

# Accuracy Report

Wonbin Song

```
source('separate_file.R')

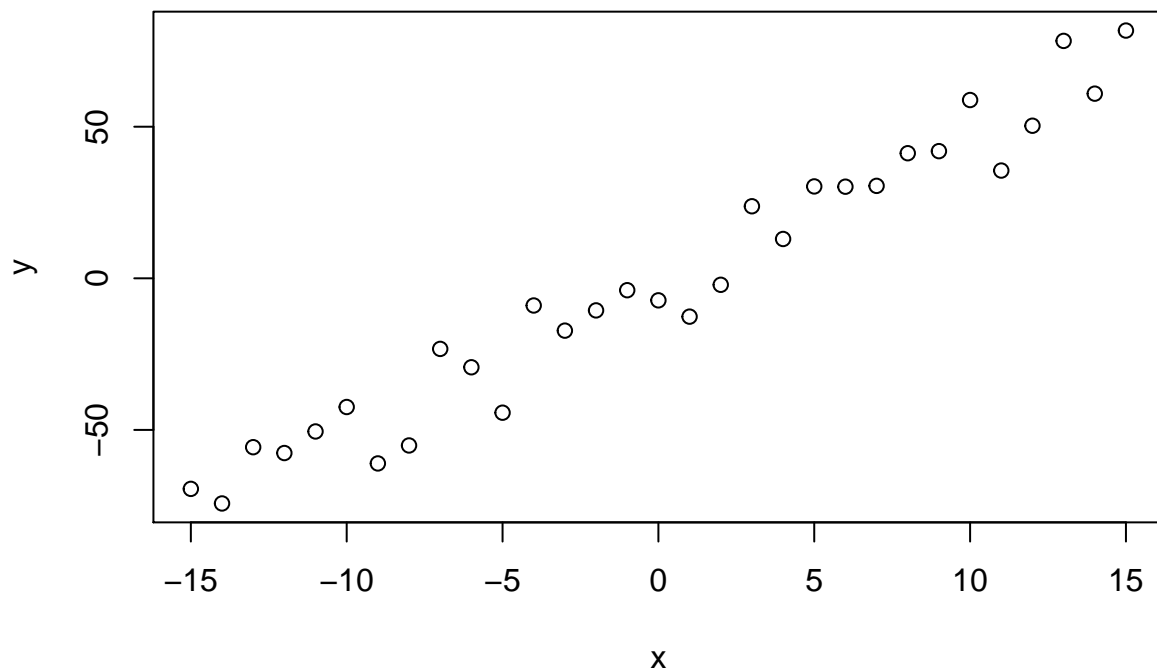
# Creating test data

trueA <- 5 # slope
trueB <- 0 # intercept
trueSd <- 10 # standard deviation of error
sampleSize <- 31 # sample size

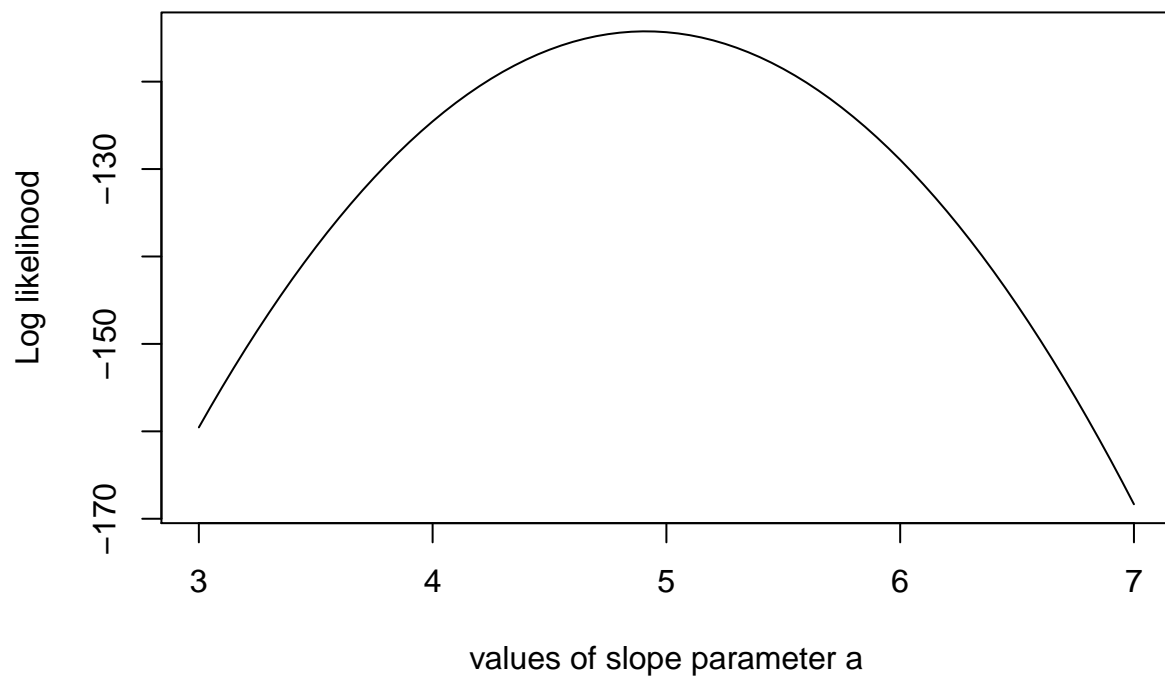
# create independent x-values
x <- (-(sampleSize-1)/2):((sampleSize-1)/2) # create a vector x
# create dependent values according to  $ax + b + N(0, sd)$ 
y <- trueA * x + trueB + rnorm(n=sampleSize, mean=0, sd=trueSd) # create a vector y

plot(x, y, main="Test Data") # plot x, y
```

