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.

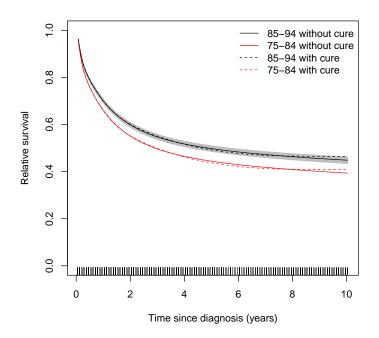
```
> require(rstpm2)
> data(popmort)
> data(colon)
> popmort2 <- transform(popmort,exitage=age,exityear=year,age=NULL,year=NULL)
> colon2 <- within(colon, {</pre>
    status <- ifelse(surv_mm>120.5,1,status)
    tm <- pmin(surv_mm,120.5)/12</pre>
    exit <- dx+tm*365.25
    sex <- as.numeric(sex)</pre>
    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 = neg11, start = coef, eval.only = TRUE, vecpar = TRUE,
    gr = function (beta)
        eta <- as.vector(X %*% beta)
        etaD <- as.vector(XD %*% beta)
```

```
h \leftarrow etaD * exp(eta) + bhazard
        g <- colSums(exp(eta) * wt * (-X + ifelse(event, 1/h,
            0) * (XD + X * etaD)))
        if (delayed) {
            eta0 <- as.vector(X0 %*% beta)
            g <- g + colSums(exp(eta0) * wt0 * X0)
        }
        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)
(Intercept)
                                     -3.977323
                                                 0.054778 - 72.6087 < 2.2e - 16
I(year8594 == "Diagnosed 85-94")TRUE -0.155612
                                                 0.025088 -6.2027 5.552e-10
nsx(log(tm), df = 5, cure = TRUE)1
                                      3.323191
                                                 0.053165 62.5071 < 2.2e-16
nsx(log(tm), df = 5, cure = TRUE)2
                                                 0.053159 68.2599 < 2.2e-16
                                      3.628630
nsx(log(tm), df = 5, cure = TRUE)3
                                                 0.022465 72.7746 < 2.2e-16
                                    1.634847
nsx(log(tm), df = 5, cure = TRUE)4
                                      6.592021
                                                 0.111503 59.1198 < 2.2e-16
nsx(log(tm), df = 5, cure = TRUE)5
                                      3.371809
                                                 0.042788 78.8032 < 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 0.8610828 0.8542898 0.8675842
2 0.7934651 0.7850103 0.8016309
3 0.6967400 0.6863191 0.7068926
4 0.8610828 0.8542898 0.8675842
5 0.8221243 0.8143226 0.8296334
6 0.8610828 0.8542898 0.8675842
```

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:

```
4. | .86108264 .8542898 .8675839 |
 5. | .82212425 .8143227 .8296332 |
    |-----|
 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 0.6063703 0.6211752 0.5916057
2 0.5515140 0.5661328 0.5370119
> 1-predict(fit, newdata.eof, type="surv", se.fit=TRUE)
  Estimate
               lower
                         upper
1 0.5913317 0.6055025 0.5772183
2 0.5350825 0.5485384 0.5217445
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")
```

3. | .69674037 .6863196 .7068927 |



And the hazard curves:

