

Document title

Subtitle

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Abstract

Lorem ipsum dolor sit amet, consectetur adipiscing elit. Aenean ut elit odio. Donec fermentum tellus neque, vitae fringilla orci pretium vitae. Fusce maximus finibus facilisis. Donec ut ullamcorper turpis. Donec ut porta ipsum. Nullam cursus mauris a sapien ornare pulvinar. Aenean malesuada molestie erat quis mattis. Praesent scelerisque posuere faucibus. Praesent nunc nulla, ullamcorper ut ullamcorper sed, molestie ut est. Donec consequat libero nisi, non semper velit vulputate et. Quisque eleifend tincidunt ligula, bibendum finibus massa cursus eget. Curabitur aliquet vehicula quam non pulvinar. Aliquam facilisis tortor nec purus finibus, sit amet elementum eros sodales. Ut porta porttitor vestibulum.

1 Introdução

2 Metodologia

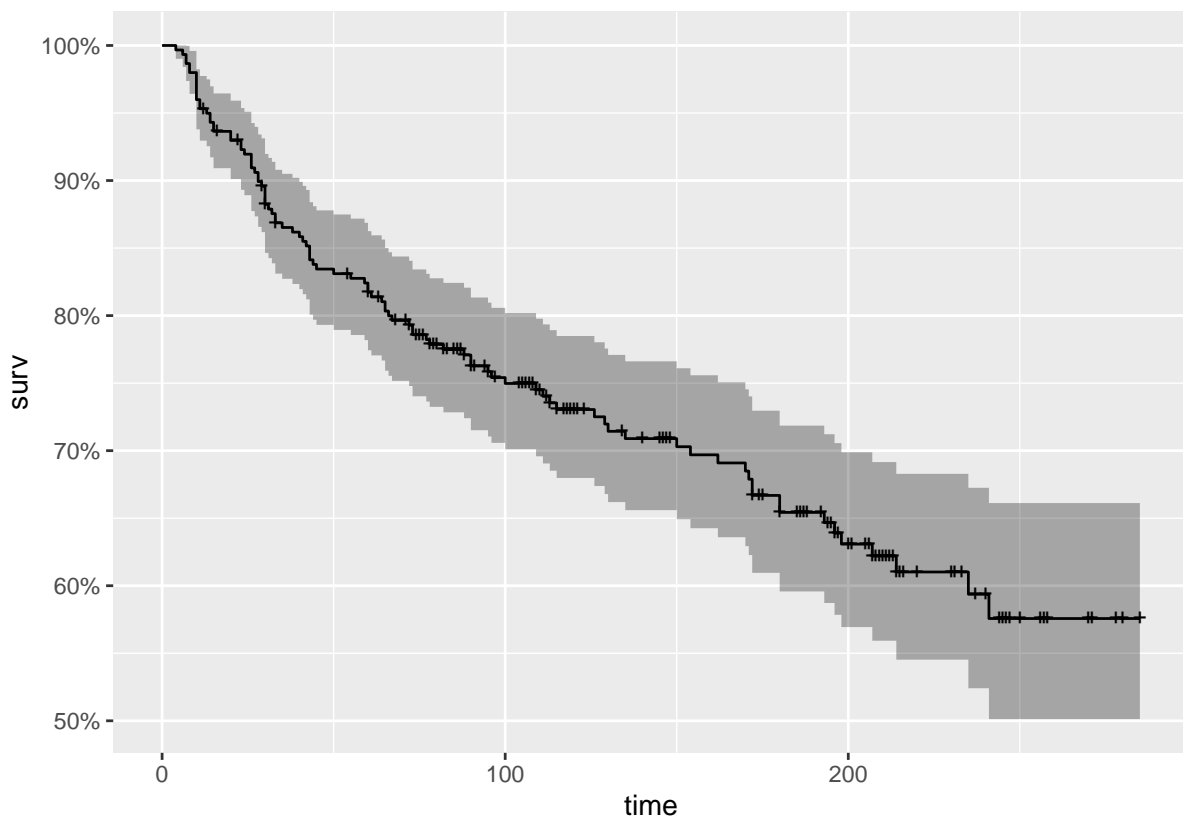
3 Resultados

3.1 Analise exploratoria

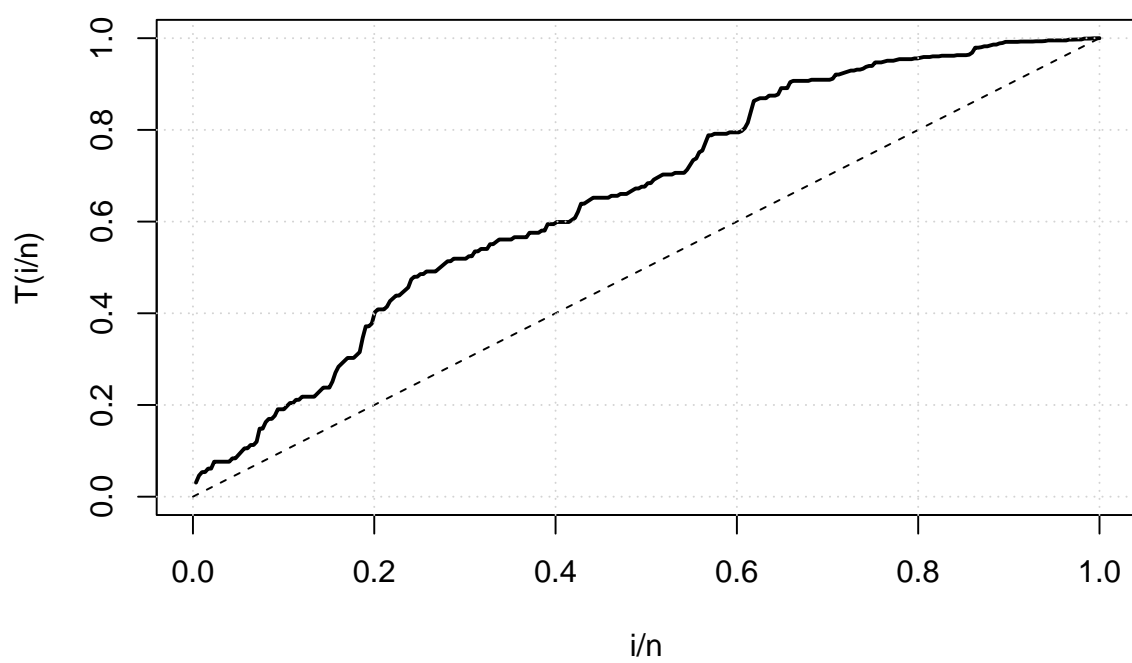
