## 2nd Delivery

## Eduardo Alarcón & Alfonso Pineda

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```
library(readxl)
SpotifySongs <- read_excel("songstats.xlsx")</pre>
```

## Model Fitting.

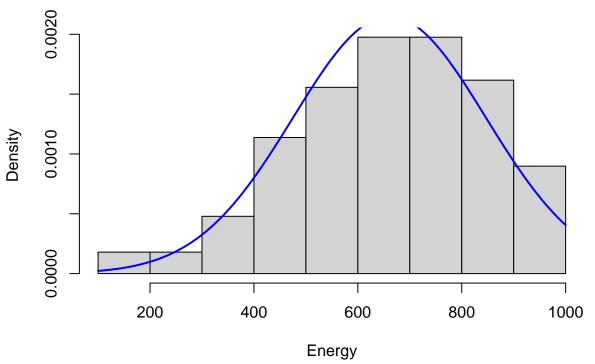
By just watching the histogram, we can suppose that our main variable, the energy, will follow a normal distribution

```
SpotifySongs <- read_excel("songstats.xlsx")
suppressWarnings(library(summarytools))
energy <- SpotifySongs$energy
descr(energy)</pre>
```

```
## Descriptive Statistics
## energy
## N: 167
##
##
                        energy
##
                        660.92
##
                Mean
##
             Std.Dev
                        185.68
##
                 Min
                        104.00
##
                   01
                        551.00
              Median
                        667.00
##
##
                  QЗ
                        804.00
##
                 Max
                        993.00
##
                 MAD
                        188.29
                  IQR
##
                        249.50
                   CV
##
                          0.28
                         -0.44
##
            Skewness
##
         SE.Skewness
                          0.19
##
            Kurtosis
                          0.07
##
             N.Valid
                        167.00
           Pct.Valid
                        100.00
```

```
hist(energy, probability = TRUE, xlab = "Energy")
curve(dnorm(x, mean(energy), sd(energy)), col="blue", lwd=2, add=TRUE, yaxt="n")
```





```
Partition
## $breaks
   [1]
##
        100 200 300 400
                            500 600 700 800 900 1000
##
## $counts
## [1] 3 3 8 19 26 33 33 27 15
##
## $density
## [1] 0.0001796407 0.0001796407 0.0004790419 0.0011377246 0.0015568862
## [6] 0.0019760479 0.0019760479 0.0016167665 0.0008982036
##
## $mids
## [1] 150 250 350 450 550 650 750 850 950
##
## $xname
## [1] "energy"
## $equidist
## [1] TRUE
##
## attr(,"class")
## [1] "histogram"
```

```
## Loading required package: MASS
## Loading required package: survival
```

library(fitdistrplus)

Partition <- hist(energy, plot=FALSE)</pre>

```
normalfit <- fitdist(energy, "norm")</pre>
normalfit
## Fitting of the distribution ' norm ' by maximum likelihood
## Parameters:
##
        estimate Std. Error
## mean 660.9222
                  14.32480
## sd
        185.1206
                  10.12919
CummulativeProbabilities = pnorm(c(-Inf, Partition\$breaks[c(-1, -10)], Inf),
                                  normalfit$estimate[1], normalfit$estimate[2])
Probabilities = diff(CummulativeProbabilities)
Expected = length(energy)*Probabilities
chisq.test(Partition$counts, p=Probabilities)
## Warning in chisq.test(Partition$counts, p = Probabilities): Chi-squared
## approximation may be incorrect
##
   Chi-squared test for given probabilities
##
## data: Partition$counts
## X-squared = 5.9441, df = 8, p-value = 0.6535
pchisq(5.9441, 6, lower.tail = FALSE)
## [1] 0.4294811
As we can see from the previous test, the normal distribution model can be used as the p-value is less than
0.05. We can confirm this data by using the following texts
library(nortest)
ad.test(energy)
##
##
   Anderson-Darling normality test
##
## data: energy
## A = 0.46126, p-value = 0.2563
cvm.test(energy)
##
##
   Cramer-von Mises normality test
##
## data: energy
## W = 0.053358, p-value = 0.4607
lillie.test(energy)
##
## Lilliefors (Kolmogorov-Smirnov) normality test
##
## data: energy
## D = 0.040216, p-value = 0.7325
pearson.test(energy)
```

##

```
Pearson chi-square normality test
##
## data: energy
## P = 10.725, p-value = 0.6339
sf.test(energy)
##
##
     Shapiro-Francia normality test
##
## data: energy
## W = 0.98224, p-value = 0.03169
plot(normalfit)
                                                                           Q-Q plot
        Empirical and theoretical dens.
                                                    Empirical quantiles
     0.0020
                                                         800
Density
     0.0000
                                                         200
             200
                     400
                                                                200
                                                                            600
                                                                                   800 1000
                            600
                                    800
                                          1000
                                                                      400
                          Data
                                                                      Theoretical quantiles
        Empirical and theoretical CDFs
                                                                           P-P plot
                                                    Empirical probabilities
     9.0
                                                         9.0
CDF
     0.0
             200
                     400
                            600
                                    800
                                          1000
                                                              0.0
                                                                     0.2
                                                                            0.4
                                                                                   0.6
                                                                                          8.0
                                                                                                 1.0
```

Data

Theoretical probabilities