

Weibull Regression with R, Part Two*

```

> rm(list=ls()); options(scipen=999)
> # install.packages("survival",dependencies=TRUE) # Only need to do this once
> library(survival) # Do this every time
> # install.packages("asaur",dependencies=TRUE) # Only need to do this once
> library(asaur)
> # help(pharmacoSmoking)
> head(pharmacoSmoking)
   id ttr relapse      grp age gender    race employment yearsSmoking
1 21 182        0 patchOnly 36  Male white      ft       26
2 113 14        1 patchOnly 41  Male white     other      27
3 39  5        1 combination 25 Female white     other      12
4 80 16        1 combination 54  Male white      ft       39
5 87  0        1 combination 45  Male white     other      30
6 29 182        0 combination 43 Male hispanic ft       30
levelSmoking ageGroup2 ageGroup4 priorAttempts longestNoSmoke
1 heavy      21-49  35-49          0            0
2 heavy      21-49  35-49          3            90
3 heavy      21-49  21-34          3            21
4 heavy      50+    50-64          0            0
5 heavy      21-49  35-49          0            0
6 heavy      21-49  35-49          2           1825
> summary(pharmacoSmoking)
   id          ttr      relapse      grp
Min. : 1.00  Min. : 0.00  Min. :0.000  combination:61
1st Qu.:33.00 1st Qu.: 8.00 1st Qu.:0.000  patchOnly :64
Median :67.00 Median :49.00 Median :1.000
Mean  :66.15 Mean  :77.44 Mean  :0.712
3rd Qu.:99.00 3rd Qu.:182.00 3rd Qu.:1.000
Max. :130.00 Max. :182.00 Max. :1.000
   age         gender      race    employment yearsSmoking
Min. :22.00  Female:81  black   :38      ft :72  Min. : 9.00
1st Qu.:41.00 Male   :44  hispanic: 8  other:39 1st Qu.:22.00
Median :49.00           other   : 2  pt  :14  Median :30.00
Mean  :48.84           white   :77           Mean  :30.88
3rd Qu.:56.00
Max. :86.00
levelSmoking ageGroup2 ageGroup4 priorAttempts longestNoSmoke
heavy:89    21-49:66  21-34:16  Min. : 0.00  Min. :  0.0
light:36    50+:59   35-49:50  1st Qu.: 1.00  1st Qu.:  7.0
                  50-64:48  Median : 2.00  Median : 90.0
                  65+:11   Mean  : 12.68  Mean  : 539.7
                           3rd Qu.: 5.00  3rd Qu.: 365.0
                           Max. :1000.00 Max. :6205.0
> # Make fixed-up data frame called quit
> quit = within(pharmacoSmoking,{
+ DayOfRelapse = Surv(ttr+1,relapse)
+ contrasts(grp) = contr.treatment(2,base=2) # Patch only is reference category
+ colnames(contrasts(grp)) = c('Combo') # Names of dummy vars -- just one
+ # Collapse race categories
+ Race = as.character(race) # Small r race is a factor. This is easier to modify.
+ Race[Race!='white'] = 'blackOther'; Race=factor(Race)
+ }) # Finished making data frame quit
> with(quit, table(race,Race) )
   Race
race   blackOther white
  black        38    0
 hispanic       8    0
  other         2    0
  white        0   77

```

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> full = survreg(DayOfRelapse ~ grp + age + gender + Race + employment
+                  + yearsSmoking + levelSmoking + priorAttempts, dist='weibull', data=quit)
> summary(full)

Call:
survreg(formula = DayOfRelapse ~ grp + age + gender + Race +
employment + yearsSmoking + levelSmoking + priorAttempts,
data = quit, dist = "weibull")
            Value Std. Error      z      p
(Intercept) 1.12177  0.97726  1.15 0.2510
grpCombo     1.09225  0.38186  2.86 0.0042
age          0.08432  0.03411  2.47 0.0134
genderMale   0.03631  0.41417  0.09 0.9301
Racewhite    0.25145  0.39143  0.64 0.5206
employmentother -1.28799  0.46719 -2.76 0.0058
employmentpt  -1.28482  0.58631 -2.19 0.0284
yearsSmoking -0.02351  0.03250 -0.72 0.4696
levelSmokinglight -0.07347  0.43151 -0.17 0.8648
priorAttempts -0.00105  0.00200 -0.52 0.6000
Log(scale)    0.54194  0.08917  6.08 0.00000000012

