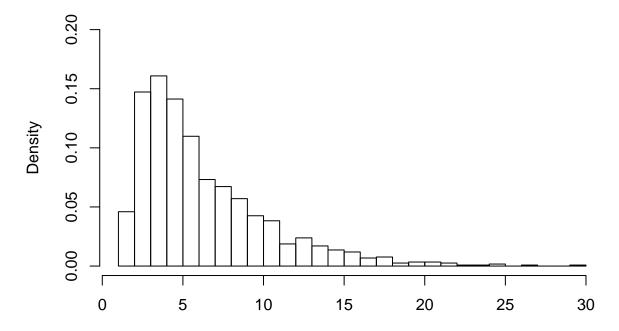
Assignment 2

1. Brain cell dataset

4.1: plot separately histograms for the distribution of the ramp spike time variable in the left and right hemisphere

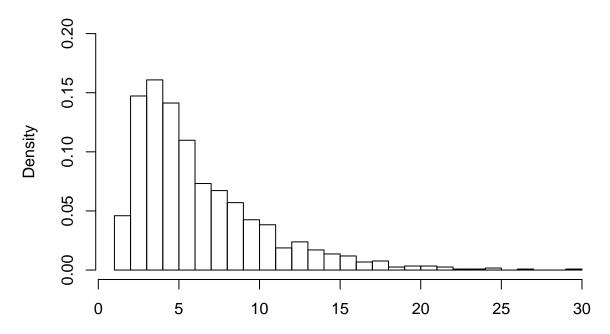
ramp spike time in left hemisphere



```
hist(left,
    breaks = 40,
```

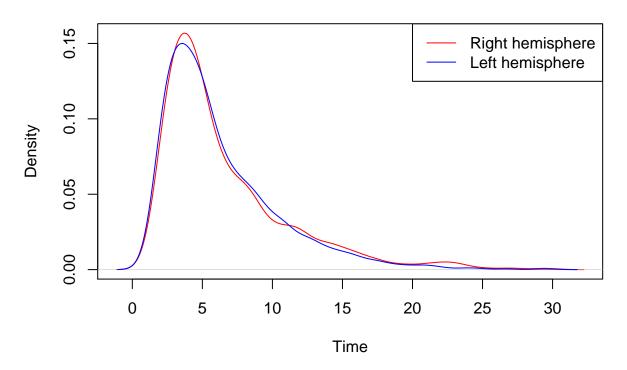
```
proba = T,
main = "ramp spike time in right hemisphere", xlab = "", ylim = c(0,0.2))
```

ramp spike time in right hemisphere



4.2: plot together kernel density estimations for the ramp spike time variable in the left and right hemisphere.

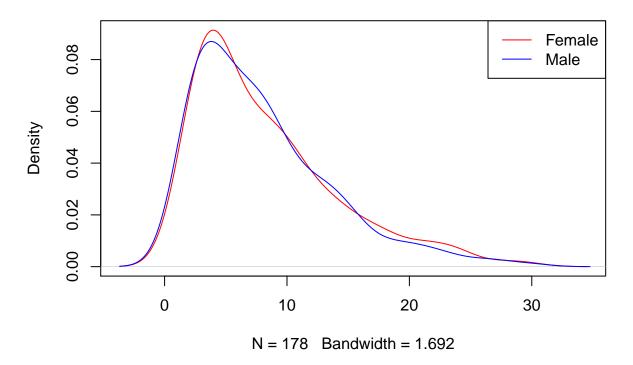
Kernel density estimation for the ramp spike time



4.3: plot together kernel density estimations for the ramp spike time variable for males and females.