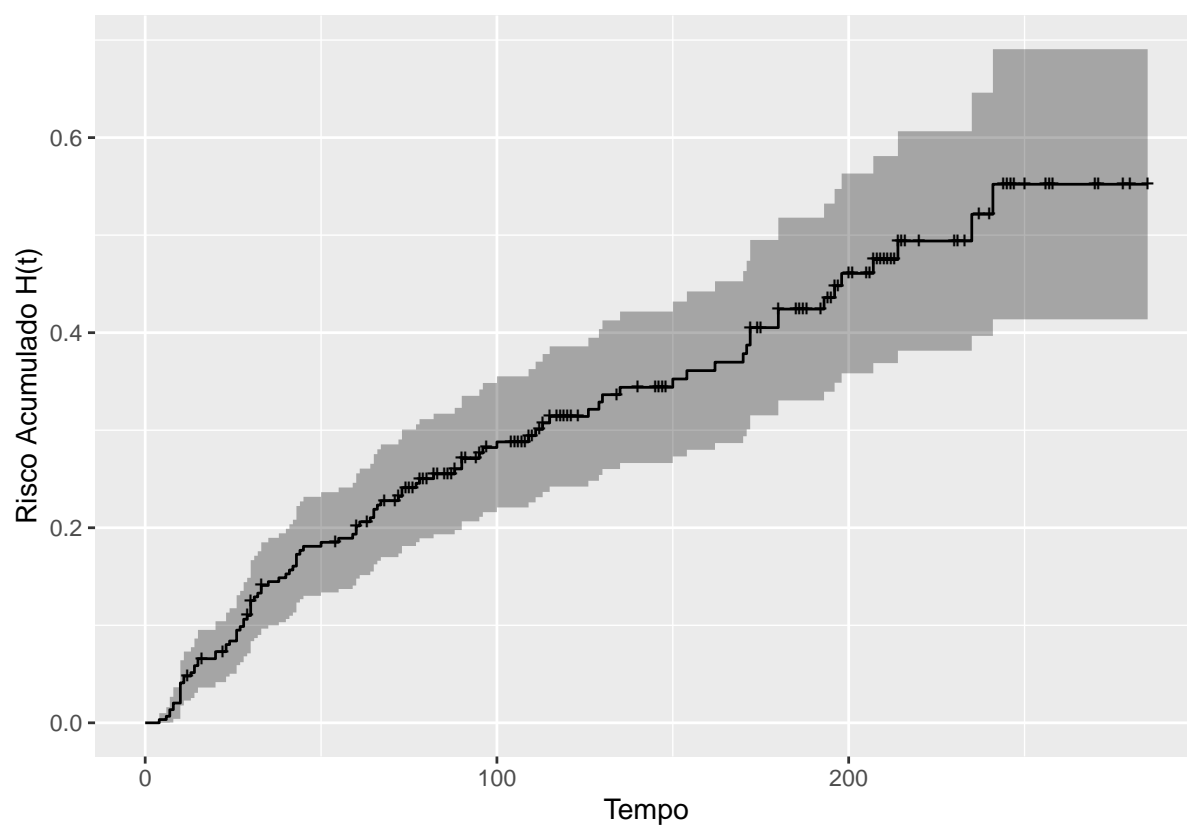
Antes de ajustar o modelo, é necessário estudarmos o comportamento dos dados antes para poder identificar qual distribuição será mais adequada e já realizar uma seleção das variáveis categóricas que são significativas para o modelo final.

```
## # A tibble: 6 x 13
##   age anaemia creatinine_phosphokinase diabetes ejection_fraction
##   <dbl> <dbl>                <dbl>    <dbl>          <dbl>
## 1    75      0                582      0            20
## 2    55      0                7861     0            38
## 3    65      0                146      0            20
## 4    50      1                111      0            20
## 5    65      1                160      1            20
## 6    90      1                 47      0            40
## # i 8 more variables: high_blood_pressure <dbl>, platelets <dbl>,
## #   serum_creatinine <dbl>, serum_sodium <dbl>, sex <dbl>, smoking <dbl>,
## #   censura <dbl>, tempo <dbl>
```

