

# Weibull Regression with R, Part One\*

## Comparing Two Treatments

### The Pharmaco-smoking study

The purpose of this study ... was to evaluate extended duration of a triple-medication combination versus therapy with the nicotine patch alone in smokers with medical illnesses. Patients with a history of smoking were randomly assigned to the triple-combination or patch therapy and followed for up to six months. The primary outcome variable was time from randomization until relapse (return to smoking); individuals who remained non-smokers for six months were censored. The data set, “pharmacoSmoking”, is available in the “asaur” package.

The variable “ttr” is the number of days without smoking (“time to relapse”), and “relapse=1” indicates that the subject started smoking again at the given time. The variable “grp” is the treatment indicator, and “employment” can take the values “ft” (full time), “pt” (part time), or “other”. This material is quoted from *Applied Survival Analysis Using R*, pages 18-19.

```
> 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)
```

---

\* Copyright information is on the last page.

```
> 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	age
Min. : 1.00	Min. : 0.00	Min. : 0.000	combination:61	Min. : 22.00
1st Qu.: 33.00	1st Qu.: 8.00	1st Qu.: 0.000	patchOnly :64	1st Qu.: 41.00
Median : 67.00	Median : 49.00	Median : 1.000		Median : 49.00
Mean : 66.15	Mean : 77.44	Mean : 0.712		Mean : 48.84
3rd Qu.: 99.00	3rd Qu.: 182.00	3rd Qu.: 1.000		3rd Qu.: 56.00
Max. : 130.00	Max. : 182.00	Max. : 1.000		Max. : 86.00

gender	race	employment	yearsSmoking	levelSmoking	ageGroup2
Female:81	black :38	ft :72	Min. : 9.00	heavy:89	21-49:66
Male :44	hispanic: 8	other:39	1st Qu.:22.00	light:36	50+ :59
	other : 2	pt :14	Median :30.00		
	white :77		Mean :30.88		
			3rd Qu.:39.00		
			Max. :56.00		

ageGroup4	priorAttempts	longestNoSmoke
21-34:16	Min. : 0.00	Min. : 0.0
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

```
> quit = within(pharmacoSmoking, {TimeToRelapse = Surv(ttr,relapse)} )
```

```
> sort(quit$TimeToRelapse)
```

[1]	0	0	0	0	0	0	0	0	0	0	0	0	1	1	1
[16]	1	1	2	2	2	2	2	2	3	4	4	4	5	5	6
[31]	7	8	8	8	10	12	12	14	14	14	14	14	14	14	15
[46]	15	15	15	16	20	21	21	25	28	28	28	30	30	30	40
[61]	42	45	49	50	56	56	56	56	56	60	60	63	63	65	75
[76]	77	77	80	84	100	105	110	140	140	140	140	155	170	170	182+
[91]	182+	182+	182+	182+	182+	182+	182+	182+	182+	182+	182+	182+	182+	182+	182+
[106]	182+	182+	182+	182+	182+	182+	182+	182+	182+	182+	182+	182+	182+	182+	182+
[121]	182+	182+	182+	182+	182+										

```

> # Fit a Weibull model with just treatment group, which was randomly assigned.
> Model0 = survreg(TimeToRelapse~grp,dist='weibull', data=quit)
Error in survreg(TimeToRelapse ~ grp, dist = "weibull", data = quit) :
  Invalid survival times for this distribution
> # Likely choking on the zeros.
> # Day of relapse starts with one.
> quit = within(quit, {DayOfRelapse = Surv(ttr+1,relapse)} )
> Model0 = survreg(DayOfRelapse~grp,dist='weibull', data=quit)
> summary(Model0)

```

Call:

```

survreg(formula = DayOfRelapse ~ grp, dist = "weibull")
              Value Std. Error      z      p
(Intercept)  5.260      0.3053 17.23 1.60e-66
grppatchOnly -1.162      0.3999 -2.91 3.67e-03
Log(scale)   0.606      0.0904  6.70 2.14e-11

```

Scale= 1.83

Weibull distribution

```

Loglik(model)= -472.1  Loglik(intercept only)= -476.5
      Chisq= 8.78 on 1 degrees of freedom, p= 0.003
Number of Newton-Raphson Iterations: 5
n= 125

```

Conclusion is that combination therapy is more effective. But the alphabetical order of treatments makes combination the reference category, and this is clumsy. Make patch-only the reference category and re-run. This is what we have.

<b>Treatment</b>	<b>d</b>	$E(Y X=x) = \beta_0 + \beta_1 d$
Combination	0	$\beta_0$
Patch Only	1	$\beta_0 + \beta_1$

This is what we want.

<b>Treatment</b>	<b>d</b>	$E(Y X=x) = \beta_0 + \beta_1 d$
Combination	1	$\beta_0 + \beta_1$
Patch Only	0	$\beta_0$

```

> quit = within(quit,{
+ contrasts(grp) = contr.treatment(2,base=2)
+ colnames(contrasts(grp)) = c('Combo') # Names of dummy vars -- just one
+ })
> Model1 = survreg(DayOfRelapse~grp,dist='weibull', data=quit)
> summary(Model1)

```

Call:

```
survreg(formula = DayOfRelapse ~ grp, dist = "weibull")
```

	Value	Std. Error	z	p
(Intercept)	4.098	0.2548	16.09	3.22e-58
grpCombo	1.162	0.3999	2.91	3.67e-03
Log(scale)	0.606	0.0904	6.70	2.14e-11

Scale= 1.83

Weibull distribution

Loglik(model)= -472.1 Loglik(intercept only)= -476.5

Chisq= 8.78 on 1 degrees of freedom, p= 0.003

Number of Newton-Raphson Iterations: 5

n= 125

```
> betahat = Model1$coefficients; betahat
```

(Intercept)	grpCombo
4.097999	1.161901

```
> betahat0 = betahat[1]; betahat1 = betahat[2]
```

```
> sigmahat = Model1$scale; sigmahat
```

```
[1] 1.832392
```

```
> Vhat = vcov(Model1); Vhat
```

	(Intercept)	grpCombo	Log(scale)
(Intercept)	0.064903153	-0.065806535	-0.001649823
grpCombo	-0.065806535	0.159909904	0.006128624
Log(scale)	-0.001649823	0.006128624	0.008179519

```
> # Asymptotic covariance matrix comes out in terms of Log(scale), which is
```

```
> # a minor pain.
```

```
>
```

```
> # 1) When patients receive the combination drug therapy rather than nicotine
patch only, expected relapse time is multiplied by _____ .
```

```
> # a) Give an estimate
```

```
> # b) Modify the CI for beta1 to get a 95% confidence interval (don't use the
delta method).
```

```
>
```

```
> # a) Give an estimate
```

```
> exp(betahat1)
```

```
grpCombo
```

```
3.196004
```

```

> # a) Give an estimate
> exp(betahat1)
grpCombo
3.196004
>
> # b) Modify the CI for beta1 to get a 95% confidence interval (don't use the
delta method).
> L = 1.162 - 1.96*0.3999; U = 1.162 + 1.96*0.3999
> c(exp(L),exp(U))
[1] 1.459649 6.999257

> summary(Model1) # Repeating

```

Call:

