f. Below is the pseudocode for the EM-algorithm. It was asked in an email thread with the professor if such a format is acceptable to represent pseudocode and it was said that this is okay.

```
Let \vec{x} = (x_1, x_2, \dots, x_n) be an observed sample.
Initialize k = 0, \vec{\theta}^{(0)} = (\pi^{(0)}, \mu_1^{(0)}, \mu_2^{(0)}, \sigma_1^{2^{(0)}}, \sigma_2^{2^{(0)}})
Epsilon = 1e^{-10}
Convergence_flag = TRUE
While (Convergence_flag) {
                                       # E-step Compute Q(\vec{\theta}|\vec{\theta}^{(k)})
                                               Q(\vec{\theta}|\vec{\theta}^{(k)}) = \log \pi^{(k)} \sum_{i=1}^{n} E[Z_i|x_i, \theta^{(k)}] + \log \phi(x_i; \mu_1^{(k)}, (\sigma_1^2)^{(k)}) + \log \phi(x_i; \mu_1^{(k)
                                       \log(1-\pi^{(k)})(n-\sum_{i=1}^{n}E[Z_{i}|x_{i},\theta^{(k)}]) + \log\phi(x_{i};\mu_{2}^{(k)},(\sigma_{2}^{2})^{(k)})(n-\sum_{i=1}^{n}E[Z_{i}|x_{i},\theta^{(k)}])
                                       # M-step Maximize Q(\vec{\theta}|\vec{\theta}^{(k)}) w.r.t. \vec{\theta}, set \vec{\theta}^{(k+1)} equal to maximizer of Q
                                       \pi^{(k+1)} = \frac{\eta^{(k)}}{n} = \frac{1}{n} \sum_{i=1}^{n} E[Z_i | x_i, \theta^{(k)}]
                                      \mu_1^{(k+1)} = \frac{1}{n(k)} \sum_{i=1}^n \eta_i^{(k)} x_i
                                       (\sigma_1^2)^{(k+1)} = \frac{1}{n^{(k)}} \sum_{i=1}^n \eta_i^{(k)} \left( x_i - \mu_1^{(k+1)} \right)^2
                                       \mu_2^{(k+1)} = \frac{1}{n-n^{(k)}} \sum_{i=1}^{n} \left(1 - \eta_i^{(k)}\right) x_i
                                        (\sigma_2^2)^{(k+1)} = \frac{1}{n-n(k)} \sum_{i=1}^n \left(1 - \eta_i^{(k)}\right) \left(x_i - \mu_2^{(k+1)}\right)^2
                                        # Check for convergence
                                       If \frac{\left|\left|\vec{\theta}^{(k)} - \vec{\theta}^{(k-1)}\right|\right|}{\left|\left|\vec{\theta}^{(k-1)}\right|\right|} < \text{Epsilon } \{
                                                                                Convergence_flag = FALSE
                                         }
 }
```

Return $\vec{\theta}^{(k+1)}$

Code Appendix (NOTE: Towards the end in the steepest ascent algorithm there are two lines commented to allow for switching of updating methods. They alternate between utilizing the identity matrix and the Fisher Information matrix.)