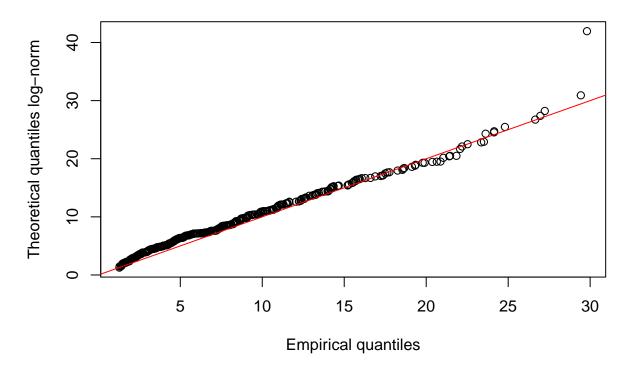
```
female_raw <-
  cells$ef__peak_t_ramp[cells$donor__species == "Homo Sapiens" &
  cells$donor sex == "Female"]
  female <- female_raw[!is.na(female_raw)]</pre>
  male_raw <-
  cells$ef__peak_t_ramp[cells$donor__species == "Homo Sapiens" &
  cells$donor__sex == "Male"]
  male <- male_raw[!is.na(male_raw)]</pre>
plot(density(female),
     col = "red",
     main = "Kernel Density Estimations for males and females")
lines(density(male),
      col = "blue")
legend("topright",
       legend = c("Female", "Male"),
       col = c("red", "blue"),
       lty = 1)
```

Kernel Density Estimations for males and females



Ex 5
5.1: Q-Q plot of ramp spike time distribution for humans, between empirical quantiles and theoretical ones (dlnorm distribution)

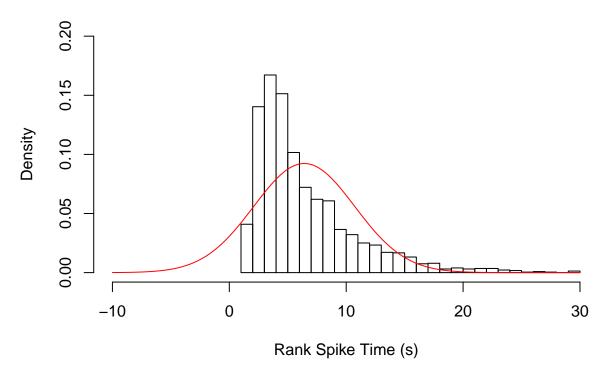
Q-Q Plot, ramp spike time d. and log-normal d.



5.2: Plot the Gaussian density on top of the histogram of the ramp spike time with better parameter values

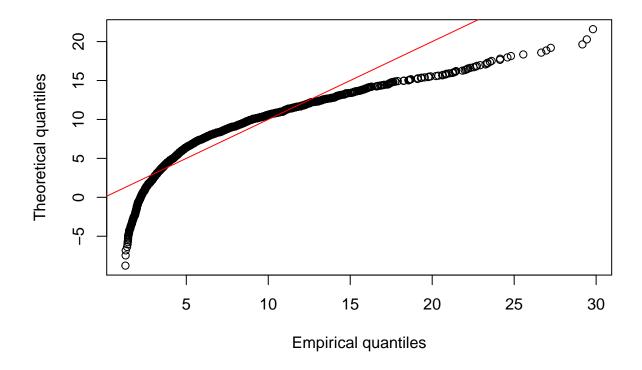
```
allcells <- cells$ef__peak_t_ramp[!is.na(cells$ef__peak_t_ramp)]
hist(allcells,
    breaks = 40,
    ylim = c(0, 0.2),
    xlim = c(-10, 30),
    probability = TRUE,
    main = "Rank Spike Time frequencies for all cells and d.norm estimation",
    xlab = "Rank Spike Time (s)")
curve(dnorm(x, mean = 6.41, sd = 4.32),
    col = "red",
    add = TRUE)</pre>
```

Rank Spike Time frequencies for all cells and d.norm estimation



> Our data daes nt have a left tail and the peak of the histogram is much larger than the curve of the norm distribution.

5.3: Q-Q plot for the ramp spike time against the Gaussian distribution with these new parameter values



We can visually observe that our data don't follow a Gaussian distribution since several quantiles in the tails don't match in the diagonal.

