## Lab 11 Solution

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## Problem 1

Estimate the parameters  $(\mu, \sigma^2)$  of a normal distribution  $\mathcal{N}(\mu, \sigma^2)$  using the **Newton-Raphson method**, and compare with their closed-form MLEs.

## Solution:

```
# Simulate data from N(mu, sigma^2), where mu is 5 and sigma = 2
set.seed(123)
x <- rnorm(100, mean = 5, sd = 2)
# No. of data points
n <- length(x)</pre>
```

```
# Log-likelihood derivatives

loglik_grad <- function(mu, sigma2, x) {
  dmu <- sum(x - mu) / sigma2
  dsigma2 <- -n / (2 * sigma2) + sum((x - mu)^2) / (2 * sigma2^2)
  return(c(dmu, dsigma2))
}</pre>
```

The above function 'loglik\_grad' returns the gradient of the log likelihood function for a particular value of  $\mu$  and  $\sigma^2$ .

```
# Setting up the Hessian matrix
loglik_hessian <- function(mu, sigma2, x) {</pre>
```

```
# Partial double derivative of log likelihood wrt mu
h11 <- -n / sigma2
# Non diagonal elements of the Hessian matrix
h12 <- -2 * sum(x - mu) / sigma2^2
h21 <- h12 # Symmetric
# Partial double derivative of log likelihood wrt sigma^2
h22 <- n / (2 * sigma2^2) - sum((x - mu)^2) / (sigma2^3)

return(matrix(c(h11, h12, h21, h22), nrow = 2, byrow = TRUE))
}</pre>
```

The above function 'loglik\_hessian' returns the Hessian matrix of the log likelihood function for a particular value of  $\mu$  and  $\sigma^2$ .

```
# Newton-Raphson to estimate both mu and sigma^2
newton_raphson_mle <- function(x, mu_init = 0, sigma2_init = 1,</pre>
                                     tol = 1e-8, max_iter = 1000) {
  # Suitable initial choices for mu and sigma^2
  mu <- mu init
  sigma2 <- sigma2_init</pre>
  for (i in 1:max_iter) {
     grad <- loglik_grad(mu, sigma2, x)</pre>
    hess <- loglik_hessian(mu, sigma2, x)</pre>
    # Hessian_inverse %*% Gradient(theta_old)
    step <- solve(hess, grad)</pre>
    # Updating parameters as per
     # ...theta_new = theta_old - Hessian_inverse %*% Gradient(theta_old)
    mu_new <- mu - step[1]</pre>
     sigma2_new <- sigma2 - step[2]</pre>
     # Stopping condition
     if (\operatorname{sqrt}((\operatorname{mu_new} - \operatorname{mu})^2 + (\operatorname{sigma2_new} - \operatorname{sigma2})^2) < \operatorname{tol}) {
     }
    mu <- mu_new
     sigma2 <- sigma2_new
```

```
return(c(mu = mu, sigma2 = sigma2))
}
```

The above function uses the Newton-Raphson Method to find the MLEs of the parameters and returns the estimated parameter values with a tolerance of  $10^{-6}$ .

```
# Closed-form MLEs
mu_mle <- mean(x)
sigma2_mle <- var(x)

# Closed-form Estimates
cat(" mu =", mu_mle)

mu = 5.180812

cat(" sigma^2 =", sigma2_mle)

sigma^2 = 3.332931</pre>
```

The above are the closed form MLEs of the parameters  $\mu$  and  $\sigma^2$ .

```
# Run Newton-Raphson
mu_init <- 5
sigma2_init <- 4
mle_estimates <- newton_raphson_mle(x, mu_init, sigma2_init)

# Newton-Raphson Estimates:
cat(" mu =", mle_estimates["mu"])

mu = 5.180812

cat(" sigma^2 =", mle_estimates["sigma2"])

sigma^2 = 3.299602</pre>
```

The above are the MLEs of the parameters  $\mu$  and  $\sigma^2$ , using the Newton-Raphson method.

```
# Comparing the results obtained
# Absolute Error
abs_error_mu <- abs(mle_estimates["mu"] - mu_mle)</pre>
abs_error_sigma2 <- abs(mle_estimates["sigma2"] - sigma2_mle)</pre>
# Relative Error
rel_error_mu <- abs_error_mu / abs(mu_mle)</pre>
rel_error_sigma2 <- abs_error_sigma2 / abs(sigma2_mle)</pre>
# Squared Error
squared_error_mu <- (mle_estimates["mu"] - mu_mle)^2</pre>
squared_error_sigma2 <- (mle_estimates["sigma2"] - sigma2_mle)^2</pre>
# Total report
comparison_error <- data.frame(</pre>
  row.names = c("Closed Form", "Newton-Raphson", "Absolute Error",
                  "Relative Error", "Squared Error"),
  mu = c(mu_mle, mle_estimates["mu"], abs_error_mu,
         rel_error_mu, squared_error_mu),
  sigma2 = c(sigma2_mle, mle_estimates["sigma2"], abs_error_sigma2,
              rel_error_sigma2, squared_error_sigma2)
)
print(comparison_error)
```

```
mu sigma2
Closed Form 5.180812e+00 3.332931321
Newton-Raphson 5.180812e+00 3.299602008
Absolute Error 5.329071e-15 0.033329313
Relative Error 1.028617e-15 0.010000000
Squared Error 2.839899e-29 0.001110843
```

From the above values of errors, we can see that the MLEs obtained by Newton-Raphson are quite close to that of the closed form MLEs of  $\mu$  and  $\sigma^2$ .