

Likelihood and maximum likelihood estimation

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Contents

1	The probability density/mass function	1
2	The likelihood function for a single observation	2
3	The likelihood function for n observations	3
4	Example	5

1 The probability density/mass function

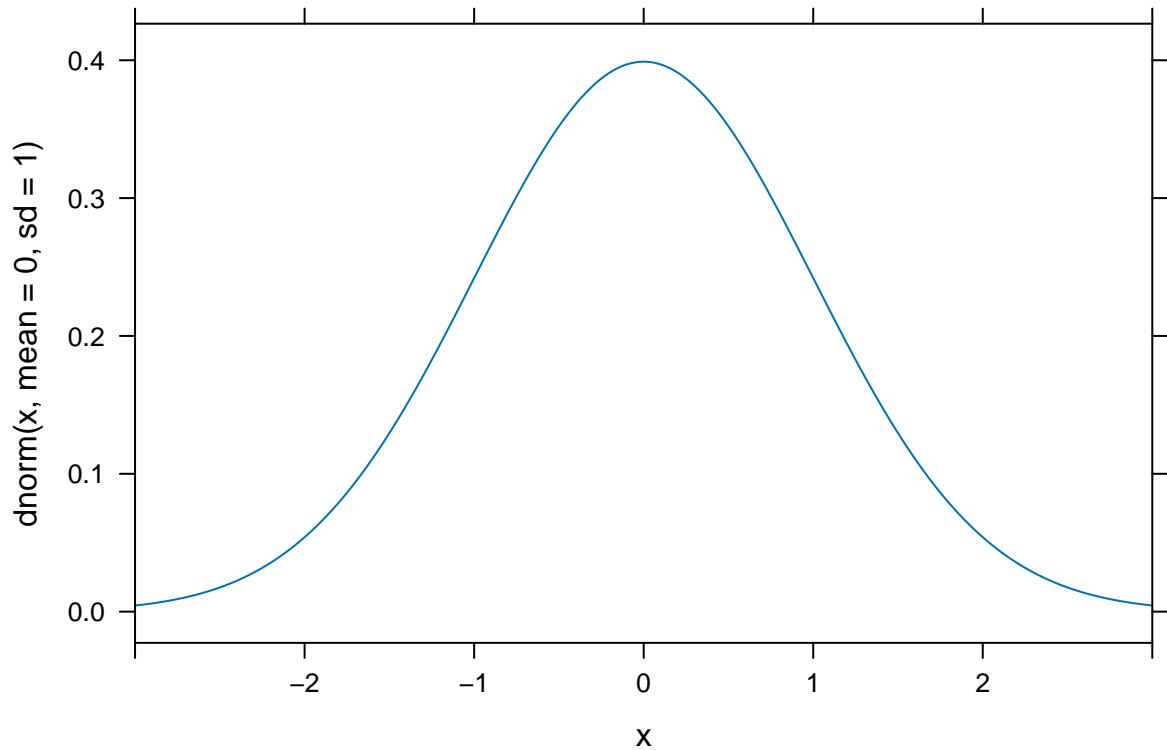
$$X \sim N(\mu, \sigma)$$

Assumes μ and σ are some fixed values:

$$f(x) = f(X = x; \mu, \sigma) = \frac{1}{\sigma\sqrt{2\pi}} \exp\left[-\frac{1}{2} \left(\frac{x-\mu}{\sigma}\right)^2\right]$$

E.g. for $\mu = 0$ and $\sigma = 1$:

```
plotFun(dnorm(x, mean = 0, sd = 1) ~ x, xlim = c(-3, 3))
```