```
library(plot3D); library(knitr) # Libraries
### Problem 2.5
oilspills <- read.table(file = "oilspills.dat", header = TRUE)
### part a
# log-likelihood of Poisson
g0 <- function(x = oilspills, alpha1, alpha2) {</pre>
  n \leftarrow x[,2]; b1 \leftarrow x[,3]; b2 \leftarrow x[,4]
  sum(n * log(alpha1 * b1 + alpha2 * b2)) +
    sum(-alpha1 * b1 - alpha2 * b2) -
    sum(log(factorial(n)))
}
# dg/dalpha1
g1 <- function(x = oilspills, alpha1, alpha2) {</pre>
  n \leftarrow x[,2]; b1 \leftarrow x[,3]; b2 \leftarrow x[,4]
  sum((n * b1) / (alpha1 * b1 + alpha2 * b2)) - sum(b1)
# dq/dalpha2
g2 <- function(x = oilspills, alpha1, alpha2) {</pre>
 n \leftarrow x[,2]; b1 \leftarrow x[,3]; b2 \leftarrow x[,4]
  sum((n * b2) / (alpha1 * b1 + alpha2 * b2)) - sum(b2)
# gradient(alpha1, alpha2)
g3 <- function(x = oilspills, alpha1, alpha2) {
 return(c(g1(x, alpha1, alpha2), g2(x, alpha1, alpha2)))
}
# d^2g/dalpha1^2
g4 <- function(x = oilspills, alpha1, alpha2) {
 n \leftarrow x[,2]; b1 \leftarrow x[,3]; b2 \leftarrow x[,4]
  -sum((n * b1^2) / (alpha1 * b1 + alpha2 * b2)^2)
# d^2g/dalpha2^2
g5 <- function(x = oilspills, alpha1, alpha2) {
 n \leftarrow x[,2]; b1 \leftarrow x[,3]; b2 \leftarrow x[,4]
  -sum((n * b2^2) / (alpha1 * b1 + alpha2 * b2)^2)
# d^2g/(dalpha1 * dalpha2)
g6 <- function(x = oilspills, alpha1, alpha2) {</pre>
 n \leftarrow x[,2]; b1 \leftarrow x[,3]; b2 \leftarrow x[,4]
  -sum((n * b1 * b2) / (alpha1 * b1 + alpha2 * b2)^2)
# 2x2 Hessian
g7 <- function(x = oilspills, alpha1, alpha2) {</pre>
  return(matrix(c(
    g4(x, alpha1, alpha2), g6(x, alpha1, alpha2),
    g6(x, alpha1, alpha2), g5(x, alpha1, alpha2)), ncol = 2))
h <- function(x = oilspills, alpha1, alpha2) { # h function
  return(-solve(g7(x, alpha1, alpha2)) %*% g3(x, alpha1, alpha2))
relative_convergence <- function(alpha1_old, alpha1_new, # Test for convergence
                                    alpha2_old, alpha2_new, epsilon) {
  old_vec <- c(alpha1_old, alpha2_old)</pre>
```

```
new vec <- c(alpha1 new, alpha2 new)
  relative_convergence_criterion <- sqrt(sum((new_vec - old_vec)^2)) /
    (sqrt(sum(old_vec^2)) + epsilon)
  bool_flag <- ifelse(relative_convergence_criterion < epsilon, FALSE, TRUE)</pre>
  return(c(relative_convergence_criterion, bool_flag))
# Plot 3D graph of the log-likelihood function
# Reference: https://www.datamentor.io/r-programming/3d-plot/
xs <- seq(-5, 10, length.out = 100)
ys <- seq(-5, 10, length.out = 100)
# Reference: https://stackoverflow.com/questions/52317124/dims-product-xx-do-not-match-the-length-of-objec
t-xx-error-in-using-r-func
g0_vectorize <- Vectorize(g0, vectorize.args = c("alpha1", "alpha2"))</pre>
z <- outer(xs, ys, function(xs, ys) g0_vectorize(alpha1 = xs, alpha2 = ys))</pre>
# Reference: https://stackoverflow.com/questions/17606906/find-row-and-column-index-of-maximum-value-in-a-
\max_{x_y} < - \text{ which}(z == \max(z, na.rm = TRUE), arr.ind = TRUE) # 41,40
# xs[max_x_y[1]]; ys[max_x_y[2]] # 1.060606, 0.9090909
persp3D(xs, ys, z,
      main = "log-likelihood of Poisson distribution",
      zlab = "log-likelihood", xlab = "alpha1", ylab = "alpha2",
theta = 45, phi = 35, ticktype = "detailed")
par(mfrow = c(2,2))
persp(xs, ys, z,
              "log-likelihood of Poisson distribution",
      main =
      zlab = "log-likelihood", xlab = "alpha1", ylab = "alpha2",
      theta = 15, phi = 10, col = "springgreen", shade = 0.3,
      ticktype = "detailed")