```
survreg(formula = DayOfRelapse ~ grp, dist = "weibull")
```

	Value	Std. Error	z	p
(Intercept)	4.098	0.2548	16.09	3.22e-58
grpCombo	1.162	0.3999	2.91	3.67e-03
Log(scale)	0.606	0.0904	6.70	2.14e-11

Scale= 1.83

Weibull distribution

Loglik(model)= -472.1    Loglik(intercept only)= -476.5

    Chisq= 8.78 on 1 degrees of freedom, p= 0.003

Number of Newton-Raphson Iterations: 5

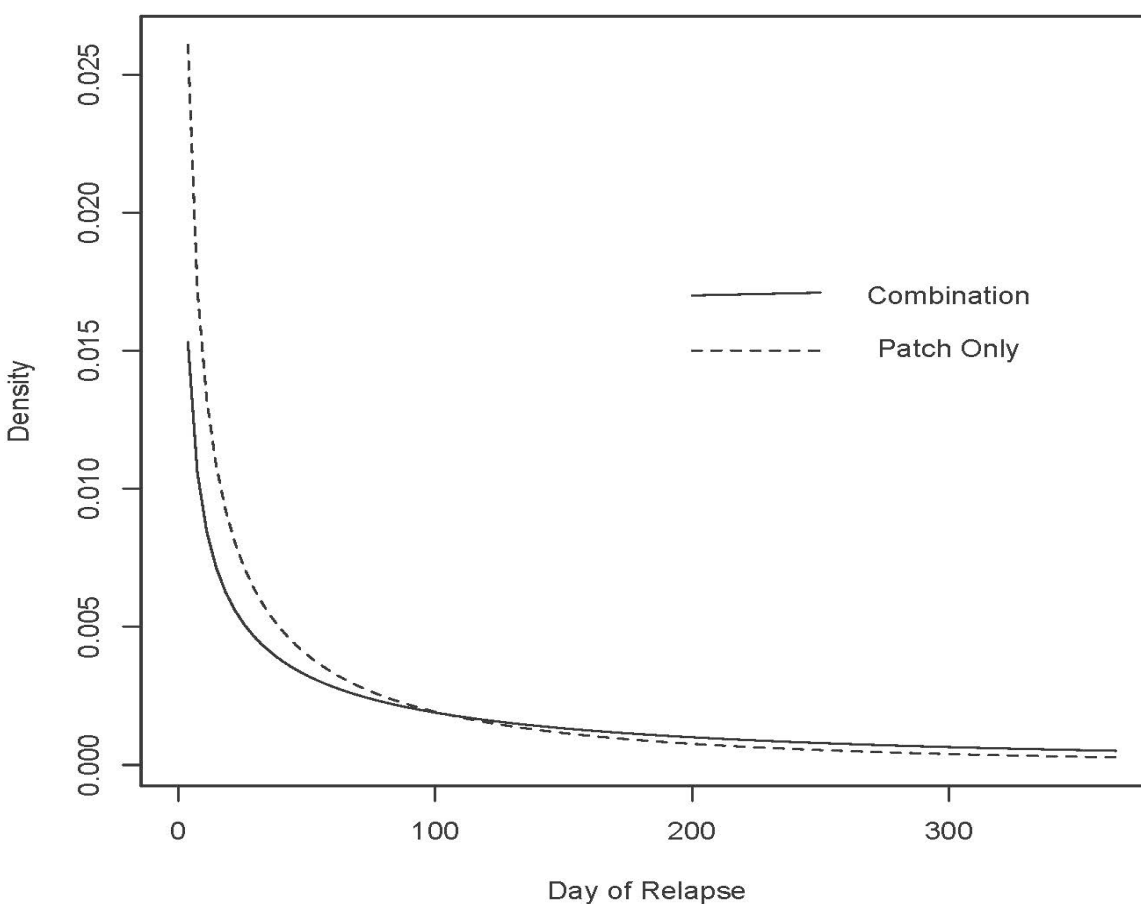
n= 125

```

> # 2) Estimate and plot the density of relapse time for the two experimental
conditions.
>
> # Okay, lambda = exp(-mu), alpha = 1/sigma
> alpha = 1/sigmahat
> lambda0 = exp(-betahat0); lambda1 = exp(-betahat0-betahat1)
> t = seq(from=0,to=300,length=101)
> f0 = alpha*lambda0^alpha * t^(alpha-1) * exp(-(lambda0*t)^alpha)
> f1 = alpha*lambda1^alpha * t^(alpha-1) * exp(-(lambda1*t)^alpha)
> plot(t,f0,pch=' ',xlab='Day of Relapse',ylab='Density') # Empty plot
> title('Estimated Density of Relapse Time')
> lines(t,f0,lty=2); lines(t,f1,lty=1)
> # Annotate the plot
> x0=c(200,250); y0 = c(0.015,0.015); lines(x0,y0,lty=2)
> text(300,0.015,'Patch Only')
> x1=c(200,250); y1 = c(0.017,0.0171); lines(x1,y1,lty=1)
> text(300,0.017,'Combination')

```

**Estimated Density of Relapse Time**



```

> max(quit$ttr) # Maximum time value (censored)
[1] 182

```

> # 3) Estimate median time to relapse for the 2 groups, with Cis

$$E(t_i^*) = \frac{\Gamma(1 + \frac{1}{\alpha})}{\lambda} = e^{\mathbf{x}_i^\top \boldsymbol{\beta}} \Gamma(1 + \sigma)$$

$$\text{Median}(t_i^*) = \frac{[\log(2)]^{1/\alpha}}{\lambda} = e^{\mathbf{x}_i^\top \boldsymbol{\beta}} \log(2)^\sigma$$