Test Data



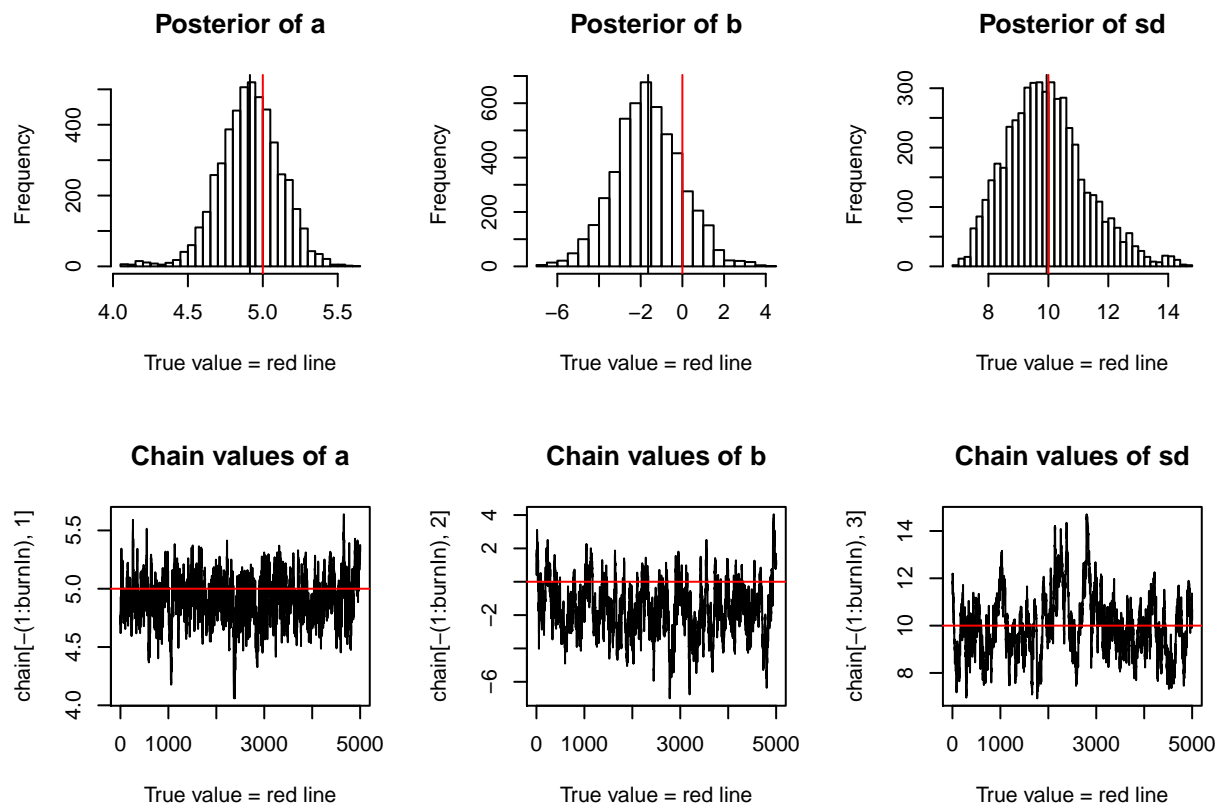
```
slopelikelihoods <- lapply(seq(3, 7, by=.05), slopevalues ) # apply the function slopevalues to the co
plot (seq(3, 7, by=.05), slopelikelihoods , type="l", xlab = "values of slope parameter a", ylab = "Log
```



```
startvalue = c(4,0,10) # startvalue = (4,0,10)
chain = run_metropolis_MCMC(startvalue, 10000) # compute the value "chain" from the function "run_metropolis_MCMC"

burnIn = 5000 # burnIn = 5000
acceptance = 1-mean(duplicated(chain[-(1:burnIn),])) # compute acceptance rates

# plot the results
summary.plot(chain, burnIn,trueA,trueB,trueSd)
```



```
# for comparison:
summary(lm(y~x)) # summaries of the results of model fitting
```

```
##
## Call:
## lm(formula = y ~ x)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -18.173  -5.966   1.716   7.799  16.047
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  -1.5882     1.7641   -0.9    0.375
## x              4.9114     0.1972  24.9   <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 9.822 on 29 degrees of freedom
## Multiple R-squared:  0.9553, Adjusted R-squared:  0.9538
## F-statistic: 620.1 on 1 and 29 DF,  p-value: < 2.2e-16
```

```
# new function
```

```
compare_outcomes <- function(iterations){ # create a function with "iterations" as an input
```

```

a.matrix <- array(dim=c(10,2)) # creat an empty matrix
colnames(a.matrix) <- c('mean', 'std') # column names
for(i in 1:10){ # for loop starting from 1 to 10
  startvalue <- c(runif(1,0,10), rnorm(1,0,5), runif(1,0,30)) # set a starting value
  chain <- run_metropolis_MCMC(startvalue, iterations) # use the function "run_metropolis_MCMC" to
  a.matrix[i,1] <- mean(chain[,1]) # compute the mean of the values in the chain for a
  a.matrix[i,2] <- sd(chain[,1]) # compute the standard deviation of the values in the chain for a
}
return(a.matrix) # return the computed mean and standard deviation
}

# Test the function for 1,000, 10,000, and 100,000 iterations
compare_outcomes(1000)