Scale= 1.72

Weibull distribution
Loglik(model)= -463.8  Loglik(intercept only)= -476.5
Chisq= 25.41 on 9 degrees of freedom, p= 0.0025
Number of Newton-Raphson Iterations: 5
n= 125
>

```

I am thinking about dropping Race, yearsSmoking, levelSmoking and priorAttempts. The last 3 variables all represent smoking history and could be correlated highly enough to wash out each other's effects. Test them simultaneously.

```

>
> # Fit the restricted model: Restricted by H0
> rest1 = survreg(DayOfRelapse ~ grp + age + gender + Race + employment ,
+                  dist='weibull', data=quit)
> anova(rest1,full) # LR test

Terms
1 grp + age + gender + Race + employment
2 grp + age + gender + Race + employment + yearsSmoking + levelSmoking +
priorAttempts
  Resid. Df -2*LL Test Df Deviance Pr(>Chi)
1       117 928.3771     NA      NA      NA
2       114 927.5513     =  3 0.8258271 0.8432801

> # Is Race significant with those variables dropped?

```

```

> # Is Race significant with those variables dropped?
> summary(rest1)

Call:
survreg(formula = DayOfRelapse ~ grp + age + gender + Race +
employment, data = quit, dist = "weibull")
      Value Std. Error     z      p
(Intercept) 1.3905    0.8684  1.60  0.1093
grpCombo     1.1021    0.3794  2.91  0.0037
age          0.0637    0.0190  3.35  0.0008
genderMale   0.0561    0.4140  0.14  0.8921
Racewhite    0.1880    0.3788  0.50  0.6196
employmentother -1.2821  0.4635 -2.77  0.0057
employmentpt  -1.2251  0.5837 -2.10  0.0358
Log(scale)    0.5444    0.0894  6.09  0.0000000011

```

Scale= 1.72

Weibull distribution
Loglik(model)= -464.2 Loglik(intercept only)= -476.5
Chisq= 24.58 on 6 degrees of freedom, p= 0.00041
Number of Newton-Raphson Iterations: 5
n= 125

Decision: Drop race and gender.

```

> full2 = survreg(DayOfRelapse ~ grp + age + employment , dist='weibull',
data=quit)
> summary(full2)

```

```

Call:
survreg(formula = DayOfRelapse ~ grp + age + employment, data = quit,
dist = "weibull")
      Value Std. Error     z      p
(Intercept) 1.4957    0.8414  1.78  0.07545
grpCombo     1.1023    0.3793  2.91  0.00366
age          0.0643    0.0186  3.45  0.00055
employmentother -1.2880  0.4617 -2.79  0.00528
employmentpt  -1.2123  0.5616 -2.16  0.03088
Log(scale)    0.5454    0.0894  6.10  0.000000001

```

Scale= 1.73

Weibull distribution
Loglik(model)= -464.3 Loglik(intercept only)= -476.5
Chisq= 24.31 on 4 degrees of freedom, p= 0.000069
Number of Newton-Raphson Iterations: 5
n= 125

```

> # Test employment status controlling for age and experimental treatment.
> rest2 = survreg(DayOfRelapse ~ grp + age , dist='weibull', data=quit)
> anova(rest2,full2) # LR test
      Terms Resid. Df     -2*LL Test Df Deviance    Pr(>Chi)
1           grp + age     121 937.9007     NA     NA      NA
2 grp + age + employment     119 928.6554     =  2  9.245333 0.009826558

```