```

> # Asymptotic covariance matrix comes out in terms of Log(scale), which is
> # unfortunate.
> # Denote log(sigma) by s, and re-write g(theta) = exp(beta0) * log(2)^sigma
> # as g(theta) = exp(beta0) * log(2)^exp(s)
> shat = log(sigmahat)
>
> # Patch Only
> medianhat0 = exp(betahat0)*log(2)^sigmahat
> # gdot will be a 1 x 3 matrix.
> gdot0 = cbind( exp(betahat0)*log(2)^exp(shat), 0, exp(betahat0) *
log(2)^exp(shat) * log(log(2)) * exp(shat) )
> se0 = sqrt( as.numeric(gdot0 %*% Vhat %*% t(gdot0)) ); se0
[1] 8.18674
> lower0 = medianhat0 - 1.96*se0; upper0 = medianhat0 + 1.96*se0
> patchonly = c(medianhat0,lower0,upper0)
> names(patchonly) = c('Median','Lower95','Upper95')
> patchonly
  Median Lower95 Upper95
30.76581 14.71980 46.81182
>
> # Combination drug treatment
> medianhat1 = exp(betahat0+betahat1)*log(2)^sigmahat
> gdot1 = cbind( exp(betahat0+betahat1)*log(2)^exp(shat),
exp(betahat0+betahat1)*log(2)^exp(shat),
+               exp(betahat0+betahat1) * log(2)^exp(shat) * log(log(2)) *
exp(shat) )
> se1 = sqrt( as.numeric(gdot1 %*% Vhat %*% t(gdot1)) ); se1
[1] 29.64109
> lower1 = medianhat1 - 1.96*se1; upper1 = medianhat1 + 1.96*se1
> combination = c(medianhat1,lower1,upper1)
> names(combination) = c('Median','Lower95','Upper95')
> combination

```

```

    Median   Lower95   Upper95
98.32767  40.23114 156.42420
>
> # There is an easier way to get these numbers
> # help(predict.survreg)
> Justpatch = data.frame(grp='patchOnly')
> # A data frame with just one case and one variable.
> Combination = data.frame(grp='combination')
> treatments = rbind(Justpatch,Combination); treatments
      grp
1  patchOnly
2  combination

> # The 0.5 quantile is the median
> medians = predict(Model1,newdata=treatments,type='quantile',p=0.5,se=TRUE)
> medians
$fit
      1      2
30.76581 98.32767

$se.fit
      1      2
8.18674 29.64109

> cbind(medians$fit,medians$se)
      [,1] [,2]
1 30.76581 8.18674
2 98.32767 29.64109
> rbind(c(medianhat0,se0),
+       c(medianhat1,se1) )
      (Intercept)
[1,] 30.76581 8.18674
[2,] 98.32767 29.64109
>

```