Modelo de sobrevivencia não paramétrico de Kaplan-Meier

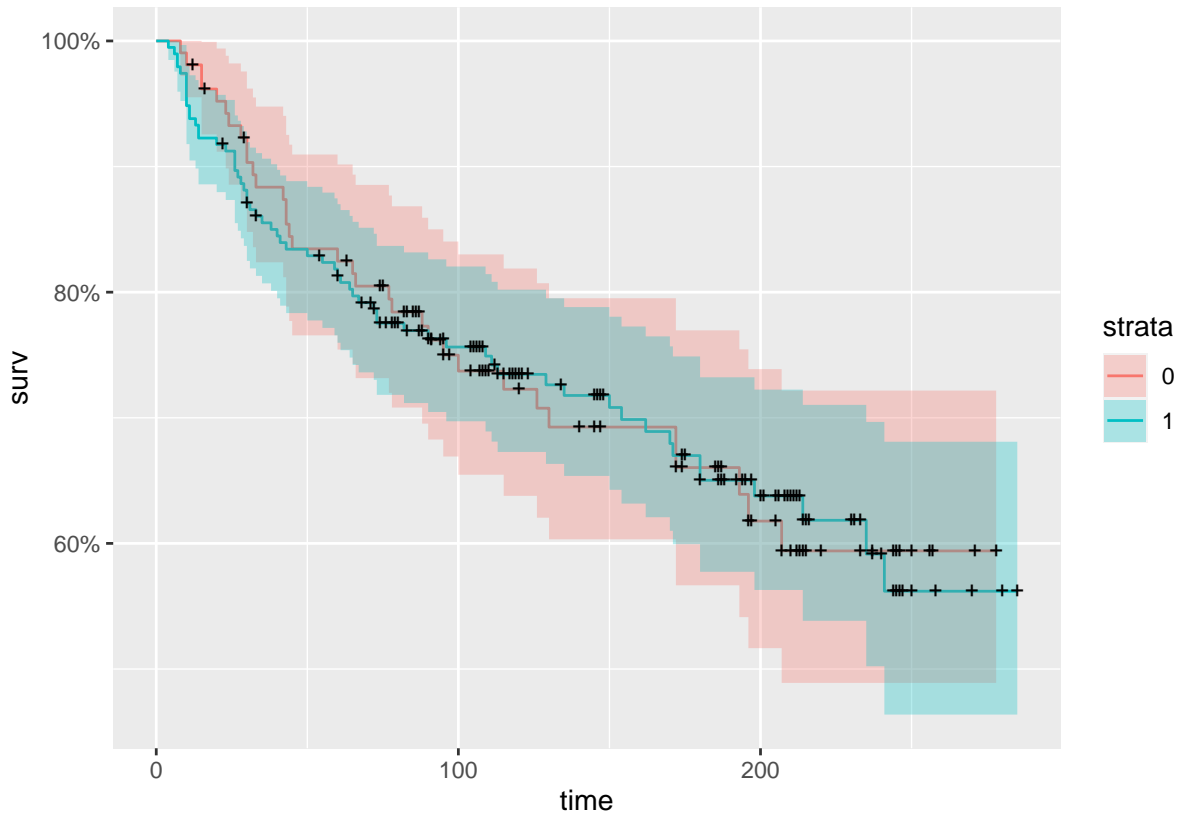


A função de risco

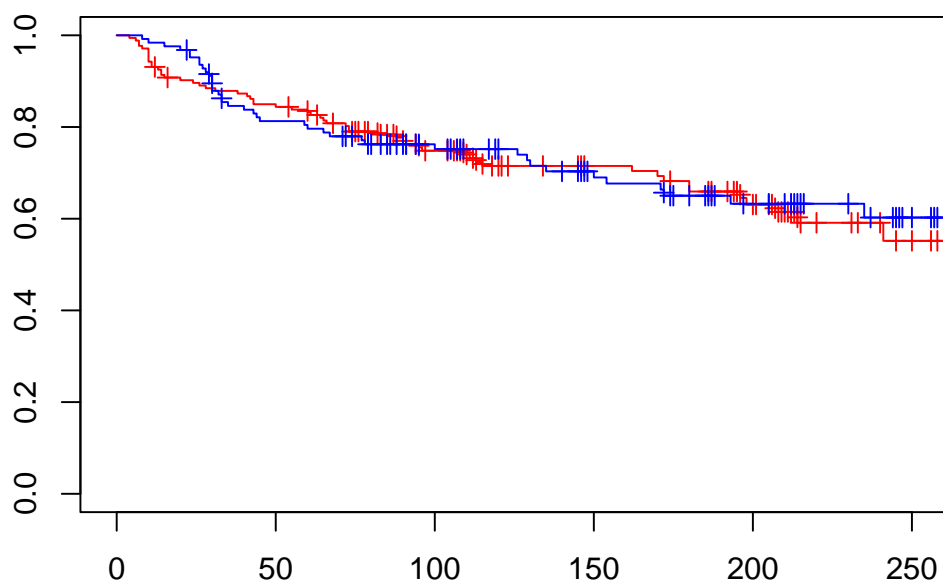


3.1.1 Análise das Variáveis Categóricas

3.1.1.1 Variável Sex Vamos comparar as curvas de sobrevivências divididas por Sexo, com o objetivo de ver se essa variável influencia na curva de sobrevivência. Em seguida iremos fazer um teste para verificar a diferença entre as curvas.