```

> # Test employment status with a Wald test.
> source("http://www.utstat.toronto.edu/~brunner/Rfunctions/Wtest.txt")
> # function(L,Tn,Vn,h=0) # H0: L theta = h
> # Tn is estimated theta, usually a vector.
> # Vn is the estimated asymptotic covariance matrix of Tn.
> # For Wald tests based on numerical MLEs, Tn = theta-hat,
> # and Vn is the inverse of the Hessian of the minus log likelihood.
>
> Vhat = vcov(full2); Vhat
            (Intercept)      grpCombo          age employmentother
(Intercept)    0.7079360800 -0.0320256900 -0.0147694486  0.111673731
grpCombo       -0.0320256900  0.1438698739 -0.0004703383 -0.013521493
age           -0.0147694486 -0.0004703383  0.0003472409 -0.003893727
employmentother 0.1116737308 -0.0135214927 -0.0038937268  0.213191081
employmentpt   -0.0003554818 -0.0078279548 -0.0013434899  0.077138486
Log(scale)     -0.0098224903  0.0050290739  0.0002048412 -0.003182291
                  employmentpt  Log(scale)
(Intercept)    -0.0003554818 -0.0098224903
grpCombo       -0.0078279548  0.0050290739
age           -0.0013434899  0.0002048412
employmentother 0.0771384860 -0.0031822913
employmentpt   0.3153999894 -0.0035442716
Log(scale)     -0.0035442716  0.0079888732

> thetahat = full2$coefficients; thetahat
            (Intercept)      grpCombo          age employmentother
(Intercept)    1.4957374    1.1023048    0.0643414   -1.2880472
employmentpt  -1.2122529
>
```

Note that the asymptotic covariance matrix includes $\log(\sigma)$, but the "coefficients" vector does not.

```

> sigmahat = full2$scale; sigmahat
[1] 1.725305
> thetahat = c(thetahat,log(sigmahat))
>
> # H0: beta3=beta4=0. Express as H0: L theta = h
> eMat = rbind( c(0,0,0,1,0,0),
+                 c(0,0,0,0,1,0) )

> Wtest(L=eMat, Tn=thetahat, Vn=Vhat)
      W      df      p-value
9.718885315 2.000000000 0.007754805
> anova(rest2,full2) # Repeating LR test for comparison
             Terms Resid. Df -2*LL Test Df Deviance Pr(>Chi)
1 grp + age           121 937.9007    NA     NA      NA
2 grp + age + employment  119 928.6554    =  2  9.245333 0.009826558

>
> # Test part time versus other
> pto = cbind(0,0,0,1,-1,0); pto
     [,1] [,2] [,3] [,4] [,5] [,6]
[1,]    0    0    0    1   -1    0
> Wtest(L=pto, Tn=thetahat, Vn=Vhat)
      W      df      p-value
0.01534747 1.000000000 0.90140640
>
```

Predict the day of relapse for a 50 year old patient who is employed full time and gets the patch-only treatment.

Weibull Regression: $t_i = \exp\{\beta_0 + \beta_1 x_{i,1} + \dots + \beta_{p-1} x_{i,p-1}\} \cdot \epsilon_i^\sigma = e^{\mathbf{x}_i^\top \boldsymbol{\beta}} \epsilon_i^\sigma$, where $\epsilon_i \sim \exp(1)$.

- $t_i \sim \text{Weibull}$, with $\alpha = 1/\sigma$ and $\lambda = e^{-\mathbf{x}_i^\top \boldsymbol{\beta}}$.
- $E(t_i) = e^{\mathbf{x}_i^\top \boldsymbol{\beta}} \Gamma(\sigma+1)$, Median(t_i) = $e^{\mathbf{x}_i^\top \boldsymbol{\beta}} \log(2)^\sigma$, $h_i(t) = \frac{1}{\sigma} \exp\{-\frac{1}{\sigma} \mathbf{x}_i^\top \boldsymbol{\beta}\} t^{\frac{1}{\sigma}-1}$.
- $S(t) = \exp\left\{-e^{-\frac{1}{\sigma} \mathbf{x}_i^\top \boldsymbol{\beta}} t^{\frac{1}{\sigma}}\right\}$

