STAT-6494 Advanced Statistical Computing with R

Homework 5

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1 Exercise: A Simple Implementation of QIF.

Write a function simpleQIF that evaluates the QIF $Q_n(\beta)$ at regression coefficient vector β for a given response vector Y, covariate matrix X, an id vector that distinguishes the clusters, and an exchangeable given working correlation structure. For the example dataset ohio in package **geepack**, feed your function to an optimizer such as optim to solve for the parameter estimate. An example call would look like simpleQIF(y, X, id, family).

1.1 Function simpleQIF

Qu, Lindsay, and Li (2000) proposed qudratic inference function (QIF) to improve generalized estimating equations. Here, we consider logit as link function for binary outcomes as default and only assume exchangeable correlation structure. The simple implementation of QIF, simpleQIF is designed to return estimates of covariate coefficients. The function body is given as follows.

```
simpleQIF <- function (y, X, id, family = binomial) {</pre>
    X <- as.matrix(X)</pre>
    y <- as.vector(y)
    famFun <- family()</pre>
    ## balanced cluster size
    idTab <- as.numeric(table(id))</pre>
    clusN <- length(idTab)</pre>
    clusSize <- unique(idTab)</pre>
    if (length(clusSize) != 1) stop("Not balanced cluster size.")
    ## define matrix m_0 and matrix m_1
    ## for exchangeable structure
    mat0 <- diag(clusSize)</pre>
    mat1 <- matrix(1, clusSize, clusSize) - mat0</pre>
    matMList <- list(mat0, mat1)</pre>
    ## initialize beta as starting point
    fit0 <- glm.fit(X, y, family = famFun)
    beta0 <- fit0$coef
    ## optim step
    res <- optim(beta0, fn = qNbeta, gr = NULL,
                  clusSize = clusSize, clusN = clusN, y = y, X = X,
                  matMList = matMList, famFun = famFun)
    ## simply output coef estimates
    res$par
}
### internal functions
## computes g_ri(beta) for cluster i and matrix r
```

```
g_riBeta <- function (matM, betaVec, yi, Xi, famFun) {</pre>
    eta <- as.vector(Xi %*% betaVec, mode = "numeric")
    muiVec <- famFun$linkinv(eta)</pre>
    matDi <- famFun$mu.eta(muiVec) * Xi # element-wise</pre>
    sqrtInvAi <- diag(1 / sqrt(famFun$variance(muiVec)))</pre>
    crossprod(matDi, sqrtInvAi) %*% (matM %*% (sqrtInvAi %*% (yi - muiVec)))
}
## computes g_i(beta) for cluster i
g_iBeta <- function (betaVec, matMList, yi, Xi, famFun) {</pre>
    out <- lapply(matMList, g_riBeta, betaVec = betaVec, yi = yi,
                   Xi = Xi, famFun = famFun)
    do.call("c", out)
}
## computes QIF: Q_N(beta)
qNbeta <- function (betaVec, clusSize, clusN, y, X, matMList, famFun) {</pre>
    nMat <- length(matMList)</pre>
    nBeta <- length(betaVec)</pre>
    gNbeta <- matrix(NA, nrow = clusN, ncol = nMat * nBeta)</pre>
    indBase <- seq(1 - clusSize, 0)</pre>
    for (i in seq_len(clusN)) {
        ind <- indBase + i * clusSize</pre>
        Xi <- X[ind, ]</pre>
        yi <- y[ind]</pre>
        gNbeta[i, ] <- g_iBeta(betaVec, matMList, yi, Xi, famFun)</pre>
    gBarN <- colMeans(gNbeta)</pre>
    cNbeta <- var(gNbeta)</pre>
    clusN * crossprod(gBarN, MASS::ginv(cNbeta) %*% gBarN)
```

1.2 Comparison with geepack on a Sample Dataset

We compare the estimates of covariate coefficients produced by **geese** in **geepack** to verify our implementation of QIF on a sample dataset, *ohio*.

From the comparison, we may find that the estimates by simpleQIF and geese are quite close to each other. So we have confidence that the implementation is correct.

```
require(geepack)
data(ohio)
y <- ohio$resp
X <- cbind("(Intercept)" = 1, ohio[, 3:4])
id <- ohio$id
simpleQIF(y, X, id) # estimates from our implementation

(Intercept) age smoke
-1.8844759 -0.1189905 0.2637976

## compare with geepack
fitGee <- geese(resp ~ age + smoke, id = id, data = ohio,</pre>
```

```
family = binomial, corstr = "exch", scale.fix = TRUE)
fitGee$beta # estimates from geepack
```

```
(Intercept) age smoke
-1.8804288 -0.1133850 0.2650833
```

Reference

Qu
, Annie, Bruce G Lindsay, and Bing Li. 2000. "Improving Generalised Estimating Equations Using Quadratic Inference Functions."
 Biometrika 87 (4). Biometrika Trust: 823–36.