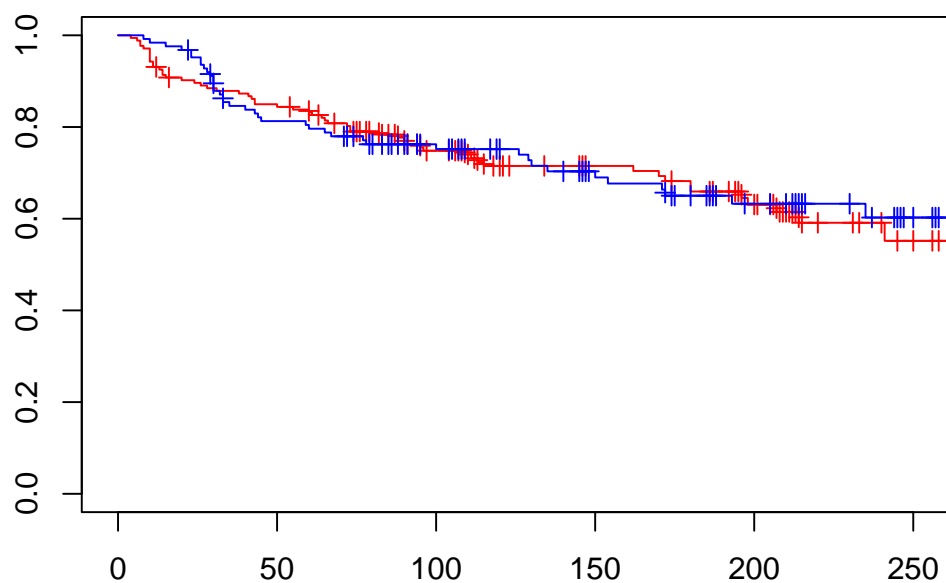


```
## Call:
## survdiff(formula = Surv(tempo, censura) ~ sex, data = dados,
##          rho = 1)
##
##               N Observed Expected (O-E)^2/E (O-E)^2/V
## sex=0 105      27.9      28.5  0.01467  0.0271
## sex=1 194      52.0      51.4  0.00814  0.0271
##
## Chisq= 0 on 1 degrees of freedom, p= 0.9
```



