## Introduction to rstpm2

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## 1 Cure models

For cure, we use the melanoma dataset used by Andersson and colleagues for cure models with Stata's stpm2 (see http://www.pauldickman.com/survival/). Initially, we merge the patient data with the all cause mortality rates.

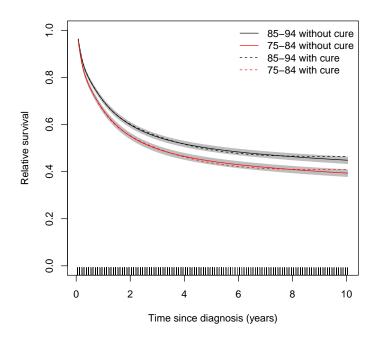
```
> require(rstpm2)
> data(popmort)
> data(colon)
> popmort2 <- transform(popmort,exitage=age,exityear=year,age=NULL,year=NULL)
> colon2 <- within(colon, {</pre>
    status <- ifelse(surv_mm>120.5,1,status)
    tm <- pmin(surv_mm,120.5)/12</pre>
    exit <- dx+tm*365.25
    sex <- as.numeric(sex)</pre>
    exitage <- pmin(floor(age+tm),99)</pre>
    exityear <- floor(yydx+tm)</pre>
    ##year8594 <- (year8594=="Diagnosed 85-94")
+ })
> colon2 <- merge(colon2,popmort2)</pre>
For comparisons, we fit the relative survival model without and with cure.
> fit0 <- stpm2(Surv(tm,status %in% 2:3)~I(year8594=="Diagnosed 85-94"),</pre>
                         data=colon2,
                         bhazard=colon2$rate, df=5)
> summary(fit <- stpm2(Surv(tm,status %in% 2:3)~I(year8594=="Diagnosed 85-94"),</pre>
                         data=colon2,
                         bhazard=colon2$rate,
                         df=5,cure=TRUE))
Maximum likelihood estimation
Call:
mle2(minuslog1 = negl1, start = init, vecpar = TRUE, gr = function (beta)
    eta <- as.vector(X %*% beta)
    etaD <- as.vector(XD %*% beta)
    h <- etaD * exp(eta) + bhazard
```

```
h[h < 0] < -1e-100
    g <- colSums(exp(eta) * wt * (-X + ifelse(event, 1/h, 0) *
        (XD + X * etaD)))
    return(-g)
\}, control = list(parscale = c(0.1, 0.1, 0.1, 0.1, 0.1, 0.1,
0.1), maxit = 300), lower = -Inf, upper = Inf)
Coefficients:
                                      Estimate Std. Error z value
                                                                        Pr(z)
                                                 0.055234 -72.0090 < 2.2e-16
(Intercept)
                                     -3.977326
I(year8594 == "Diagnosed 85-94")TRUE -0.155613
                                                 0.025090
                                                           -6.2021 5.573e-10
nsx(log(tm), df = 5, cure = TRUE)1
                                      3.323188
                                                 0.053960 61.5858 < 2.2e-16
nsx(log(tm), df = 5, cure = TRUE)2
                                      3.628637
                                                 0.053874 67.3537 < 2.2e-16
nsx(log(tm), df = 5, cure = TRUE)3
                                      1.634854
                                                 0.022294 73.3323 < 2.2e-16
nsx(log(tm), df = 5, cure = TRUE)4
                                      6.592018
                                                 0.112810 58.4346 < 2.2e-16
nsx(log(tm), df = 5, cure = TRUE)5
                                      3.371806
                                                 0.043119 78.1974 < 2.2e-16
(Intercept)
I(year8594 == "Diagnosed 85-94")TRUE ***
nsx(log(tm), df = 5, cure = TRUE)1
                                     ***
nsx(log(tm), df = 5, cure = TRUE)2
                                     ***
nsx(log(tm), df = 5, cure = TRUE)3
                                     ***
nsx(log(tm), df = 5, cure = TRUE)4
nsx(log(tm), df = 5, cure = TRUE)5
                                     ***
Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
-2 log L: 42190.77
> predict(fit,head(colon2),se.fit=TRUE)
   Estimate
                lower
                          upper
1 0.8610835 0.8543192 0.8675585
2 0.7934666 0.7850400 0.8016060
3 0.6967422 0.6863396 0.7068773
4 0.8610835 0.8543192 0.8675585
5 0.8221253 0.8143552 0.8296052
6 0.8610835 0.8543192 0.8675585
```

The estimate for the year parameter from the model without cure is within three significant figures with that in Stata. For the predictions, the Stata model gives:

	Δ.				4
		surv	surv_lci	surv_uci	
	1				•
1.	-	.86108264	.8542898	.8675839	
2.	-	.79346526	.7850106	.8016309	I
3.		.69674037	.6863196	.7068927	١
4.	-	.86108264	.8542898	.8675839	١
5.	-	.82212425	.8143227	.8296332	١
	-				١

```
6. | .86108264 .8542898 .8675839 |
     +----+
We can estimate the proportion of failures at the end of follow-up using:
> newdata.eof <- data.frame(year8594 = unique(colon2$year8594),</pre>
                            tm=max(colon2$tm))
> 1-predict(fit0, newdata.eof, type="surv", se.fit=TRUE)
   Estimate
                lower
                          upper
1 0.6063672 0.6211834 0.5915914
2 0.5515138 0.5661462 0.5369984
> 1-predict(fit, newdata.eof, type="surv", se.fit=TRUE)
   Estimate
                lower
                          upper
1 0.5913307 0.6055044 0.5772144
2 0.5350813 0.5485375 0.5217431
We can plot the predicted survival estimates:
> plot(fit0,newdata=data.frame(year8594 = "Diagnosed 85-94"), ylim=0:1,
       xlab="Time since diagnosis (years)", ylab="Relative survival")
> plot(fit0, newdata=data.frame(year8594 = "Diagnosed 75-84"),
       add=TRUE,line.col="red",rug=FALSE)
> plot(fit,newdata=data.frame(year8594 = "Diagnosed 85-94"),
       add=TRUE, ci=FALSE, lty=2, rug=FALSE)
> plot(fit,newdata=data.frame(year8594="Diagnosed 75-84"),
       add=TRUE,rug=FALSE,line.col="red",ci=FALSE,lty=2)
> legend("topright",c("85-94 without cure","75-84 without cure",
                      "85-94 with cure", "75-84 with cure"),
         col=c(1,2,1,2), lty=c(1,1,2,2), bty="n")
```



## And the hazard curves:

