Lista 1

Modelagem com Apoio Computacional

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Carregamento dos dados

```
fatigue_df <- data.frame(work_MJ_m3 = c(11.5,13.0,14.3,15.6,16.0,17.3,19.3,
21.1,21.5,22.6,22.6,24.0,24.0,24.6,25.2,25.5,26.3,27.9,28.3,28.4,28.6,30.9,
31.9,34.5,40.1,40.1,43.0,44.1,46.5,47.3,48.7,52.9,56.6,59.9,60.2,60.3,60.5,
62.1,62.8,66.5,67.0,67.1,67.9,68.8,75.4,100.5),life_cycles = c(3280,5046,
1563,4707,977,2834,2266,2208,1040,700,1583,482, 804,1093,1125,884,1300,
852,580,1066,1114,386,745,736, 750,316,456,552,355,242,190,127,185,255,
195, 283,212,327,373,125,187,135,245,137,200,190))</pre>
```

```
work_MJ_m3 life_cycles
1
        11.5
                      3280
2
        13.0
                     5046
3
        14.3
                     1563
4
        15.6
                     4707
5
        16.0
                      977
        17.3
                     2834
```

Ajuste do modelo com base na moda

```
mle_moda_bs <- function(x, t) {
   x <- as.matrix(x)
   t <- as.matrix(t)</pre>
```

```
fit <- lm.fit(x, log(t))</pre>
beta <- c(fit$coef)</pre>
k <- length(beta)</pre>
n <- length(t)
mu init <- x %*% beta
phi_init <- alpha^2 #chute inicial</pre>
thetaStar <- c(beta, phi_init)</pre>
loglik_moda <- function(par) {</pre>
  log_moda_Y <- x %*% par[1:k]
  phi_param <- par[k+1]</pre>
  if (phi_param <= 0 || phi_param >= 1) {
   return(NA)}
  alpha_art <- sqrt(phi_param) #proposta do artifo</pre>
  beta_art <- exp(log_moda_Y) / (1 - phi_param) #proposta do artigo</pre>
  #pdf da distribuição log-BS
  z <- (sqrt(t/beta_art) - sqrt(beta_art/t))/alpha_art</pre>
  pdf_bs <- (1/(alpha_art*sqrt(t)))*(sqrt(t/beta_art) +</pre>
  sqrt(beta_art/t))*dnorm(z)
  pdf_bs[pdf_bs <= 0] <- .Machine$double.xmin</pre>
  #substitui valor menor ou igual a O por um número muito pequeno
  #log-verossimilhança
 1_i <- log(pdf_bs)</pre>
  if (any(!is.finite(l_i))) {
   return(.Machine$double.xmax)}
    #retorna um valor muito grande se algum log for infinito
  #verossimilhança total (loglik-plus)
  return(-sum(l_i))}
est <- optim(</pre>
 par = thetaStar,
 fn = loglik_moda,
 method = "BFGS",
```

```
hessian = TRUE,
  control = list(maxit = 2000, reltol = 1e-12))
if (est$conv != 0) {
  warning("FUNCTION DID NOT CONVERGE!")}
coef <- (est$par)[1:k]</pre>
phi_est <- est$par[k + 1]</pre>
moda_hat_log <- x %*% coef</pre>
moda_hat <- exp(moda_hat_log)</pre>
SHess = solve(est$hessian)
SE = sqrt(diag(SHess))
tval = est$par / SE
matcoef = cbind(est$par, SE, tval, 2 * (1 - pnorm(abs(tval))))
AIC <- 2 * \text{ est}$value + 2 * (k + 1)
BIC <- 2 * \text{est}$value + (k + 1) * \log(n)
result <- list(
  phiHat = phi_est,
  betaHat_log_moda = coef,
  modaHat = moda_hat,
  AIC = AIC,
  BIC = BIC,
  matcoef = matcoef)
return(result)}
```

