

STAT 600 - HW 3

Kevin Korsurat

All Rcpp/RcppArmadillo can be found in my [GitHub](#).

Derivation

First, I will derive the log-likelihood function given the completed data, $l(p, \lambda, \mu | \mathbf{Y}, \boldsymbol{\delta})$.

$$\begin{aligned} l(p, \lambda, \mu | \mathbf{Y}, \boldsymbol{\delta}) &= \log(L(p, \lambda, \mu | \mathbf{Y}, \boldsymbol{\delta})) \\ &= \log\left(\prod_{i=1}^n p(Y_i, \delta_i | p, \lambda, \mu)\right) \\ &= \log\left(\prod_{i=1}^n (p\lambda \exp(-\lambda y_i))^{\delta_i} ((1-p)\mu \exp(-\mu y_i))^{1-\delta_i}\right) \\ &= \sum_{i=1}^n \log\left\{(p\lambda \exp(-\lambda y_i))^{\delta_i} ((1-p)\mu \exp(-\mu y_i))^{1-\delta_i}\right\} \\ &= \sum_{i=1}^n [\delta_i \{-\lambda y_i + \log(p\lambda)\} + (1-\delta_i) \{-\mu y_i + \log((1-p)\mu)\}] \end{aligned}$$

Then, the E-step in the EM algorithm is shown below.

$$\begin{aligned} Q(p, \lambda, \mu | p^{(t)}, \lambda^{(t)}, \mu^{(t)}) &= E[l(p, \lambda, \mu | \mathbf{Y}, \boldsymbol{\delta}) | \mathbf{Y}, p^{(t)}, \lambda^{(t)}, \mu^{(t)}] \\ &= E\left[\sum_{i=1}^n [\delta_i \{-\lambda y_i + \log(p\lambda)\} + (1-\delta_i) \{-\mu y_i + \log((1-p)\mu)\}] \middle| \mathbf{Y}, p^{(t)}, \lambda^{(t)}, \mu^{(t)}\right] \\ &= \sum_{i=1}^n E\left[\delta_i \{-\lambda y_i + \log(p\lambda)\} + (1-\delta_i) \{-\mu y_i + \log((1-p)\mu)\} \middle| y_i, p^{(t)}, \lambda^{(t)}, \mu^{(t)}\right] \\ &= \sum_{i=1}^n \left\{ \hat{\delta}_i^{(t)} [-\lambda y_i + \log(p\lambda)] + (1 - \hat{\delta}_i^{(t)}) [-\mu y_i + \log((1-p)\mu)] \right\} \end{aligned}$$

$$\text{where } \hat{\delta}_i^{(t)} = E[\delta_i | y_i, p^{(t)}, \lambda^{(t)}, \mu^{(t)}] = \frac{p^{(t)} \lambda^{(t)} \exp(-\lambda^{(t)} y_i)}{p^{(t)} \lambda^{(t)} \exp(-\lambda^{(t)} y_i) + (1-p^{(t)}) \mu^{(t)} \exp(-\mu^{(t)} y_i)}$$

Then, I will derive the M-step by starting the finding the $p^{(t+1)}$.

$$\begin{aligned}
\frac{d}{dp}Q\left(p, \lambda, \mu|p^{(t)}, \lambda^{(t)}, \mu^{(t)}\right) &\stackrel{\text{set}}{=} 0 \\
\sum_{i=1}^n \left[\frac{\hat{\delta}_i^{(t)}}{p} - \frac{1 - \hat{\delta}_i^{(t)}}{1 - p} \right] &= 0 \\
\sum_{i=1}^n \left[(1 - p)\hat{\delta}_i^{(t)} - (1 - \hat{\delta}_i^{(t)})p \right] &= 0 \\
\sum_{i=1}^n \hat{\delta}_i^{(t)} - np &= 0 \\
p^{(t+1)} &= \frac{1}{n} \sum_{i=1}^n \hat{\delta}_i^{(t)}
\end{aligned}$$

Below is the derivation for $\lambda^{(t+1)}$.

$$\begin{aligned}
\frac{d}{d\lambda}Q\left(p, \lambda, \mu|p^{(t)}, \lambda^{(t)}, \mu^{(t)}\right) &\stackrel{\text{set}}{=} 0 \\
\sum_{i=1}^n \left[-\hat{\delta}_i^{(t)}y_i + \frac{\hat{\delta}_i^{(t)}}{\lambda} \right] &= 0 \\
-\lambda \sum_{i=1}^n \hat{\delta}_i^{(t)}y_i + \sum_{i=1}^n \hat{\delta}_i^{(t)} &= 0 \\
\lambda^{(t+1)} &= \frac{\sum_{i=1}^n \hat{\delta}_i^{(t)}}{\sum_{i=1}^n \hat{\delta}_i^{(t)}y_i}
\end{aligned}$$

Lastly, this is the derivation for $\mu^{(t+1)}$

$$\begin{aligned}
\frac{d}{d\mu}Q\left(p, \lambda, \mu|p^{(t)}, \lambda^{(t)}, \mu^{(t)}\right) &\stackrel{\text{set}}{=} 0 \\
\sum_{i=1}^n \left[(1 - \hat{\delta}_i^{(t)})y_i + \frac{1 - \hat{\delta}_i^{(t)}}{\mu} \right] &= 0 \\
-\mu \sum_{i=1}^n (1 - \hat{\delta}_i^{(t)})y_i + \sum_{i=1}^n (1 - \hat{\delta}_i^{(t)}) &= 0 \\
\mu^{(t+1)} &= \frac{\sum_{i=1}^n (1 - \hat{\delta}_i^{(t)})}{\sum_{i=1}^n (1 - \hat{\delta}_i^{(t)})y_i}
\end{aligned}$$

Appendix

```
knitr::opts_chunk$set(echo = FALSE)

library(tidyverse)
library(knitr)
library(Rcpp)
library(RcppArmadillo)
library(foreach)
library(doParallel)
library(ggplot2)
library(latex2exp)

path <- "/Users/kevin-imac/Desktop/Github - Repo/"
if(! file.exists(path)){
  path <- "/Users/kevinkvp/Desktop/Github Repo/"
}

# sourceCpp(paste0(path, "HW3EM/src/main.cpp"))

### User-defined functions -----
meanSD <- function(x, dplace = 5){
  mm <- round(mean(x), digits = dplace)
  ss <- round(sd(x), digits = dplace)
  paste0(mm, " (SD = ", ss, ")")
}
```