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getwd()
Series1 = read.csv("Series1.csv")
Series2 = read.csv("Series2.csv")
#Functions that are used in this assignment
getest = function(x, y, m, beta_ini, lambda_ini, ..., eps = 10^(-10)){
 beta = beta_ini
 lambda = lambda ini
 xi = c(beta, lambda)
  iter = 0
  while(TRUE){
   iter = iter + 1
   eta = x \% *\% beta
   mu = 1 - 1 / (lambda * exp(eta) + 1)^(1 / lambda)
   mu = ifelse(abs(mu) > eps, mu, eps)
   mu = ifelse(abs(1 - mu) > eps, mu, 1 - eps)
   eta_mu = lambda / (1 - mu) / (1 - (1 - mu)^lambda)
   w = m / (eta_mu^2 * mu * (1 - mu))
   eta_lambda = -\log(1 - mu) / (1 - (1 - mu)^lambda) - 1 / lambda
   X_A = cbind(x, -eta_lambda)
   W = diag(c(w))
   z = (y - mu) * eta_mu
   xi_2 = xi + solve(t(X_A) %*% W %*% X_A, t(X_A) %*% W %*% z)
   \#xi_2 = solve(t(X_A) \% \% \% \% \% X_A, t(X_A) \% \% \% \% \% (X_A \% \% xi + z))
   if(any(is.na(xi_2))){
      stop("Fatal error:: NA's generated")
   if(norm(xi - xi_2) < 10^(-10))
      break
   else{
      beta = c(xi_2)[1:2]
     lambda = c(xi_2)[3]
     xi = c(beta, lambda)
   }
 }
 return(list(beta = c(xi_2)[1:2], lambda = c(xi_2)[3], iteration = iter))
}
#Series = Series2
#link_ini = "cloglog"
glmfit = function(Series, link_ini){
 model_ini = glm(deathrate ~ logCS2, data = Series, family = binomial(link_ini), weights = Total)
  beta_ini = model_ini$coefficients
  if(link_ini == "logit") lambda_ini = 1
  if(link_ini == "cloglog") lambda_ini = 0.001
 x = cbind(1, Series$logCS2)
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y = Series$deathrate
  m = Series$Total
  eps = 10^{(-10)}
  res = getest(x, y, m, beta_ini, lambda_ini)
  beta = res$beta
  lambda = res$lambda
  model_logit = glm(deathrate ~ logCS2, data = Series, family = binomial("logit"), weights = Total)
  chi2statistics = 2 * (loglik(lambda, beta, x, y, m)- loglik(1, model_logit$coefficients, x, y, m))
  return(list(beta = beta, lambda = lambda, loglik = loglik(lambda, beta, x, y, m),
              loglik_logit = loglik(1, model_logit$coefficients, x, y, m), chi2 = chi2statistics,
              iter = res$iteration))
}
loglik = function(lambda, beta, x, y, m){
  eta = x \% *\% beta
  mu = 1 - 1 / (lambda * exp(eta) + 1)^(1 / lambda)
 if(any(mu == 1)) return(Inf)
  else return(sum(m * (y * log(mu / (1 - mu)) + log(1 - mu))))
}
# Prob 1
fit1 = glmfit(Series1, "logit")
#glmfit(Series2, "logit")
fit2 = glmfit(Series2, "cloglog")
qchisq(0.95, 1) # Series1: do not reject HO; Series2: reject HO
# Prob 2
Series_combined = rbind(Series1, Series2)
fit3 = glmfit(Series_combined, "cloglog")
2 * (fit1$loglik + fit2$loglik - fit3$loglik)
qchisq(0.95, 3) # Do not reject HO
# Prob 3
2 * 6 - 2 * (fit1$loglik + fit2$loglik)
2 * 3 - 2 * fit3$loglik
2 * 4 - 2 * (fit1$loglik_logit + fit2$loglik_logit)
2 * 2 - 2 * fit3$loglik_logit
#Series_combined; estimated link
beta = fit3$beta
lambda = fit3$lambda
x_{\text{vec}} = \text{seq}(\text{from} = 3.8, \text{ to} = 4.4, \text{ length} = 100)
eta_vec = beta[1] + x_vec * beta[2]
mu_vec = 1 - 1 / (lambda * exp(eta_vec) + 1)^(1 / lambda)
plot(mu_vec ~ x_vec, type = "1", xlab = "logCS2", ylab = "estimated probability",
     main = "Estimated response curve")
sprintf("eta = %g + %g * x; lambda = %g", beta[1], beta[2], lambda)
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mu_x = -beta[1] / beta[2]
sigma_x = 1 / beta[2]
G_{\text{vec}} = 1 - 1 / (lambda * exp((x_{\text{vec}} - mu_x) / sigma_x) + 1)^(1 / lambda)
plot(G_vec ~ x_vec, type = "1", xlab = "logCS2", ylab = "estimated probability",
     main = "Estimated tolerance distribution")
# Prob 4
x = Series_combined$logCS2
y = Series_combined$deathrate
m = Series_combined$Total
eta = beta[1] + x * beta[2]
mu = 1 - 1 / (lambda * exp(eta) + 1)^(1 / lambda)
sum(2 * m * (y * log(y / mu) + (1 - y) * log((1 - y) / (1 - mu))), na.rm = TRUE)
qchisq(0.95, 13 - 3) # Do not reject HO
di = 2 * m * (y * log(y / mu) + (1 - y) * log((1 - y) / (1 - mu)))
plot(y = sqrt(di[!is.na(di)]) * sign(y[!is.na(di)] - mu[!is.na(di)]), x = y[!is.na(di)],
    ylab = "Deviance Residual", xlab = "Fitted Value", main = "residual plot")
```