=NULL,
                  err.var="Xstar1", mis.var="Zstar", err.true="X", mis.true="Z",
                  err.mat = NULL,
                  repeated = TRUE,repind=list(Xstar1=c("Xstar1","Xstar2")),
                  lambda=NULL, M=5, B=i, nBoot=5, extrapolation=c("both"))
    end <- proc.time() - start
    timerecord<-rbind(timerecord,c(5,i,5,"RM","B",end[1]))
    start <- proc.time()
    Re <- augSIMEX(mainformula = Y~Xstar1+Zstar+W, family = binomial(link=logit),
                  data=Data$Main,validationdata=Data$Validation, subset=NULL,
                  err.var="Xstar1", mis.var="Zstar", err.true="X", mis.true="Z",
                  err.mat = NULL,
                  repeated = TRUE,repind=list(Xstar1=c("Xstar1","Xstar2")),
                  lambda=NULL, M=5, B=5, nBoot=i, extrapolation=c("both"))
    end <- proc.time() - start
    timerecord<-rbind(timerecord,c(5,5,i,"RM","nBoot",end[1]))
  }
}

TimeSheet <- data.frame(M=as.numeric(timerecord[,1]),
                        B=as.numeric(timerecord[,2]),
                        nBoot=as.numeric(timerecord[,3]),

```

```

        type=timerecord[,4],
        parameter=timerecord[,5],
        time=as.numeric(timerecord[,6]))

TimeSheetagg <- aggregate(TimeSheet$time,by=list(TimeSheet$M,TimeSheet$B,
        TimeSheet$nBoot,TimeSheet$parameter,TimeSheet$type),FUN=mean)
TimeSheetagg <- data.frame(TimeSheetagg, choice = rep(c(5,10,15,20,25),9))
colnames(TimeSheetagg) <- c("M","B","T","parameter","study","time","choice")
# save(TimeSheetagg,file="TimeSheet.Rdata")
TimeSheetagg$parameter <- as.character(TimeSheetagg$parameter)
TimeSheetagg[TimeSheetagg$parameter=="nBoot","parameter"] <-
        rep("T",length(TimeSheetagg[TimeSheetagg$parameter=="nBoot","parameter"]))

g <- ggplot( TimeSheetagg , aes(x=choice, y= time, colour = study)) +
        geom_line(aes(linetype=study),size=0.8)+
        scale_colour_brewer(palette = "Set1") +
        scale_linetype_manual(values=c("twodash", "dotted","solid")) +
        facet_wrap(~parameter)+
        theme(text=element_text(size=13, family="mono"))

cbPalette <- c("#56B4E9","#999999", "#E69F00")

ggplotly(g+ scale_color_manual(values=cbPalette))

## figure 3: speed, sample size
timerecord2<-NULL
for (N in 1:100){
  for (i in c(300+0:12*100)){
    for (j in c(500+0:10*50)){
      Data <- DataGenerator2(Seeds[1],0.5,"1D",nm=i,nv=j,link="logit")
      start <- proc.time()
      R <- augSIMEX(mainformula = Y~Xstar+W+Zstar, mismodel=pi|qi~1, family = binomial,
        data=Data$Main,validationdata=Data$Validation, subset=NULL,
        err.var="Xstar", mis.var="Zstar", err.true="X", mis.true="Z",
        err.mat = NULL,lambd=0, M=5, B=5, nBoot=5, extrapolation=c("both"))
    }
  }
}

```

```

end <- proc.time() - start
timerecord2<-rbind(timerecord2,c(i,j,"1D",end[1]))
Data <- DataGenerator2(Seeds[1],0.5,"2D",nm=i,nv=j,link="logit")
meformula <- Xstar.X1|Xstar.X2~-1+X.X1|-1+X.X2
start <- proc.time()
Re<-augSIMEX(mainformula = Y~Xstar.X1+Xstar.X2+Zstar+W.W1+W.W2,
             pimodel=pi~X.X1+X.X2+W.W1+W.W2,
             qimodel=qi~X.X1+X.X2+W.W1+W.W2, family = binomial,
             data=Data$Main,meformula=meformula,validationdata=Data$Validation,
             subset=NULL, err.var=c("Xstar.X1","Xstar.X2"), mis.var="Zstar",
             err.true=c("X.X1","X.X2"), mis.true="Z", err.mat = NULL,
             lambda=NULL, M=5, B=5, nBoot=5, extrapolation=c("both"))
end <- proc.time() - start
timerecord2<-rbind(timerecord2,c(i,j,"2D",end[1]))
Data <- DataGenerator2(Seeds[1],0.5,"RM",nm=i,nv=j,link="logit")
start <- proc.time()
Re <- augSIMEX(mainformula = Y~Xstar1+Zstar+W, family = binomial(link=logit),
             data=Data$Main,validationdata=Data$Validation, subset=NULL,
             err.var="Xstar1", mis.var="Zstar", err.true="X", mis.true="Z",
             err.mat = NULL, repeated = TRUE,
             repind=list(Xstar1=c("Xstar1","Xstar2")),
             lambda=NULL, M=5, B=5, nBoot=5, extrapolation=c("both"))
end <- proc.time() - start
timerecord2<-rbind(timerecord2,c(i,j,"RM",end[1]))
}
}
cat("N=",N,"\n")
}

TimeSheet <- data.frame(nm=as.numeric(timerecord2[,1]),
                      nv=as.numeric(timerecord2[,2]),
                      type=timerecord2[,3],
                      time=as.numeric(timerecord2[,4]))

TimeSheetagg2 <- aggregate(TimeSheet$time,by=list(TimeSheet$nm,TimeSheet$nv,
                                                TimeSheet$type),FUN=mean)
colnames(TimeSheetagg2) <- c("nm","nv","type","time")

# save(TimeSheetagg2,file="Sheetagg2.RData")

```

```

v <- ggplot(TimeSheetagg2, aes(nm, nv, z = time)) +
  facet_wrap(~type)+
  theme(text=element_text(size=13, family="mono"))+
  geom_raster(aes(fill = time)) +
  geom_contour(colour = "white")

ggplotly(v)

timerecord2<-matrix(timerecord2,nrow=7)
timerecord2<-rowMeans(timerecord2)

```