```

> # 4) Plot the Kaplan-Meier estimates and MLEs of S(t)
>
> KM = survfit(DayOfRelapse~grp, type="kaplan-meier", data=quit)
# Kaplan-Meier is the default estimation method anyway.
> summary(KM)
Call: survfit(formula = DayOfRelapse ~ grp, type = "kaplan-meier")

```

grp=combination							
time	n.risk	n.event	survival	std.err	lower 95% CI	upper 95% CI	
1	61	4	0.934	0.0317	0.874	0.999	
3	57	3	0.885	0.0408	0.809	0.969	
5	54	1	0.869	0.0432	0.788	0.958	
6	53	2	0.836	0.0474	0.748	0.934	
9	51	2	0.803	0.0509	0.709	0.909	
11	49	1	0.787	0.0524	0.691	0.897	
13	48	1	0.770	0.0538	0.672	0.884	
15	47	1	0.754	0.0551	0.653	0.870	
16	46	2	0.721	0.0574	0.617	0.843	
17	44	1	0.705	0.0584	0.599	0.829	
21	43	1	0.689	0.0593	0.582	0.815	
22	42	1	0.672	0.0601	0.564	0.801	
31	41	2	0.639	0.0615	0.530	0.772	
43	39	1	0.623	0.0621	0.512	0.757	
51	38	1	0.607	0.0625	0.496	0.742	
57	37	2	0.574	0.0633	0.462	0.712	
61	35	2	0.541	0.0638	0.429	0.682	
64	33	2	0.508	0.0640	0.397	0.650	
66	31	1	0.492	0.0640	0.381	0.635	
76	30	1	0.475	0.0639	0.365	0.619	
111	29	1	0.459	0.0638	0.350	0.603	
141	28	3	0.410	0.0630	0.303	0.554	
171	25	1	0.393	0.0625	0.288	0.537	

grp=patchOnly							
time	n.risk	n.event	survival	std.err	lower 95% CI	upper 95% CI	
1	64	8	0.875	0.0413	0.798	0.960	
2	56	5	0.797	0.0503	0.704	0.902	
3	51	3	0.750	0.0541	0.651	0.864	
4	48	1	0.734	0.0552	0.634	0.851	
5	47	2	0.703	0.0571	0.600	0.824	
7	45	1	0.688	0.0579	0.583	0.811	
8	44	1	0.672	0.0587	0.566	0.797	
9	43	1	0.656	0.0594	0.550	0.784	
13	42	1	0.641	0.0600	0.533	0.770	
15	41	6	0.547	0.0622	0.438	0.684	
16	35	2	0.516	0.0625	0.407	0.654	
22	33	1	0.500	0.0625	0.391	0.639	
26	32	1	0.484	0.0625	0.376	0.624	

29	31	3	0.437	0.0620	0.331	0.578
31	28	1	0.422	0.0617	0.317	0.562
41	27	1	0.406	0.0614	0.302	0.546
46	26	1	0.391	0.0610	0.288	0.530
50	25	1	0.375	0.0605	0.273	0.515
57	24	3	0.328	0.0587	0.231	0.466
78	21	2	0.297	0.0571	0.204	0.433
81	19	1	0.281	0.0562	0.190	0.416
85	18	1	0.266	0.0552	0.177	0.399
101	17	1	0.250	0.0541	0.164	0.382
106	16	1	0.234	0.0530	0.151	0.365
141	15	1	0.219	0.0517	0.138	0.348
156	14	1	0.203	0.0503	0.125	0.330
171	13	1	0.187	0.0488	0.113	0.312

> # Look at K-M estimates of medians

> KM[1] # Combination

Call: survfit(formula = DayOfRelapse ~ grp, type = "kaplan-meier")

n	events	median	0.95LCL	0.95UCL
61	37	66	51	NA

> KM[2] # Patch Only

Call: survfit(formula = DayOfRelapse ~ grp, type = "kaplan-meier")

n	events	median	0.95LCL	0.95UCL
64	52	24	15	57

> # Repeat MLEs for comparison

> combination

Median	Lower95	Upper95
98.32767	40.23114	156.42420

> patchonly

Median	Lower95	Upper95
30.76581	14.71980	46.81182