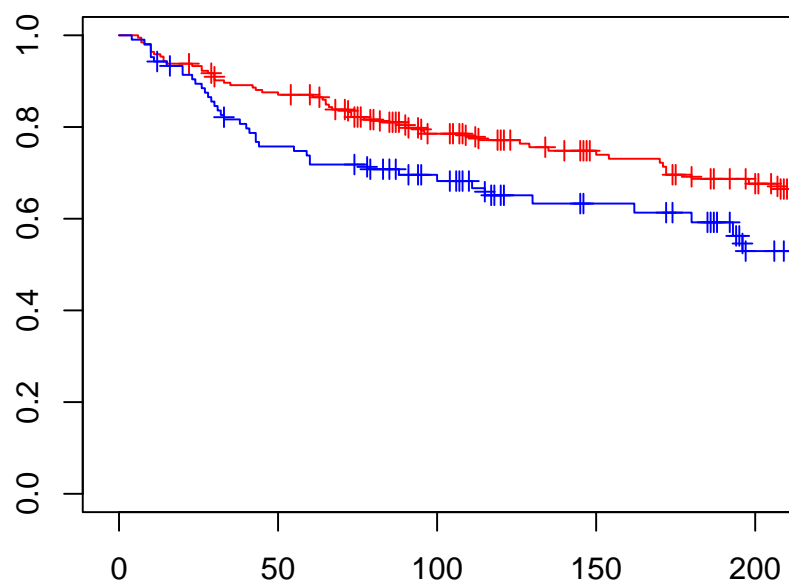
3.1.1.2 Variavel Diabetes

```
## Call:
## survdiff(formula = Surv(tempo, censura) ~ diabetes, data = dados,
##          rho = 1)
##
##              N Observed Expected (O-E)^2/E (O-E)^2/V
## diabetes=0 174    46.7    45.9    0.0125    0.0349
## diabetes=1 125    33.2    34.0    0.0168    0.0349
##
## Chisq= 0 on 1 degrees of freedom, p= 0.9
```



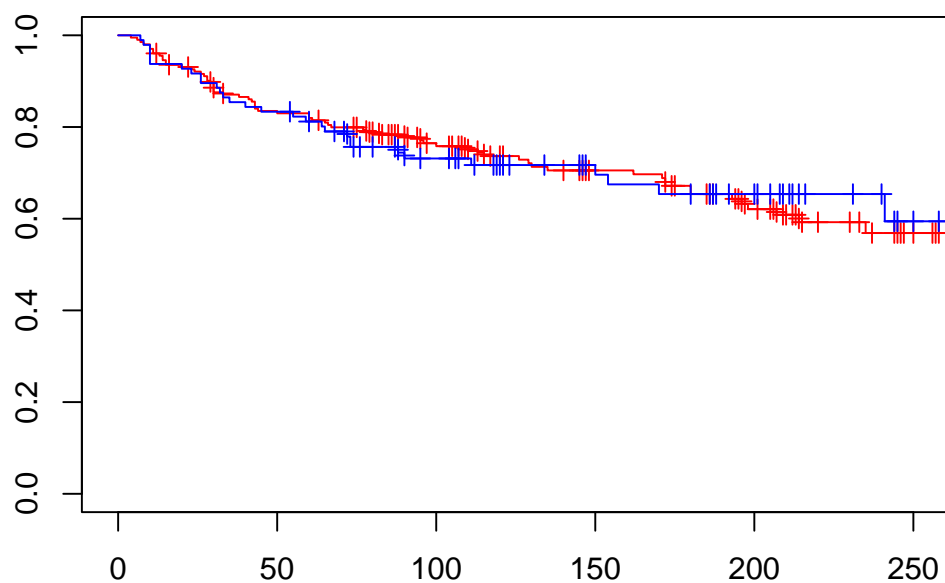
3.1.1.3 Variavel Anaemia

```
## Call:
## survdiff(formula = Surv(tempo, censura) ~ diabetes, data = dados,
##          rho = 1)
##
##              N Observed Expected (O-E)^2/E (O-E)^2/V
## diabetes=0 174      46.7      45.9    0.0125    0.0349
## diabetes=1 125      33.2      34.0    0.0168    0.0349
##
## Chisq= 0  on 1 degrees of freedom, p= 0.9
```

