

# APTS Statistical Modelling: Practical 1

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Suppose

$$y_{im} \sim \text{Poisson}(\mu(x_{im})),$$

independently, for  $i = 1, \dots, n$  and  $m = 1, \dots, M$ , where

$$\mu(x_{im}) = 8 \exp(q(x_{im})),$$

for some function  $q$ .

Suppose  $x_{im} = x_i = -10 + 20(i-1)/(n-1)$ ,  $M = 3$  and

$$q(x) = 0.001(100 + x + x^2 + x^3).$$

The code below performs the following simulation study. For  $b = 1, \dots, B$ :

- For  $i = 1, \dots, n$  and  $m = 1, \dots, M$ , generate

$$y_{im} \sim \text{Poisson}(\mu(x_{im})).$$

- Record the AIC for models

$$\begin{aligned} y_{im} &\sim \text{Poisson}(\mu(x_{im})) \\ \mu(x_{im}) &= \exp\left(\beta_0 + \sum_{j=1}^p \beta_j x_{im}^j\right), \end{aligned}$$

for  $p = 0, \dots, p_{\max} = 20$ .

```
> B <- 1000
> n <- 25
> M <- 3
> pmax <- 20
>
> x <- rep(seq(from = -10, to = 10, length = n), each = M)
>
> mu <- function(x) {
+   8 * exp(q(x))
+ }
>
> q <- function(x) {
+   0.001 * (100 + x + x^2 + x^3)
+ }
>
> aics <- matrix(0, nrow = B, ncol = pmax)
```

```

> for (b in 1:B) {
+
+   y <- rpois(n = M * n, lambda = mu(x))
+
+   mod <- glm(y ~ 1, family = poisson)
+   aics[b, 1] <- AIC(mod)
+
+   formula <- "y~x"
+   mod <- glm(formula, family = poisson)
+   aics[b, 2] <- AIC(mod)
+
+   for (j in 3:pmax) {
+     formula <- paste(formula, " + I(x^", j - 1, ")", sep = "")
+     mod <- glm(formula, family = poisson)
+     aics[b, j] <- AIC(mod)
+   }
+
+ }
>
> AIOrder <- apply(aics, 1, which.min) - 1
> tAIC <- table(AIOrder)
> tAIC

```

1. Investigate the performance of AIC as a model selection tool for  $n = 25, 50, 100, 1000$ .
2. Vary the simulation model, using

$$q(x) = \frac{1.2}{1+\exp(-x)},$$

to see how AIC performs when the fitted models do not include the simulation model.

3. Modify the code above to compute the values of BIC and  $AIC_c$ , where

$$AIC_c = AIC + \frac{2p^2+2p}{n-p-1}.$$