

MTH210: Lab 8

Linear Regression, MLE, Ridge, and Newton-Raphson

1. Load the `cars` dataset in R:

```
data(cars)
```

Fit a linear regression model using maximum likelihood with response y being the distance and x being speed. Remember to include an intercept term in X by making the first column as a column of 1s. Do not use inbuilt functions in R to fit the model.

2. Load the `fuel2001` dataset in R:

```
fuel2001 <- read.csv("https://dvats.github.io/assets/fuel2001.csv", row.names = 1)
```

- a. Fit the linear regression model using maximum likelihood with response `FuelC`. Remember to include an intercept in X . What is your final estimate of β and σ^2 ?
 - b. For $\lambda = 1$, what is the ridge regression estimator of β ?
3. **Simulating data in R:** Let $X \in \mathbb{R}^{n \times p}$ be the design matrix, where all entries in its first column equal one (to form an intercept). Let $x_{i,j}$ be the (i,j) th element of X . For the i^{th} case, $x_{i1} = 1$ and x_{i2}, \dots, x_{ip} are the values of the $p - 1$ predictors. Let y_i be the response for the i th case and define $y = (y_1, \dots, y_n)^T$. The model assumes that y is a realization of the random vector:

$$Y \sim N_n(X\beta_*, \sigma_*^2 I_n),$$

where $\beta_* \in \mathbb{R}^p$ are unknown regression coefficients and $\sigma_*^2 > 0$ is the unknown variance. We would like to *generate data* that actually follows the following model. This is useful when building too methods and different estimation techniques for β .

- a. Study the code below and understand how the data is being generated according to the model:

```
set.seed(1)
n <- 50
p <- 5
sigma2.star <- 1/2
beta.star <- rnorm(p)
```

```
X <- cbind(1, matrix(rnorm(n*(p-1)), nrow = n, ncol = (p-1)))
y <- X %*% beta.star + rnorm(n, mean = 0, sd = sqrt(sigma2.star))
```

- b. Having generated the above (y, X) , obtain the MLE of β . Is the MLE roughly close to β^* ? What happens when you increase $n = 500$?
 - c. Repeat the data generation process, but now change $p = 100$ and keep $n = 50$. Can you find the traditional MLE in this case?
 - d. For the above data find the ridge regression estimator of β for $\lambda = 0.01, 0.1, 1, 10$.
4. Write a Newton-Raphson code to find the MLE of α for Gamma $(\alpha, 1)$ distribution. That is, suppose $X_1, X_2, \dots, X_n \stackrel{iid}{\sim} \text{Gamma}(\alpha, 1)$, then write a function to obtain $\hat{\alpha}_{MLE}$.

In order to implement this, you will need data that is from $\text{Gamma}(\alpha, 1)$. You may use the following:

```
set.seed(100)
alpha <- 5 #true value of alpha
n <- 10 # actual data size is small first
dat <- rgamma(n, shape = alpha, rate = 1)
```

The above generates $n = 10$ observations. Use `dat` to obtain the MLE of α . You will need the `pracma` library in R to calculate the derivatives of $\Gamma(\cdot)$ function.

```
library(pracma) #for psi function
?psi
```

5. Using both Newton-Raphson and gradient ascent algorithm, maximize objective function

$$f(x) = \cos(x) \quad x \in [-\pi, 3\pi].$$