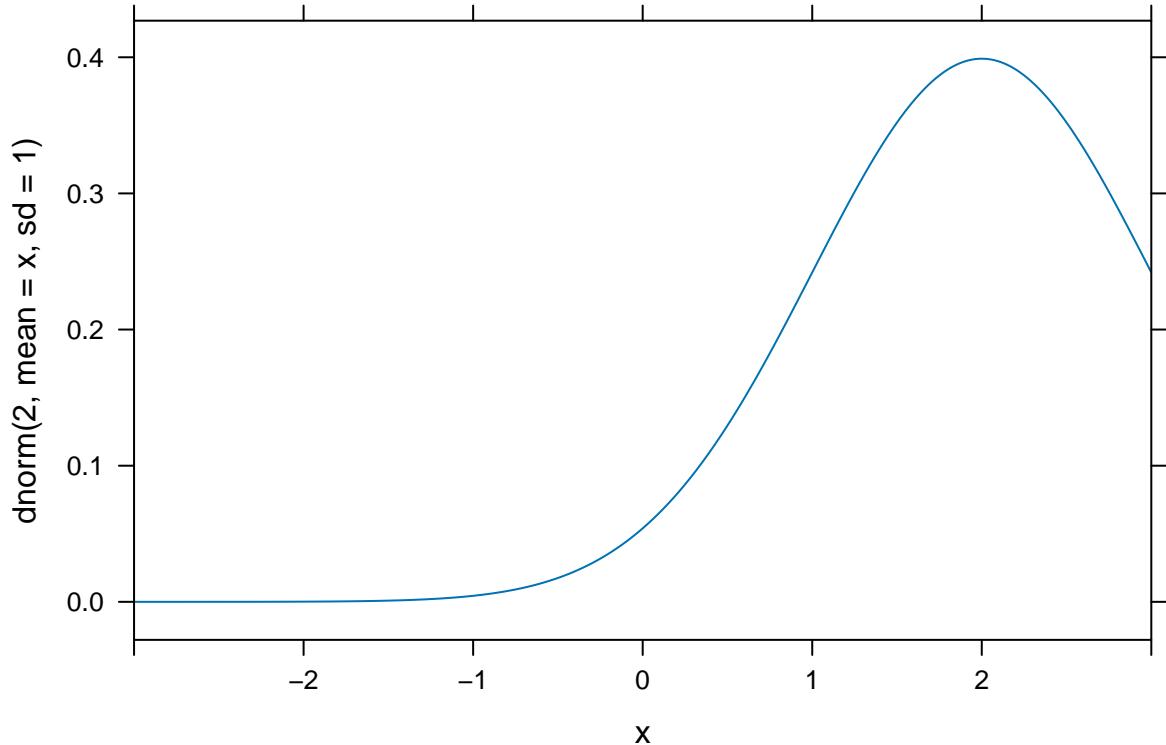
2 The likelihood function for a single observation

What if instead the data is fixed and we let μ and σ vary?

$$L(\mu, \sigma) = L(\mu, \sigma; X = x) = f(X = x; \mu, \sigma)$$

E.g. assume that $X = 2$ was observed:

```
plotFun(dnorm(2, mean = x, sd = 1) ~ x, xlim = c(-3, 3))
```



The value of μ that gives the highest probability (density) of observing $X = 2$ is $\mu = 2$.

3 The likelihood function for n observations

Assume independence for n observations, then

$$L(\mu, \sigma) = L(\mu, \sigma; X_1, X_2, \dots, X_n) = \prod_{i=1}^n L(\mu, \sigma; X_i) = \prod_{i=1}^n L(\mu, \sigma; X_i). \quad (1)$$

- Maximum likelihood estimation (MLE):
 - Find the values of μ and σ that gives the largest value of L for fixed X 's.
 - Would you rather differentiate products or sums?