zlab = "log-likelihood", xlab = "alpha1", ylab = "alpha2",
      theta = 90, phi = 10, col = "springgreen", shade = 0.3,
      ticktype = "detailed")
persp(xs, ys, z,
      main = "log-likelihood of Poisson distribution",
      zlab = "log-likelihood", xlab = "alpha1", ylab = "alpha2",
      theta = -40, phi = 10, col = "springgreen", shade = 0.3,
      ticktype = "detailed")
persp(xs, ys, z,
              "log-likelihood of Poisson distribution",
      main =
      zlab = "log-likelihood", xlab = "alpha1", ylab = "alpha2",
      theta = -180, phi = 10, col = "springgreen", shade = 0.3,
      ticktype = "detailed")
dev.off()
# Newton-Raphson
nr <- function(x = oilspills, alpha1_init, alpha2_init, epsilon = 1e-5) {</pre>
  alpha1_new <- alpha1_init; alpha2_new <- alpha2_init # Initialize variables</pre>
  convergence_flag <- TRUE; count <- 0</pre>
  while(convergence_flag) { # Loop through algorithm
    alpha1_old <- alpha1_new; alpha2_old <- alpha2_new</pre>
    alpha_vec <- c(alpha1_old, alpha2_old) + h(x, alpha1_old, alpha2_old)</pre>
    alpha1_new <- alpha_vec[1]; alpha2_new <- alpha_vec[2]; count <- count + 1</pre>
    convergence_vec <- relative_convergence(alpha1_old, alpha1_new, alpha2_old,</pre>
                                             alpha2_new, epsilon)
    convergence_value <- convergence_vec[1]; convergence_flag <- convergence_vec[2]</pre>
  cat('Iterations: ', count, '\n')
```

```
return(c(alpha_vec, count))
}
# Test values for Newton-Raphson
test_initializations <- expand.grid(-1:3, -1:3)</pre>
# Reference: https://stackoverflow.com/questions/4227223/convert-a-list-to-a-data-frame
test_initializations <- matrix(unlist(test_initializations),</pre>
                                 ncol = length(test_initializations))
test_results <- mapply(function(a1, a2) {</pre>
    tryCatch(nr(alpha1_init = a1, alpha2_init = a2),
              error = function(e) NA,
              warning = function(w) NA,
              message = function(c) NA
  }, a1 = test_initializations[,1], a2 = test_initializations[,2]
test_results <- do.call(rbind, test_results)</pre>
test_results <- as.data.frame(test_results)</pre>
test_results <- cbind(test_initializations, test_results)
colnames(test_results) <- c("alpha1_init", "alpha2_init", "alpha1", "alpha2", "Iterations")</pre>
kable(test_results, format = "latex", caption = "Results for Newton-Raphson")
nr(alpha1_init = -1, alpha2_init = -1)
nr_fisher(alpha1_init = 1, alpha2_init = -1)
# part b
fisher_information <- function(x = oilspills,
                                 alpha1, alpha2) { # Fisher Information matrix
  n \leftarrow x[,2]; b1 \leftarrow x[,3]; b2 \leftarrow x[,4]
  lambda_i <- alpha1 * b1 + alpha2 * b2</pre>
  top_left <- -sum(b1^2 / lambda_i)</pre>
  bottom_left <- top_right <- -sum(b1 * b2 / lambda_i)</pre>
  bottom_right <- -sum(b2^2 / lambda_i)</pre>
  matrix(c(top_left, bottom_left, top_right, bottom_right), ncol = 2)
h_fisher <- function(x, alpha1, alpha2) { # h function with Fisher scoring
  return(-solve(fisher_information(x, alpha1, alpha2)) %*%
            g3(x, alpha1, alpha2))
nr_fisher <- function(x = oilspills, alpha1_init, alpha2_init, epsilon = 1e-5) {</pre>
  alpha1_new <- alpha1_init; alpha2_new <- alpha2_init # Initialize variables
  convergence_flag <- TRUE; count <- 0</pre>
  while(convergence_flag) { # Loop through algorithm
    alpha1_old <- alpha1_new; alpha2_old <- alpha2_new # Update old variables