```

```

##           mean      std
## [1,] 4.789681 0.3461592
## [2,] 4.718021 0.7076853
## [3,] 5.331610 1.0409435
## [4,] 4.747596 0.5963721
## [5,] 5.495010 1.3035896
## [6,] 4.673074 0.8734947
## [7,] 4.837919 0.4190807
## [8,] 4.831589 0.1966257
## [9,] 4.719721 0.4989302
## [10,] 4.396092 1.2077150

```

```
compare_outcomes(10000)
```

```

##           mean      std
## [1,] 4.931565 0.2238603
## [2,] 4.876886 0.3656853
## [3,] 4.908385 0.2416311
## [4,] 4.904625 0.2476450
## [5,] 4.907536 0.2291882
## [6,] 4.907507 0.2297831
## [7,] 4.891732 0.2786114
## [8,] 4.912359 0.2057280
## [9,] 4.947299 0.2964820
## [10,] 4.863630 0.3968799

```

```
compare_outcomes(100000)
```

```

##           mean      std
## [1,] 4.907988 0.2141141
## [2,] 4.910638 0.2131626
## [3,] 4.913832 0.2069193
## [4,] 4.915440 0.2319948
## [5,] 4.910926 0.2097763
## [6,] 4.909756 0.2113765
## [7,] 4.906740 0.2426225
## [8,] 4.918152 0.2090453
## [9,] 4.905929 0.2086539
## [10,] 4.911696 0.2089489

```

These results are the mean and the standard deviation of the values in the chain for  $a$  with 1000, 10000, and 100000 iterations. As the number of iteration increases, the mean of the values in the chain for  $a$  converge to trueA, which is 5, and the standard deviation of the values in the chain for  $a$  become smaller. In other words, the accuracy of this algorithm increases as the number of iteration increases.