```

> plot(KM, xlab='t', ylab='Survival Probability', lwd=2, col=1:2)
> # 1 is black, 2 is red
> legend(x=125,y=1.0, col=1:2, lwd=2, legend=c('Combination','Patch Only'))
> title(expression(paste(hat(S)(t),': Kaplan-Meier and Maximum Likelihood
Estimates')))
> # MLEs

```

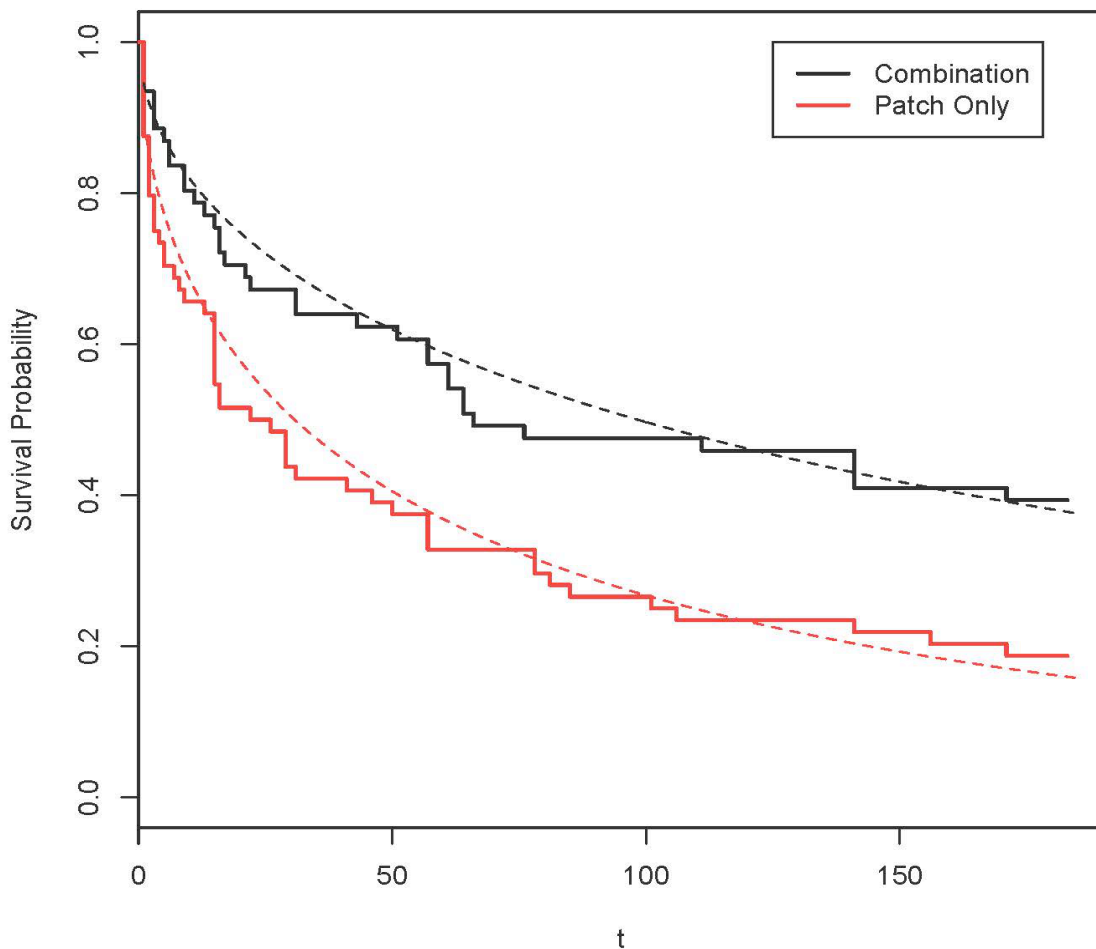
$$S(t) = e^{-(\lambda t)^\alpha} \quad \lambda = e^{-\mathbf{x}_i^\top \boldsymbol{\beta}} \quad \alpha = 1/\sigma$$

```

> x = 1:185
> lambda0 = exp(-betahat0); lambda1 = exp(-betahat0-betahat1); alpha=1/sigmahat
> Shat0 = exp(-(lambda0*x)^alpha); Shat1 = exp(-(lambda1*x)^alpha)
> lines(x,Shat0,lty=2,col=2) # Patch only is red
> lines(x,Shat1,lty=2)      # Combination is black (default)

```

$\hat{S}(t)$ : Kaplan-Meier and Maximum Likelihood Estimates



```
>
> # Non-parametric rank test of equal survival functions
> # See http://dwo11.de/rexrepos/posts/survivalKM.html
> survdiff(DayOfRelapse~grpgrp, data=quit)
```

Call:

```
survdiff(formula = DayOfRelapse ~ grp)
```

	N	Observed	Expected	(O-E)^2/E	(O-E)^2/V
grp=combination	61	37	49.9	3.36	8.03
grp=patchOnly	64	52	39.1	4.29	8.03

Chisq= 8 on 1 degrees of freedom, p= 0.00461

```
> # Compare p = 0.00367 from Z-test of H0: beta1=0
```

```
>
```

	relapse	
grp	0	1
combination	24	37
patchOnly	12	52

```
> prop.table(twoby2,1) # Proportions of row totals
```

	relapse	
grp	0	1
combination	0.3934426	0.6065574
patchOnly	0.1875000	0.8125000

```
> chisq.test(twoby2)
```

Pearson's Chi-squared test with Yates' continuity correction

data: twoby2

X-squared = 5.4945, df = 1, p-value = 0.01908

```
> fisher.test(twoby2) # p = 0.01713
```

---

This document was prepared by [Jerry Brunner](#), University of Toronto. It is licensed under a Creative Commons Attribution - ShareAlike 3.0 Unported License:

[http://creativecommons.org/licenses/by-sa/3.0/deed.en\\_us](http://creativecommons.org/licenses/by-sa/3.0/deed.en_us). Use any part of it as you like and share the result freely. It is available in OpenOffice.org format from the course website:

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