    # Update new variables
    alpha_vec <- c(alpha1_old, alpha2_old) + h_fisher(x, alpha1_old, alpha2_old)</pre>
    alpha1_new <- alpha_vec[1]; alpha2_new <- alpha_vec[2]; count <- count + 1</pre>
    # Check for convergence
    convergence_vec <- relative_convergence(alpha1_old, alpha1_new, alpha2_old,</pre>
                                                alpha2_new, epsilon)
    convergence_value <- convergence_vec[1]; convergence_flag <- convergence_vec[2]</pre>
  cat('Iterations: ', count, '\n')
  return(c(alpha_vec, count))
test_results_fisher <- mapply(function(a1, a2) {</pre>
  tryCatch(nr_fisher(alpha1_init = a1, alpha2_init = a2),
            error = function(e) NA,
            warning = function(w) NA,
            message = function(c) NA
  )
```

```
}, a1 = test_initializations[,1], a2 = test_initializations[,2]
test_results_fisher <- do.call(rbind, test_results_fisher)</pre>
test_results_fisher <- as.data.frame(test_results_fisher)</pre>
test_results_fisher <- cbind(test_initializations, test_results_fisher)
colnames(test_results_fisher) <- c("alpha1_init", "alpha2_init", "alpha1", "alpha2", "Iterations")
kable(test_results_fisher, format = "latex", caption = "Results for Newton-Raphson with Fisher Scoring")</pre>
# part e
backtrack_check <- function(x = oilspills,</pre>
                                alpha1 old btc,
                                alpha1_test,
                                alpha2_old_btc,
                                alpha2_test,
                                alpha_btc,
                                backtracks_btc,
                               backtrack_flag_btc) {
  step_old <- g0(x = x), # Calculate steps
                   alpha1 = alpha1_old_btc,
                   alpha2 = alpha2_old_btc)
  step_test <- g0(x = x)
                    alpha1 = alpha1_test,
                    alpha2 = alpha2_test)
  if(step_test < step_old) { # Check if "downhill" or sideways</pre>
    alpha_btc <- alpha_btc / 2
    backtracks_btc <- backtracks_btc + 1</pre>
  } else { # Stop tuning alpha
    backtrack flag btc <- FALSE
  return(c(alpha_btc, backtracks_btc, backtrack_flag_btc))
alpha_tuner <- function(x = oilspills,</pre>
                           alpha1_old_at,
                           alpha2_old_at,
                           alpha_at,
                           backtracks_at) {
  backtrack_flag <- TRUE # (Re-)Initialize blacktracking variables</pre>
  while (backtrack_flag) { # Loop until a suitable alpha is found
    # theta test <- c(alpha1 old at, alpha2 old at) + # trying to find first update
       # alpha_at * h_fisher(x, alpha1_old_at, alpha2_old_at)
    theta_test <- c(alpha1_old_at, alpha2_old_at) + # trying to find first update
       alpha_at * g3(x, alpha1_old_at, alpha2_old_at)
    backtrack_vec <- backtrack_check(x = x,</pre>
                                          alpha1_old_btc = alpha1_old_at,
                                          alpha1_test = theta_test[1],
                                          alpha2 old btc = alpha2 old at,
                                          alpha2_test = theta_test[2],
                                          alpha_btc = alpha_at,
                                          backtracks_btc = backtracks_at,
                                          backtrack_flag_btc = backtrack_flag)
     alpha_at <- backtrack_vec[1]; backtracks_at <- backtrack_vec[2]</pre>
