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>
+ })
> 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$rate, df=5)
> summary(fit <- stpm2(Surv(tm, status %in% 2:3)~I(year8594=="Diagnosed 85-94"),
                        data=colon2,
+
                        bhazard=colon2$rate,
                        df=5,cure=TRUE))
Maximum likelihood estimation
Call:
mle2(minuslogl = negll, start = init, vecpar = TRUE, 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

```
(Intercept)
                                                           0.054778 -72.6078
                                               -3.977326
I(year8594 == "Diagnosed 85-94")TRUE
                                                           0.025102 -6.1992
                                               -0.155613
nsx(log(tm), df = 5, cure = TRUE, log = TRUE)1 3.323188
                                                           0.053166 62.5064
nsx(log(tm), df = 5, cure = TRUE, log = TRUE)2 3.628637
                                                           0.053160 68.2594
nsx(log(tm), df = 5, cure = TRUE, log = TRUE)3 1.634854
                                                           0.022465 72.7748
nsx(log(tm), df = 5, cure = TRUE, log = TRUE)4 6.592018
                                                           0.111504 59.1193
nsx(log(tm), df = 5, cure = TRUE, log = TRUE)5 3.371806
                                                           0.042788 78.8029
                                                   Pr(z)
(Intercept)
                                               < 2.2e-16 ***
I(year8594 == "Diagnosed 85-94")TRUE
                                               5.675e-10 ***
nsx(log(tm), df = 5, cure = TRUE, log = TRUE)1 < 2.2e-16 ***
nsx(log(tm), df = 5, cure = TRUE, log = TRUE)2 < 2.2e-16 ***
nsx(log(tm), df = 5, cure = TRUE, log = TRUE)3 < 2.2e-16 ***
nsx(log(tm), df = 5, cure = TRUE, log = TRUE)4 < 2.2e-16 ***
nsx(log(tm), df = 5, cure = TRUE, log = TRUE)5 < 2.2e-16 ***
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 0.8542896 0.8675857
2 0.7934666 0.7850101 0.8016338
3 0.6967422 0.6863187 0.7068971
4 0.8610835 0.8542896 0.8675857
5 0.8221253 0.8143224 0.8296357
6 0.8610835 0.8542896 0.8675857
```

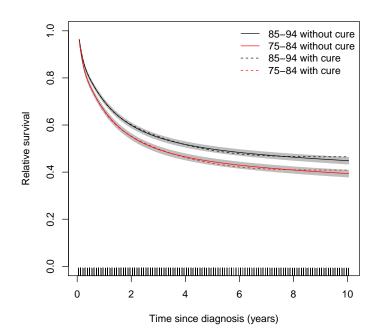
The estimate for the year parameter from the model without cure is within three significant figures with that in Stata. For the predictions, the Stata model gives:

	+-			+
	 -	surv	surv_lci	surv_uci
1.	i	.86108264	.8542898	.8675839
2.	1	.79346526	.7850106	.8016309
3.	1	.69674037	.6863196	.7068927
4.	-	.86108264	.8542898	.8675839
5.		.82212425	.8143227	.8296332
	-			
6.	I	.86108264	.8542898	.8675839
	+-			+

We can estimate the proportion of failures at the end of follow-up using:

```
> newdata.eof <- data.frame(year8594 = unique(colon2$year8594),
+ tm=max(colon2$tm))
> 1-predict(fit0, newdata.eof, type="surv", se.fit=TRUE)
```

```
Estimate
                lower
                          upper
1 0.6063672 0.6211742 0.5916005
2 0.5515138 0.5661344 0.5370100
> 1-predict(fit, newdata.eof, type="surv", se.fit=TRUE)
   Estimate
                lower
                          upper
1 0.5913307 0.6055051 0.5772138
2 0.5350813 0.5485401 0.5217406
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:

```
> plot(fit0,newdata=data.frame(year8594 = "Diagnosed 85-94"),
+ ylim=c(0,0.5), type="hazard",
```

```
+ xlab="Time since diagnosis (years)",ylab="Excess hazard")
> plot(fit0,newdata=data.frame(year8594 = "Diagnosed 75-84"),
+ type="hazard",
+ add=TRUE,line.col="red",rug=FALSE)
> plot(fit,newdata=data.frame(year8594 = "Diagnosed 85-94"),
+ type="hazard",
+ add=TRUE,ci=FALSE,lty=2,rug=FALSE)
> plot(fit,newdata=data.frame(year8594="Diagnosed 75-84"),
+ type="hazard",
+ 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")
```