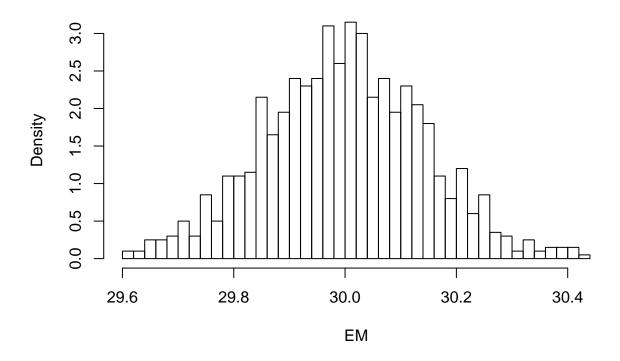
2. Empirical mean and variance

6.1: Sample n values of $X \sim Bi(parameter\ size=100,\ prob=0.3)$ for $n=10,\ 100,\ 1000,\ 1000.$

```
n = 100
               # n. of experiments (size or i.e. nb of coin tosses)
p = 0.3
               # p. of success (i.e. head)
EM = c()
EV = c()
Esd = c()
rep = c()
for (x in c(10, 100, 1000, 10000)){
                                            \# x = nb of sampling repetitions
  sample <- rbinom(x, size = n, prob = p)</pre>
  EM <- c(EM, mean(sample))</pre>
                                             # empirical mean
  EV <- c(EV, var(sample))</pre>
                                             # empirical variance
  Esd <- c(Esd, sd(sample))
                                             # empirical standard deviation
                                             # n. of sampling repetition
  rep <- c(rep, x)
}
mean <- n * p
                                      # true or theoretical mean
var <- n * p * (1-p)
                                      # true or theoretical variance
sd <- sqrt(var)</pre>
                                      # true or theoretical standard deviation
```

6.2: Write a function to repeat the sampling and estimation for n = 1000 and plot.

```
samp_est_f <- function(nb_rep){</pre>
 nb_trials = 100
 p = 0.3
 EM = c()
  EV = c()
 Esd = c()
  for (i in 1:1000){
   sample <- rbinom(nb_rep, size = nb_trials, prob = p)</pre>
   EM <- c(EM, mean(sample)) # that's an easy way to "append" values to a vector
 }
  hist (EM,
       probability = TRUE,
       breaks = 50)
                             # histogram of the distribution of the empirical means
  data = list()
                               # list to store the output of the function
  data$EM <- EM
                                # means distributions for 6.4
  data$sample <- sample</pre>
                               # the last sample generated
  data$rep <- nb_rep</pre>
                               # n. of repetition
  data$var_EM <- var(EM)</pre>
                               # (emp.) variance of empirical/sample means
  data$sd_EM <- sd(EM)
                                # (emp.) standard deviation of the empirical/sample means
  cat("rep:", data$rep, # cat() to print numbers and strings in the same line
      ", var_EM:", data$var_EM,
      ", sd_EM:", data$sd_EM)
  return(data)
}
data <- samp_est_f(1000)</pre>
```



```
## rep: 1000 , var_EM: 0.02079311 , sd_EM: 0.1441982
```

6.3: Compute the standard error of the mean

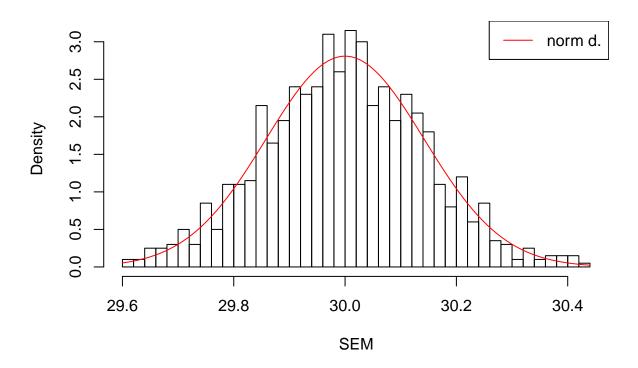
```
SEM_f <- function(x){
    SEM <- (sd(data$sample) / sqrt(length(data$sample) - 1))
    return(SEM)
}
data$SEM <- SEM_f(1000)
cat("sd_EM:", data$sd_EM, ", SEM:", data$SEM)

## sd_EM: 0.1441982 , SEM: 0.1420409

# SEM = ESTIMATOR of the error between the sample means and the population mean
# sd_EM = standard deviation of the sample means</pre>
```