```

> theta_hat
   (Intercept)           grpCombo            age employment_other
   1.4957374          1.1023048         0.0643414        -1.2880472
employment_pt
   -1.2122529          0.5454037

> x = c(1,0,50,0,0,0)
> xb = sum(x*theta_hat)
>
> # a) The estimated mean
> exp(xb) * gamma(sigmahat+1)
[1] 175.5516
>
> # b) The estimated mean
> exp(xb) * log(2)^sigmahat
[1] 59.17273
>
> # I think the median is preferable to mean because the Weibull distribution
> # is skewed. Also, the predict function for Weibull regression works as expected
> # for medians (but not means).
>
> oldguy = data.frame(grp='patchOnly',age=50,employment='ft')
> predict(full2,newdata=oldguy,type='quantile',p=0.5,se=TRUE)
$fit
      1
59.17273

$se.fit
      1
18.87577

> # The 0.5 quantile is the median. se is from the delta method.
>
> # Estimate and plot S(t) for the old guy.

```

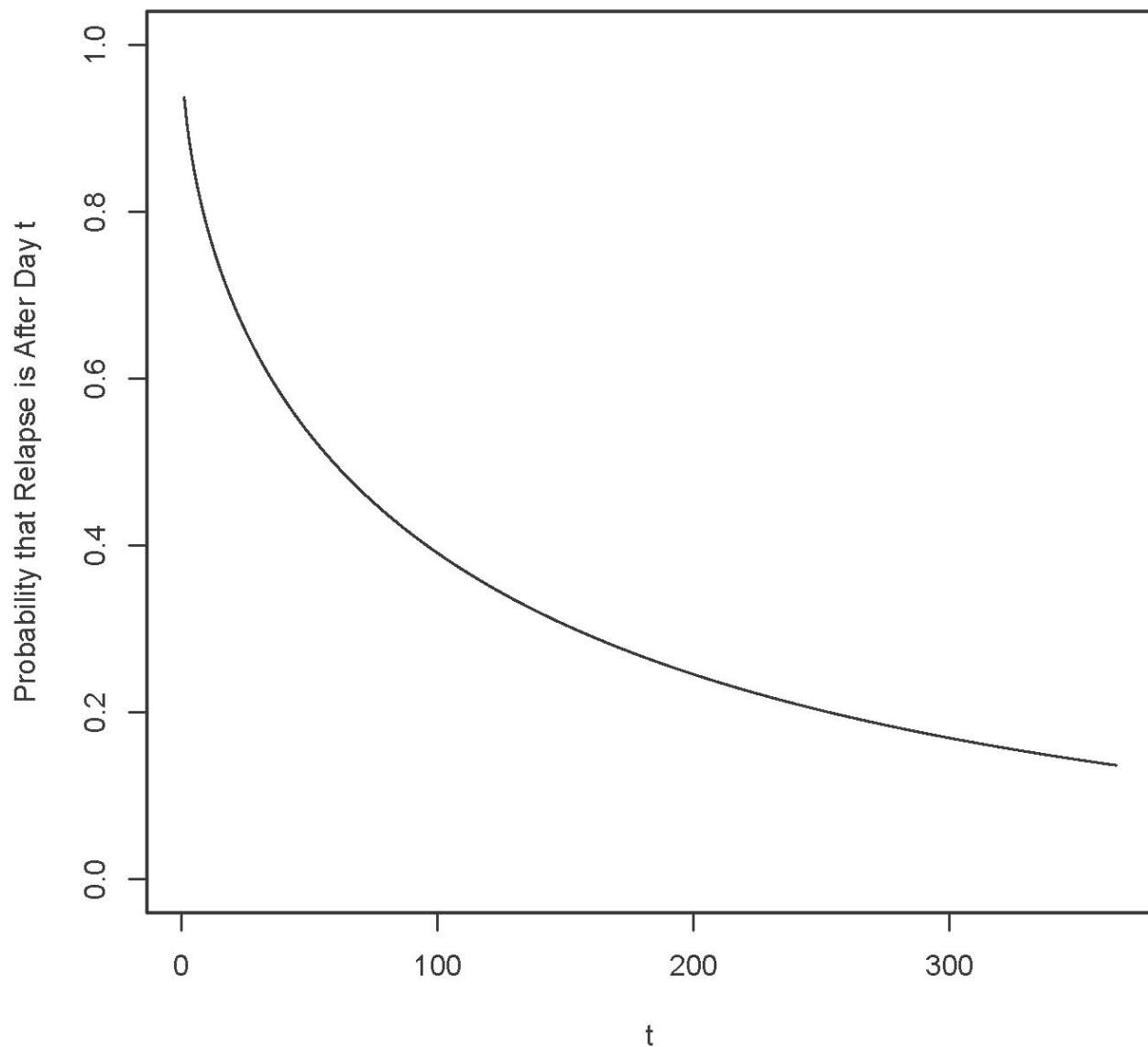
The se of $S\text{-hat}(t)$ is straightforward in theory, but messy in practice. G-dot from Mathematica is very ugly. For example, in Wolfram Alpha, try $D[\exp(-t^{1/s}) \exp(-(b0 + 50 b2)/s), b0]$

