## Assignment 5

Shinjon Ghosh

2025-02-14

```
# 1. Simulate Predictor Variables
set.seed(123)
n <- 100
X <- as.data.frame(matrix(rnorm(n * 10), nrow = n, ncol = 10))</pre>
colnames(X) <- paste0("X", 1:10)</pre>
# 2. Calculate the Mean Parameter (mu_i)
mu \leftarrow 1 + 2 * X$X1 + X$X2 + 0.5 * X$X5 + 1.5 * X$X10
# 3. Generate the Count Response (Y i)
Y <- rpois(n, lambda = exp(mu))
# 4. Split Data into Training and Testing Sets
set.seed(456)
train_index <- sample(1:n, size = 80, replace = FALSE)</pre>
test_index <- setdiff(1:n, train_index)</pre>
X train <- X[train index, ]</pre>
X test <- X[test index, ]</pre>
Y_train <- Y[train_index]</pre>
Y_test <- Y[test_index]</pre>
# 5. Fit a Poisson Regression Model
m1 <- glm(Y_train ~ X1 + X2 + X5 + X10, family = poisson(), data = X_train)
summary(m1)
##
## Call:
## glm(formula = Y_train \sim X1 + X2 + X5 + X10, family = poisson(),
##
       data = X_train)
##
## Coefficients:
               Estimate Std. Error z value Pr(>|z|)
                                      20.79 <2e-16 ***
## (Intercept) 1.02650
                           0.04937
                                      84.91
                                              <2e-16 ***
## X1
                1.98425
                           0.02337
## X2
                1.08449 0.03863
                                      28.07 <2e-16 ***
                           0.02115 21.61 <2e-16 ***
## X5
                0.45694
                         0.03034 48.58 <2e-16 ***
## X10
                1.47379
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## (Dispersion parameter for poisson family taken to be 1)
##
       Null deviance: 17728.02 on 79 degrees of freedom
## Residual deviance: 91.47 on 75 degrees of freedom
```

```
## AIC: 340.29
##
## Number of Fisher Scoring iterations: 4
Y pred <- predict(m1, newdata = X test, type = "response")</pre>
print(Y_pred)
              7
##
                            9
                                        11
                                                                    21
                                                      18
26
## 8.561482e+00 5.764463e-01 9.457957e+02 2.006728e-03 4.668347e-01 2.750095e
-01
##
             29
                           37
                                         39
                                                      42
                                                                    44
49
## 8.484655e+00 8.650274e-01 3.180309e+00 1.006892e+00 1.050172e+01 3.353672e
+01
##
             53
                           55
                                         57
                                                      60
                                                                    61
72
## 5.644910e+00 4.023417e+00 5.469277e-01 1.019496e+01 8.784807e+01 1.200804e
-03
             74
##
                           81
## 1.215115e+00 2.687758e+01
# Compare predicted values with actual test values
comparison <- data.frame(Actual = Y_test, Predicted = Y_pred)</pre>
print(comparison)
##
      Actual
                Predicted
           7 8.561482e+00
## 7
## 9
           1 5.764463e-01
## 11
        1048 9.457957e+02
           0 2.006728e-03
## 18
## 21
           0 4.668347e-01
## 26
           0 2.750095e-01
## 29
           9 8.484655e+00
## 37
           0 8.650274e-01
           1 3.180309e+00
## 39
## 42
           2 1.006892e+00
## 44
           7 1.050172e+01
## 49
          32 3.353672e+01
## 53
           7 5.644910e+00
## 55
           8 4.023417e+00
## 57
           0 5.469277e-01
## 60
           6 1.019496e+01
## 61
          81 8.784807e+01
## 72
           0 1.200804e-03
## 74
           1 1.215115e+00
## 81
          25 2.687758e+01
```