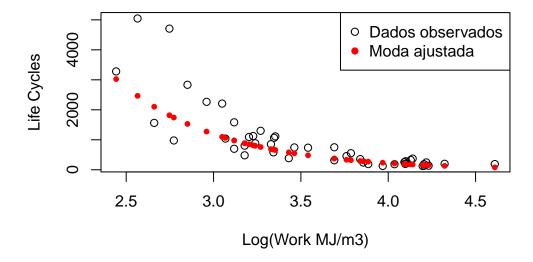
Estimativas do modelo

```
Zx <- model.matrix(~ log(fatigue_df$work_MJ_m3))
Zy <- fatigue_df$life_cycles
fit_moda <- mle_moda_bs(x = Zx , t = Zy)
fit_moda$matcoef</pre>
```

SE tval

```
(Intercept) 12.095313 0.39169573 30.879358 0.00000e+00 log(fatigue_df$work_MJ_m3) -1.670763 0.10844480 -15.406574 0.00000e+00 0.168396 0.03510972 4.796278 1.61641e-06
```

Dados observados e moda ajustada

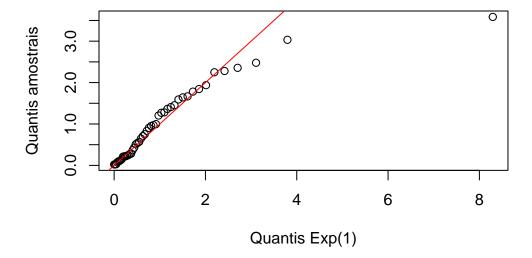


Verificação de ajuste do modelo

```
S_bs_moda <- function(y, mu, phi) {
  alpha <- sqrt(phi)
  beta <- mu / (1 - phi)
  xi <- (sqrt(y/beta) - sqrt(beta/y)) / alpha
  S <- 1 - pnorm(xi)</pre>
```

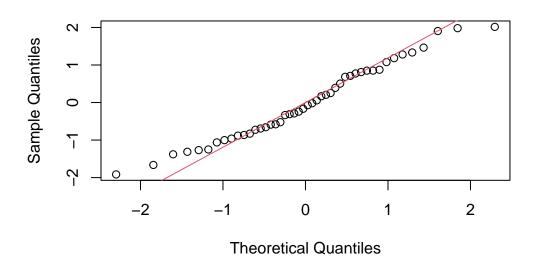
```
return(S)}
residuos_bs_moda <- function(y, mu_hat, phi_hat) {</pre>
  S <- S_bs_moda(y, mu_hat, phi_hat)</pre>
  r_{cox} \leftarrow -log(S)
  r_quant <- qnorm(S)</pre>
  list(coxsnell = r_cox, quantile = r_quant)}
res_moda <- residuos_bs_moda(</pre>
  y = fatigue_df$life_cycles,
  mu_hat = fit_moda$modaHat,
  phi_hat = fit_moda$phiHat)
# QQ plot
a <- ppoints(2000)
QGG <- qexp(a)
qqplot(QGG, res_moda$coxsnell,
       xlab = "Quantis Exp(1)", ylab = "Quantis amostrais",
       main = "QQ-plot Residuos")
abline(0,1,col="red")
```

QQ-plot Resíduos



```
# QQ plot quantilicos
qqnorm(res_moda$quantile); qqline(res_moda$quantile, col=2)
```

Normal Q-Q Plot



Os dois gráficos são bem similares aos gráficos para o modelo baseado na média. Indicando também adequação do modelo com base na moda.

Comparação dos modelos

Moda

```
SE tval
(Intercept) 12.095313 0.39169573 30.879358 0.00000e+00
log(fatigue_df$work_MJ_m3) -1.670763 0.10844480 -15.406574 0.00000e+00
0.168396 0.03510972 4.796278 1.61641e-06
```

Média

```
se.coef tval
(Intercept) 12.2797340 0.3893978 31.535187 0
log(fatigue_df$work_MJ_m3) -1.6707690 0.1084439 -15.406766 0
0.4103574 0.0427819 9.591846 0
```