Do you know a function that turns products into sums?

$$\log(a \cdot b) = \log(a) + \log(b)$$

Assume independence for n observations, then

$$L(\mu, \sigma) = L(\mu, \sigma; X_1, X_2, \dots, X_n) = \prod_{i=1}^n L(\mu, \sigma; X_i) = \prod_{i=1}^n L(\mu, \sigma; X_i) \quad (2)$$

$$l(\mu, \sigma) = \log L(\mu, \sigma) = \log L(\mu, \sigma; X_1, X_2, \dots, X_n) \quad (3)$$

$$= \log \left(\prod_{i=1}^n L(\mu, \sigma; X_i) \right) \quad (4)$$

$$= \sum_{i=1}^n \log L(\mu, \sigma; X_i) \quad (5)$$

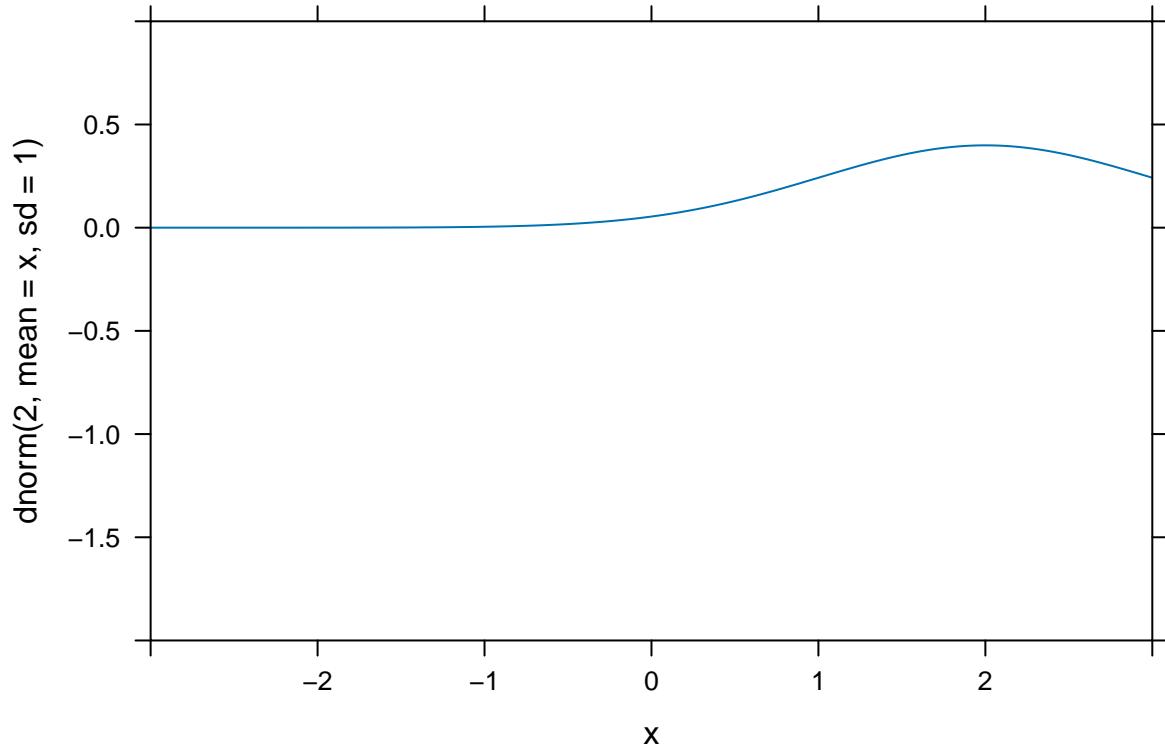
$$= \sum_{i=1}^n l(\mu, \sigma; X_i). \quad (6)$$

Note that \log is a monotonic (increasing) function, so maximum of $l = \log L$ is the same as that of L .

