# Metropolis-Hastings example

## Florian Hartig

This example has been taken from this blog post.

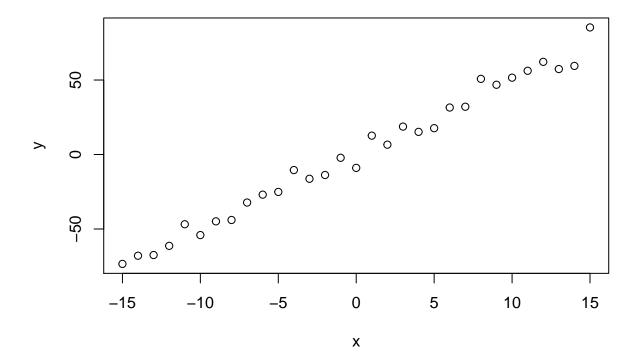
## simulate test data

```
trueA <- 5
trueB <- 0
trueSd <- 5
sampleSize <- 31

set.seed(1500)
# create independent x-values
x <- (-(sampleSize-1)/2):((sampleSize-1)/2)
# create dependent values according to ax + b + N(0,sd)
y <- trueA * x + trueB + rnorm(n=sampleSize,mean=0,sd=trueSd)

plot(x,y, main="Test Data")</pre>
```

### **Test Data**



## Implementing MH algorithm

```
run_metropolis_MCMC <- function(startvalue, iterations){</pre>
    chain = array(dim = c(iterations+1,3))
    chain[1,] = startvalue
   for (i in 1:iterations){
        proposal = proposalfunction(chain[i,])
       probab = exp(posterior(proposal) - posterior(chain[i,]))
        if (runif(1) < probab){</pre>
            chain[i+1,] = proposal
        }else{
            chain[i+1,] = chain[i,]
   }
   return(chain)
# propose new parameter values
proposalfunction <- function(param){</pre>
   return(rnorm(3,mean = param, sd= c(0.1,0.5,0.3)))
}
# evaluate log posterior at given parameter values
posterior <- function(param){</pre>
  return (likelihood(param) + prior(param))
}
# evaluate log prior at given parameter values
prior <- function(param){</pre>
   a = param[1]
   b = param[2]
   sd = param[3]
   aprior = dunif(a, min=0, max=10, log = T)
   bprior = dnorm(b, sd = 5, log = T)
    sdprior = dunif(sd, min=0, max=10, log = T)
   return(aprior+bprior+sdprior)
}
# evaluate log likelihood at given parameter values
likelihood <- function(param){</pre>
   a = param[1]
   b = param[2]
   sd = param[3]
   pred = a*x + b
   singlelikelihoods = dnorm(y, mean = pred, sd = sd, log = T)
   sumll = sum(singlelikelihoods)
   return(sumll)
}
```

### Run MH algorithm

```
set.seed(1)
# initial value
startvalue = c(4,0,4)
# simulate 10000 samples
chain = run_metropolis_MCMC(startvalue, 10000)
# remove the first 5000 as burn-in
burnIn = 5000
# computing average acceptance probability
acceptance = 1-mean(duplicated(chain[-(1:burnIn),]))
acceptance
## [1] 0.6414717
par(mfrow = c(2,2))
hist(chain[-(1:burnIn),1],nclass=30, , main="Posterior of a", xlab="True value = red line")
abline(v = mean(chain[-(1:burnIn),1]), col="green")
abline(v = trueA, col="red" )
hist(chain[-(1:burnIn),2],nclass=30, main="Posterior of b", xlab="True value = red line")
abline(v = mean(chain[-(1:burnIn),2]), col="green")
abline(v = trueB, col="red" )
```

hist(chain[-(1:burnIn),3],nclass=30, main="Posterior of sd", xlab="True value = red line")

#### Posterior of a

abline(v = mean(chain[-(1:burnIn),3]), col="green" )

# -requency 5.0 5.2 5.4

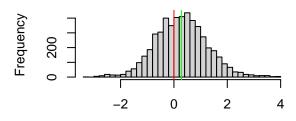
4.8

abline(v = trueSd, col="red" )

4.6

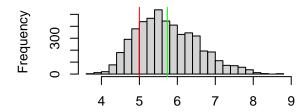
True value = red line

#### Posterior of b



True value = red line

#### Posterior of sd



True value = red line

```
# for comparison:
summary(lm(y~x))
```

```
##
## Call:
## lm(formula = y \sim x)
##
## Residuals:
       Min 1Q Median 3Q
                                       Max
## -10.3580 -3.8445 0.8254 2.4071 10.7585
##
## Coefficients:
             Estimate Std. Error t value Pr(>|t|)
##
## (Intercept) 0.2710
                      0.9989 0.271
                                         0.788
## x
               4.9678
                         0.1117 44.482 <2e-16 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 5.562 on 29 degrees of freedom
## Multiple R-squared: 0.9856, Adjusted R-squared: 0.9851
## F-statistic: 1979 on 1 and 29 DF, p-value: < 2.2e-16
```