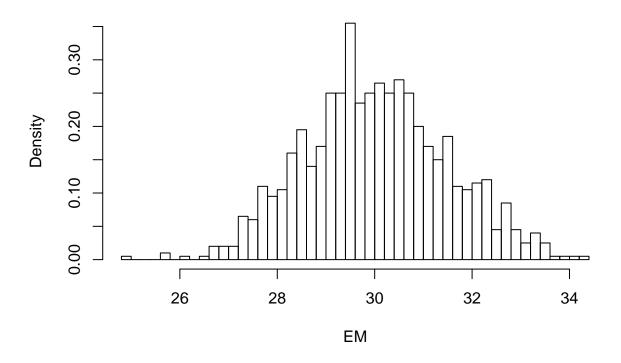
6.4: draw the density of Gaussian distribution with parameters mean = 100 \times 0.3 and sd equal to the sem

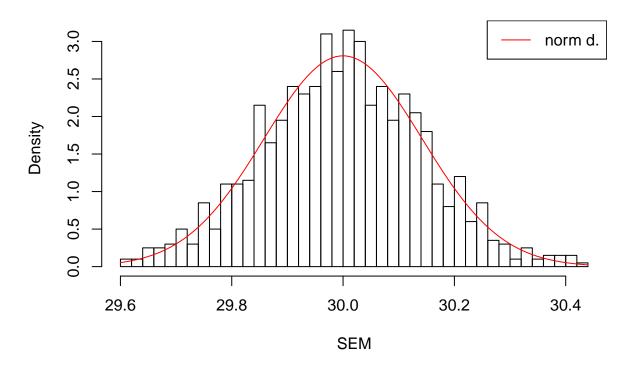
```
plot_f <- function(x){
  hist(data$EM,
      probability = TRUE,
      breaks = 50,
      xlab = "SEM")
  curve(dnorm(x, mean = 100 * 0.3, sd = data$SEM),
      col = "red",</pre>
```



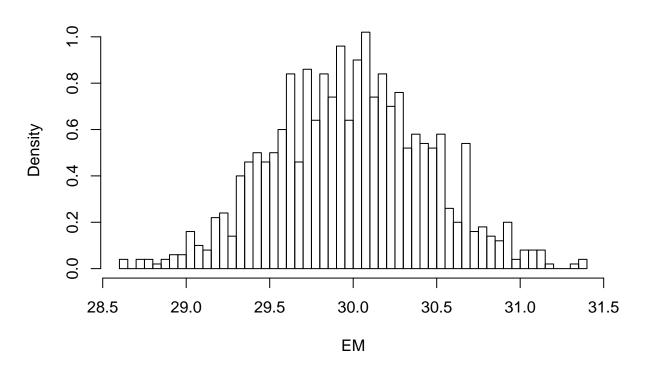
check the result for different values of x:

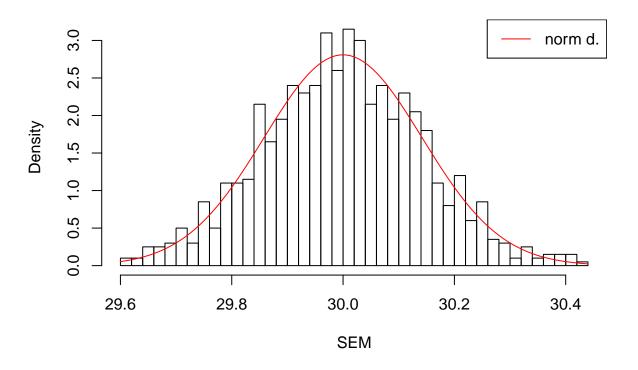
```
data_10 <- samp_est_f(10)
```





data_100 <- samp_est_f(100)





data_10000 <- samp_est_f(10000)