```

> t = 1:365
> Shat = exp( -(exp(-xb/sigmahat)*t^(1/sigmahat)) )
>
> plot(t,Shat,type='l',ylim=c(0,1),xlab='t',ylab='Probability that Relapse is After
Day t')
> tstring = expression(paste(hat(S)(t), " = Probability Relapsing After Day t"))
> title(tstring)

```

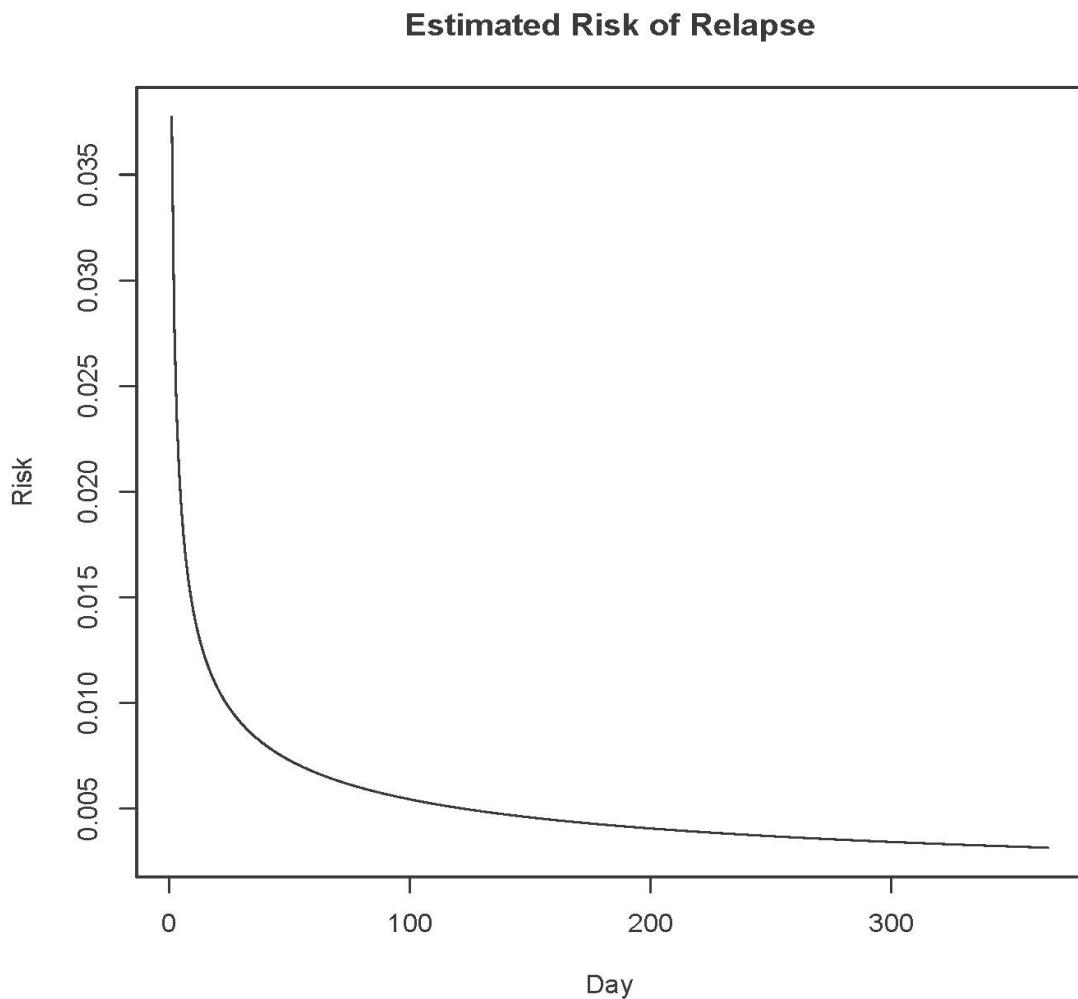
$\hat{S}(t)$ = Probability Relapsing After Day t



Plot estimated hazard function for that 50 year old patient who is employed full time and gets the patch-only treatment.

$$\begin{aligned}
 h(t) &= \frac{f(t)}{S(t)} \\
 &= \frac{\alpha\lambda(\lambda t)^{\alpha-1} e^{-(\lambda t)^\alpha}}{e^{-(\lambda t)^\alpha}} \\
 &= \alpha\lambda^\alpha t^{\alpha-1} \\
 &= \frac{1}{\sigma} e^{-\frac{1}{\sigma} \mathbf{x}^\top \boldsymbol{\beta}} t^{\frac{1}{\sigma}-1}
 \end{aligned}$$