3.1.1.4 Variavel High Blood Pressure

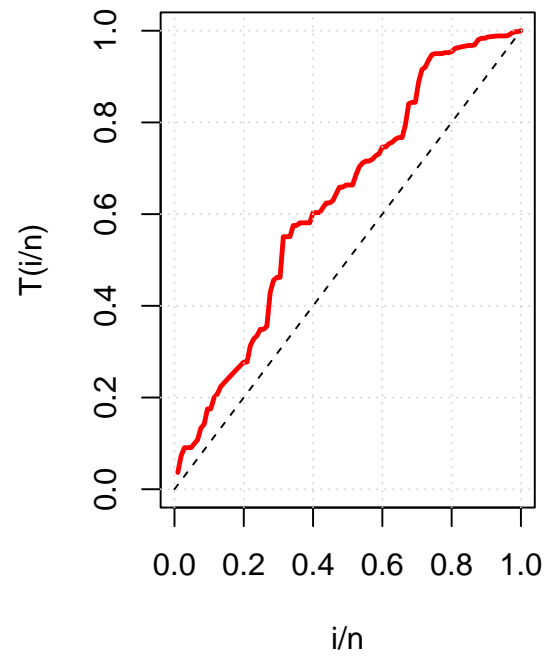
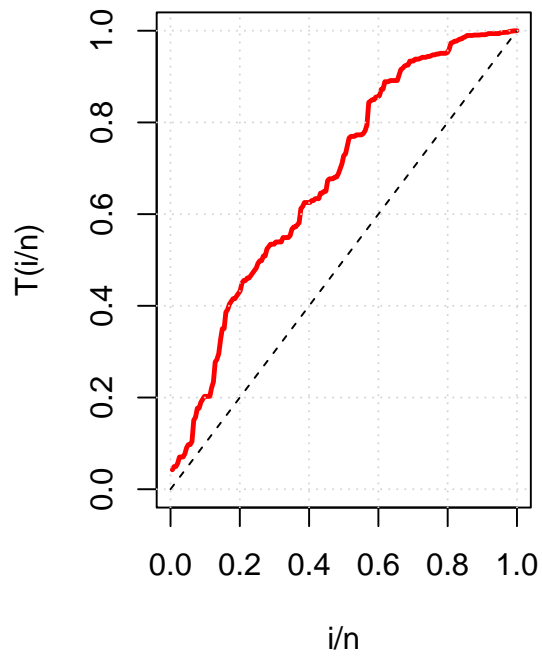
```
## Call:
## survdiff(formula = Surv(tempo, censura) ~ high_blood_pressure,
##          data = dados, rho = 1)
##
##               N Observed Expected (O-E)^2/E (O-E)^2/V
## high_blood_pressure=0 194      46.6      54.8      1.25      4.71
## high_blood_pressure=1 105      33.3      25.1      2.72      4.71
##
## Chisq= 4.7  on 1 degrees of freedom, p= 0.03
```



