Pratical work: EM algorithm

Thomas Roiseux

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Simulation

We first generate:

- 1. A sample of n = 100 observations with a Poisson law using $\lambda = 3$.
- 2. A sample of n=200 observations with a Poisson law using $\lambda=15$.
- 3. A vector of 300 coordinates, which the 100 first are 1 and the others are 2.

```
sample3 <- rpois(100, 3)</pre>
print(sample3)
      [38] \ 1 \ 3 \ 5 \ 1 \ 3 \ 3 \ 4 \ 0 \ 0 \ 2 \ 5 \ 3 \ 3 \ 2 \ 2 \ 3 \ 2 \ 3 \ 1 \ 2 \ 4 \ 5 \ 1 \ 0 \ 1 \ 1 \ 2 \ 5 \ 4 \ 3 \ 2 \ 0 \ 5 \ 2 \ 4 \ 3 \ 2 
   [75] 0 4 3 3 1 3 4 3 2 7 1 2 4 1 6 4 4 7 4 4 2 3 1 7 5 3
sample15 <- rpois(200, 15)</pre>
print(sample15)
     [1] 16 15 17 17 16 24 17 19 12 13 13 13 13 16 14 18 13 15 14 10 16 12 17 12 17
    [26] 18 12 9 19 15 16 12 21 19 21 19 16 15 14 17 14 17
                                                             8 17 14 14 13 12 15 11
   [51] 9 12 14 14 15 15 13 13 15 19 23 15 16 18 21 10 12 14 18 15 11 20 17 20 14
  [76] 12 16 15 14 16 20 12 23 15 12 11 19 9 23 19 15 16 17 12 7 17 16 13 11 15
## [101] 15 20 16 18 22 22 21 21 9 20 15 17 22 11 17 13 14 13 18 18 9 15 14 22 18
## [126] 16 10 11 23 18 20 17 20 12 17 14 19
                                              9 14 18 11 15 17 16 15
                                                                       9 13 13 20
## [151] 15 19 17 12 13 18 18 9 8 17 20 15 9 17 20 21 18 18 11 18 13 18 15 18 11
## [176] 14 10 17 27 12 9 13 15 17 10 17 21 19 16 13 15 17 15 16 17 10 17 18 12
v <- c()
for (i in 1:100)
{
  v <- c(v, 1, recursive = TRUE)</pre>
}
for (i in 1:200)
  v <- c(v, 2, recursive = TRUE)</pre>
}
```

Now, we are going to generate a Poisson law using two components:

```
#Settings constants
pi1 <- 0.4
pi2 <- 0.6
lambda1 <- 3
```

```
lambda2 <- 15
sample1 <- rpois(300, lambda1)</pre>
sample2 <- rpois(300,lambda2)</pre>
mixed_sample = pi1 * sample1 + pi2 * sample2
print(mixed_sample)
    [1] 12.8 8.8 12.8 7.2 16.6 10.6 7.8 10.4 11.2 6.8 10.0 12.0 7.8 9.6 10.2
   [16] 8.8 13.8 9.2 10.0 13.4 11.2 9.4 10.4 12.2 12.6 13.0 13.0 11.8
   [31] 12.2 12.4 11.8 10.2 9.4 10.0 9.0 10.4 8.4 12.2
                                                        9.8 9.8 13.6 11.0 7.6
   [46] 10.4
              6.4 10.8 10.0 8.8 11.2 9.4 13.0 7.0 7.2
                                                         9.0
                                                             8.8
                                                                  8.8 5.6
   [61] 9.2
             9.2 9.6 11.6 12.4 13.6 11.0 11.8 8.2 11.6
                                                         9.4 14.2
                                                                  8.2 10.0 5.4
              7.2 8.8 9.8 9.2 14.4 11.2 13.6 12.0 9.6
                                                         6.4
                      8.2 5.8 11.4 10.0 9.0 12.8 14.6 13.8
   [91] 9.2 9.8 14.0
                                                             7.0
                                                                  8.4 10.0 15.0
## [106] 9.0 10.8 13.0 7.0 11.4 12.8 5.8 11.6 8.4 10.4 12.0 11.4
                                                                  8.0 17.4 7.8
## [121] 17.2 12.8 8.4 8.0 13.0 9.4 14.8 6.8 8.2 9.8 8.2 8.0
                                                                 8.6 4.8 10.2
## [136] 15.0 11.2 10.2 10.6 11.4 6.4 12.4 11.0 12.8 9.0
                                                         8.4 11.6 14.4 11.8 9.6
## [151] 11.6 8.8 11.0 10.6 11.2 17.6 10.0 14.2 12.0
                                                   6.6
                                                         8.8 12.0
                                                                  7.8 12.0 12.0
## [166] 15.2 9.6 12.8 10.0 9.6 10.6 10.0 12.0 9.2 8.8
                                                         8.0 11.6
                                                                 8.2 15.0 14.4
## [181] 11.4 12.2 12.0 12.8 11.6 10.0 8.8 10.6 13.8 9.4 12.2
                                                             7.4 8.0 10.4 14.0
## [196] 7.2 13.6 11.0 7.8 7.4 7.8 4.4 8.4 7.4 12.4 13.8
                                                             8.2 12.2 16.2 9.0
## [211] 11.8 8.0 10.4 7.2 10.8 13.6
                                     9.2 8.8 12.0 8.2 11.0
                                                             6.0
                                                                  7.6
## [226] 8.6 6.2 10.0 8.4 10.0 9.0 9.8 10.2 12.6 7.2 10.0 12.4
                                                                  8.0 11.4 9.4
## [241] 13.6 13.8 9.6 17.0 12.0 10.2 9.8 8.2
                                               6.4 13.2 13.6 11.0
## [256] 12.8 10.8 14.0 11.4 13.6 9.2 7.0 8.0
                                               7.8
                                                   8.6
                                                         9.4
                                                             6.4
                                                                  9.6
## [271] 15.2 8.6 10.6 12.6 9.8 16.6 10.0 10.4 8.6
                                                    7.4
                                                         9.4
                                                             6.2 13.0 9.4 8.4
## [286] 9.4 13.0 10.6 8.0 10.4 14.2 11.2 6.0 6.6 9.6 7.2 11.6 11.8 10.2 15.0
```

EM-algorithm

For a K-components Poisson law, the density function is:

$$P(x) = \sum_{i=1}^{k} \pi_i \frac{e^{-\lambda_i}}{x!} \lambda_i^x$$

We first want the initialization of the EM-algorithm:

```
em_init = function(K){
}
```