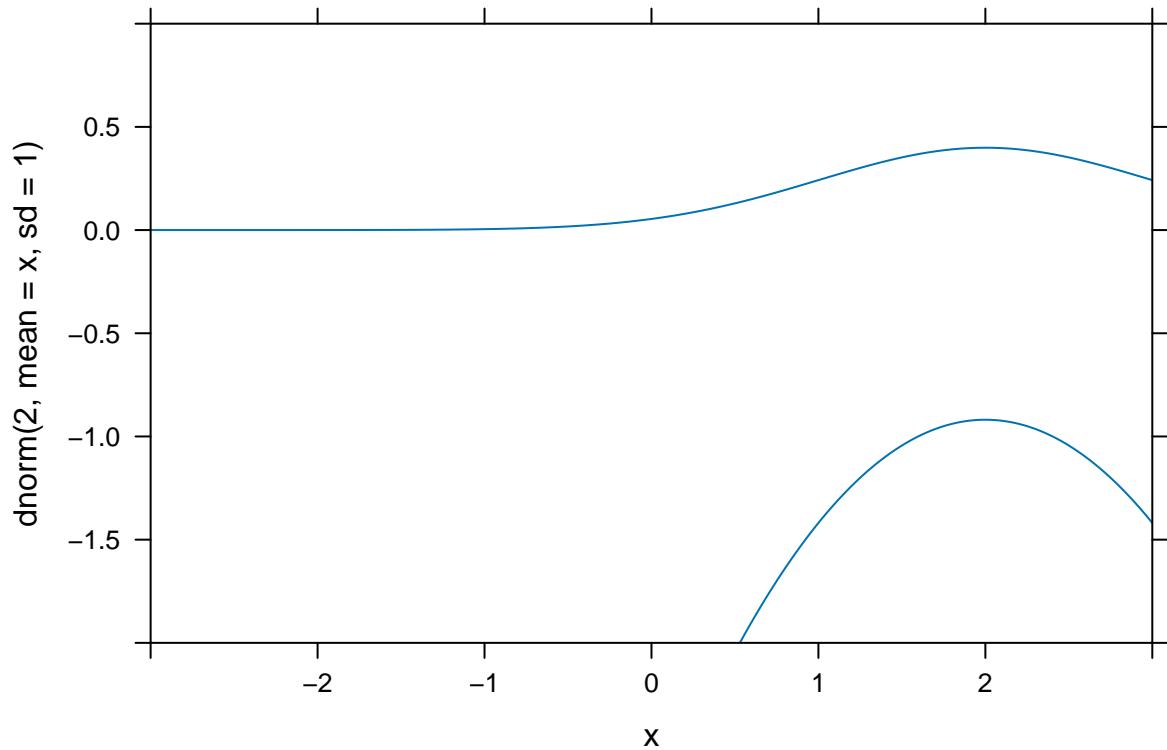
$$L(\mu, \sigma) = L(\mu, \sigma; X = x) = f(X = x; \mu, \sigma) \quad (7)$$

$$l(\mu, \sigma; X) = \log L(\mu, \sigma; X) = \log f(X; \mu, \sigma) \quad (8)$$

```
plotFun(dnorm(2, mean = x, sd = 1) ~ x, xlim = c(-3, 3), ylim = c(-2, 1));
```



```
plotFun(log(dnorm(2, mean = x, sd = 1)) ~ x, xlim = c(-3, 3), add = TRUE)
```