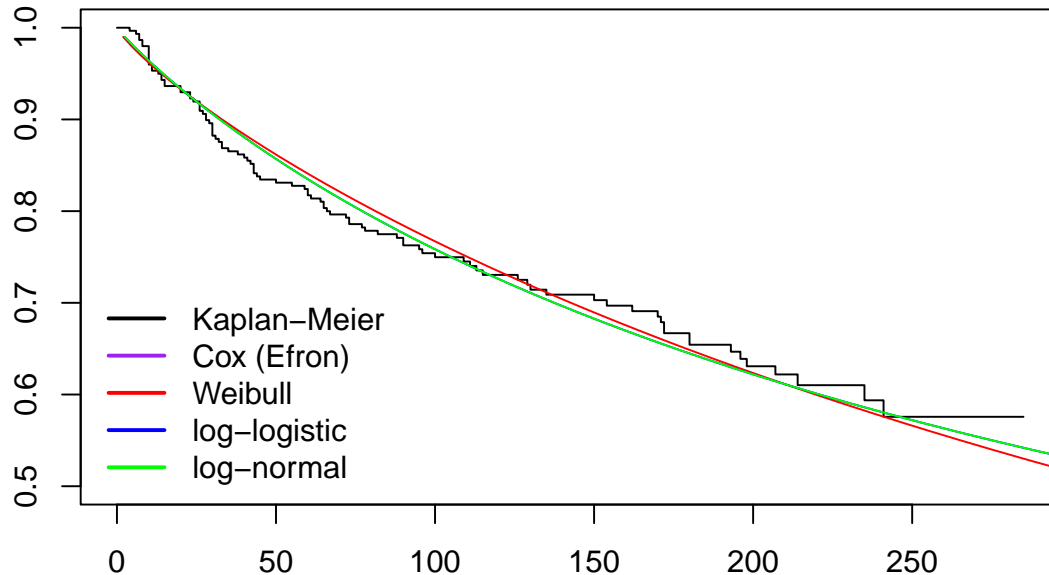
3.1.1.5 Variavel Smoking

```
## Call:
## survdiff(formula = Surv(tempo, censura) ~ smoking, data = dados,
##          rho = 1)
##
##              N Observed Expected (O-E)^2/E (O-E)^2/V
## smoking=0 203     54.5     54.7  0.000902  0.00339
## smoking=1  96     25.5     25.2  0.001955  0.00339
##
## Chisq= 0 on 1 degrees of freedom, p= 1
```

3.1.2 Curvas TTT



3.1.3 Seleção da distribuição



```
## Weibull ~( 0.8333038 , 491.7358 )
```

```
##           AICws  AICcws  BICws
## [1,] 1344.876 1344.916 1352.277
```

```
##
## Call:
## survreg(formula = s ~ 1, data = dados, dist = "weibull")
##           Value Std. Error      z      p
## (Intercept) 6.1979      0.1638 37.83 <2e-16
## Log(scale)  0.1824      0.0923  1.98  0.048
##
## Scale= 1.2
##
## Weibull distribution
## Loglik(model)= -670.4  Loglik(intercept only)= -670.4
## Number of Newton-Raphson Iterations: 5
## n= 299
```

```
## Log-Normal ~( 5.914729 , 1.916652 )
```

```
##           AIClns  AICclns  BIClns
## [1,] 1336.546 1336.587 1343.947
```

```
##
## Call:
## survreg(formula = s ~ 1, data = dados, dist = "loglogistic")
##              Value Std. Error      z      p
## (Intercept)  5.8326      0.1613 36.17 <2e-16
## Log(scale)   0.0703      0.0897  0.78   0.43
##
## Scale= 1.07
##
## Log logistic distribution
## Loglik(model)= -669.2   Loglik(intercept only)= -669.2
## Number of Newton-Raphson Iterations: 4
## n= 299

## Log-Normal ~( 0.9320725 , 341.255 )

##           AIClls  AICc1ls  BIC1ls
## [1,] 1342.334 1342.375 1349.735

##
## Call:
## survreg(formula = s ~ 1, data = dados, dist = "loglogistic")
##              Value Std. Error      z      p
## (Intercept)  5.8326      0.1613 36.17 <2e-16
## Log(scale)   0.0703      0.0897  0.78   0.43
##
## Scale= 1.07
##
## Log logistic distribution
## Loglik(model)= -669.2   Loglik(intercept only)= -669.2
## Number of Newton-Raphson Iterations: 4
## n= 299

##           AICws  AICcws  BICws
## [1,] 1344.876 1344.916 1352.277

##           AIClns  AICc1ns  BIClns
## [1,] 1336.546 1336.587 1343.947

##           AIClls  AICc1ls  BIC1ls
## [1,] 1342.334 1342.375 1349.735
```

4 Conclusão