```
> h = 1/sigmahat * exp(-xb/sigmahat) * t^(1/sigmahat - 1)
> plot(t,h,type='l',xlab='Day',ylab='Risk',main='Estimated Risk of Relapse')
```



LaTeX code for the record

```
\noindent
Weibull Regression: $t_i = \exp\{\beta_0 + \beta_{1x_{i,1}} + \dots + \beta_{p-1}x_{i,p-1}\} \cdot \epsilon_i^\sigma = e^{\mathbf{x}_i^\top \boldsymbol{\beta}} \epsilon_i^\sigma,
where \epsilon_1 \sim \exp(1).
\begin{itemize}
    \item $t_i \sim$ Weibull, with $\alpha = 1/\sigma$ and $\lambda = e^{-(\mathbf{x}_i^\top \boldsymbol{\beta})}$.
    \item $E(t_i) = e^{\mathbf{x}_i^\top \boldsymbol{\beta}} \Gamma(\sigma + 1)$,
        Median($t_i$) = $e^{\mathbf{x}_i^\top \boldsymbol{\beta}} \log(2)^\sigma$,
        $h_i(t) = \frac{1}{\sigma} e^{-\frac{1}{\sigma} t^{\frac{1}{\sigma}-1}}$.
    \item $S(t) = e^{-e^{-\frac{1}{\sigma} t^{\frac{1}{\sigma}}}}$ 
\end{itemize}
% Hazard calculation
\begin{eqnarray*}
h(t) &= \frac{f(t)}{S(t)} \\
&= \frac{\alpha \lambda}{(\lambda t)^{\alpha-1}} e^{-(\lambda t)^\alpha} \\
&= \frac{1}{\sigma} e^{-\frac{1}{\sigma} t^{\frac{1}{\sigma}}} \\
&= \frac{1}{\sigma} \mathbf{x}_i^\top \boldsymbol{\beta} - 1
\end{eqnarray*}
```

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<http://www.utstat.toronto.edu/~brunner/oldclass/312f23>