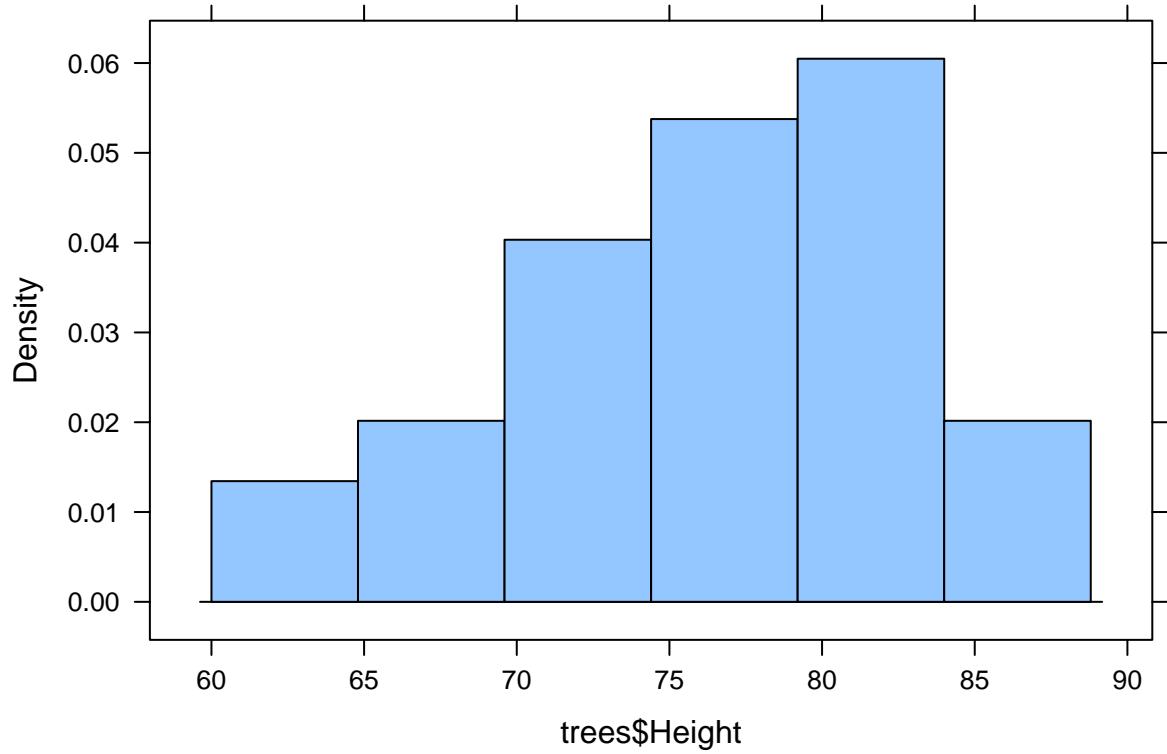


4 Example

```
trees <- read.delim("https://asta.math.aau.dk/datasets?file=trees.txt")
head(trees)

##   Girth Height Volume
## 1   8.3     70     10
## 2   8.6     65     10
## 3   8.8     63     10
## 4  10.5     72     16
## 5  10.7     81     19
## 6  10.8     83     20

histogram(trees$Height)
```