    backtrack_flag <- backtrack_vec[3]</pre>
  return(c(alpha_at, backtracks_at))
steepest_ascent <- function(x = oilspills,</pre>
                                alpha1_init,
                                alpha2_init,
                                epsilon = 1e-10) {
  alpha1_new <- alpha1_old <- alpha1_init # Initialize variables</pre>
  alpha2_new <- alpha2_old <- alpha2_init</pre>
  convergence_flag <- TRUE; count <- 0; backtracks <- 0; alpha <- 1</pre>
  while(convergence_flag) { # Loop through algorithm
```

```
alpha1 old <- alpha1 new; alpha2 old <- alpha2 new; alpha <- 1
    # Tune alpha
    alpha_tune_vec <- alpha_tuner(x = x,</pre>
                                     alpha1_old_at = alpha1_old,
                                     alpha2_old_at = alpha2_old,
                                     alpha_at = alpha,
                                     backtracks_at = backtracks)
    alpha <- alpha_tune_vec[1]; backtracks <- alpha_tune_vec[2]</pre>
    # Update old / new variables
    # theta_vec <- c(alpha1_old, alpha2_old) + alpha * h_fisher(x, alpha1_old, alpha2_old)
    theta_vec <- c(alpha1_old, alpha2_old) + alpha * g3(x, alpha1_old, alpha2_old)
    alpha1_new <- theta_vec[1]; alpha2_new <- theta_vec[2]; count <- count + 1</pre>
    # Check for convergence
    convergence_vec <- relative_convergence(alpha1_old, alpha1_new, alpha2_old,
                                                alpha2_new, epsilon)
    convergence_flag <- convergence_vec[1]; convergence_value <- convergence_vec[2]</pre>
  cat('Iterations: ', count, '\n')
cat('Backtracks: ', backtracks, '\n')
  return(c(theta_vec, count, backtracks))
}
test_results_steepest_ascent_identity <- mapply(function(a1, a2) {</pre>
  tryCatch(steepest_ascent(alpha1_init = a1, alpha2_init = a2),
            error = function(e) NA,
            warning = function(w) NA,
            message = function(c) NA
 }, a1 = test_initializations[,1], a2 = test_initializations[,2]
test_results_steepest_ascent_identity <- do.call(rbind, test_results_steepest_ascent_identity)</pre>
test_results_steepest_ascent_identity <- as.data.frame(test_results_steepest_ascent_identity)</pre>
test_results_steepest_ascent_identity <- cbind(test_initializations, test_results_steepest_ascent_identit
colnames(test_results_steepest_ascent_identity) <- c("alpha1_init", "alpha2_init", "alpha1", "alpha2", "It</pre>
erations", "Backtracks")
kable(test_results_steepest_ascent_identity, format = "latex", caption = "Results for Steepest Ascent with
Identity Matrix")
test_results_steepest_ascent_fisher <- mapply(function(a1, a2) {</pre>
  tryCatch(steepest_ascent(alpha1_init = a1, alpha2_init = a2),
            error = function(e) NA,
            warning = function(w) NA,
            message = function(c) NA
 }, a1 = test_initializations[,1], a2 = test_initializations[,2]
test_results_steepest_ascent_fisher <- do.call(rbind, test_results_steepest_ascent_fisher)
test_results_steepest_ascent_fisher <- as.data.frame(test_results_steepest_ascent_fisher)</pre>
test_results_steepest_ascent_fisher <- cbind(test_initializations, test_results_steepest_ascent_fisher) colnames(test_results_steepest_ascent_fisher) <- c("alpha1_init", "alpha2_init", "alpha1", "alpha2", "Iter
ations". "Backtracks")
kable(test_results_steepest_ascent_fisher, format = "latex", caption = "Results for Steepest Ascent with F
isher Matrix")
```