If Height is normally distributed, what are the parameters (mean and standard deviation)?

```

sd(trees$Height)

## [1] 6.4

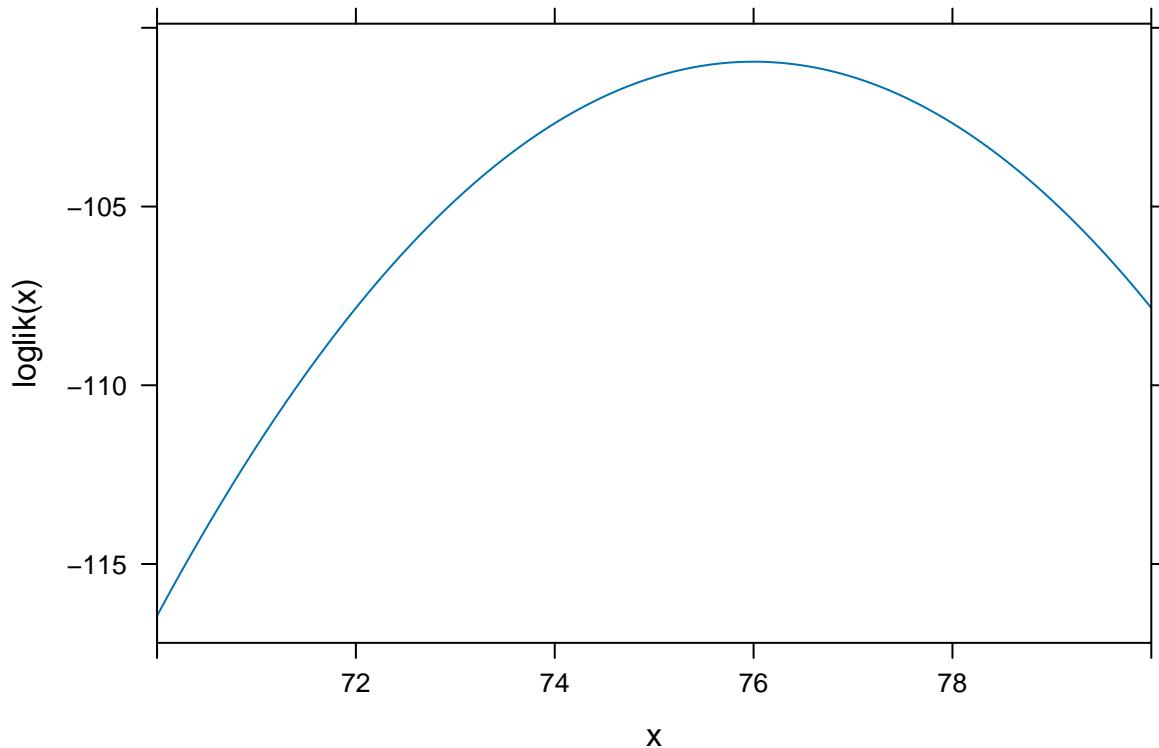
single_loglik <- function(mu) {
  sum(log(dnorm(trees$Height, mean = mu, sd = 6)))
}
single_loglik(65)

## [1] -153
single_loglik(c(60, 65, 70))

## [1] -159
loglik <- Vectorize(single_loglik)
loglik(c(60, 65, 70))

## [1] -211 -153 -116
plotFun(loglik(x) ~ x, xlim = c(70, 80))

```



Maximum around 76.

```
optimise(single_loglik, interval = c(0, 100))

## $minimum
## [1] 4.6e-05
##
## $objective
## [1] -2588

optimise(single_loglik, interval = c(0, 100), maximum = TRUE)

## $maximum
## [1] 76
##
## $objective
## [1] -101
```

Maximum of l is minimum of $-l$.

```
single_loglik <- function(mu) {
  -sum(log(dnorm(trees$Height, mean = mu, sd = 6)))
}
optimise(single_loglik, interval = c(0, 100))

## $minimum
## [1] 76
##
## $objective
## [1] 101
```

Both parameters:

```
single_loglik <- function(pars) {  
  mu <- pars[1]  
  sigma <- pars[2]  
  -sum(log(dnorm(trees$Height, mean = mu, sd = sigma)))  
}  
optim(c(1, 50), single_loglik)  
  
## Warning in dnorm(trees$Height, mean = mu, sd = sigma): NaNs produced  
  
## Warning in dnorm(trees$Height, mean = mu, sd = sigma): NaNs produced  
  
## Warning in dnorm(trees$Height, mean = mu, sd = sigma): NaNs produced  
  
## $par  
## [1] 76.0 6.3  
##  
## $value  
## [1] 101  
##  
## $counts  
## function gradient  
##      87      NA  
##  
## $convergence  
## [1] 0  
##  
## $message  
## NULL
```

```
single_loglik <- function(pars) {  
  mu <- pars[1]  
  sigma <- exp(pars[2])  
  -sum(log(dnorm(trees$Height, mean = mu, sd = sigma)))  
}  
mle <- optim(c(1, 50), single_loglik)  
mle  
  
## $par  
## [1] 76.0 1.8  
##  
## $value  
## [1] 101  
##  
## $counts  
## function gradient  
##      113      NA  
##  
## $convergence  
## [1] 0  
##  
## $message
```

```
## NULL
exp(mle$par[2])

## [1] 6.3
mean(trees$Height)

## [1] 76
sd(trees$Height)

## [1] 6.4
n <- length(trees$Height)
sd(trees$Height)

## [1] 6.4
```