

# model\_cox.R

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```
#####  
## load packages ##  
#####  
# Clear workspace variables  
rm(list = ls())  
cat("\014")
```

```
# options(stringsAsFactors=F)  
# install.packages("survival")  
# install.packages("Formula")  
# install.packages("ggplot2")  
# install.packages("Hmisc")  
# install.packages("lattice")  
# install.packages("SparseM")  
# install.packages("tinytex")  
# tinytex::install_tinytex(force = TRUE)  
library(survival)  
library(Formula)  
library(ggplot2)  
library(Hmisc)
```

```
## Loading required package: lattice
```

```
##
```

```
## Attaching package: 'Hmisc'
```

```
## The following objects are masked from 'package:base':
```

```
##
```

```
##      format.pval, units
```

```
library(SparseM)
```

```
##
```

```
## Attaching package: 'SparseM'
```

```
## The following object is masked from 'package:base':
```

```
##
```

```
##      backsolve
```

```
library(lattice)
```

```
# install.packages("sandwich")
```

```
library(sandwich)
```

```
# install.packages("rms")
```

```

library(rms)

#install.packages("devtools")
#library(devtools)
#install.version("rmarkdown",version=1.8)

#####
## 1. Set working directory and get data      ##
#####
setwd("~/Dropbox/Documents/Projects/DataScience/SurvivalELSA")
#elsa_cf <- read.csv("ELSA_CF_TRUE.csv", sep=",")
#save(elsa_cf,file='elsa_cf.rdata')

load("elsa.RData")
attach(elsa_cf)

#####
## 2. Cox regression models                  ##
#####

#####
## 2.1. Simple models                      ##
#####

# Sex only
cox1 <- coxph(Surv(time, death)~ factor(sex), data=elsa_cf)
summary(cox1)

## Call:
## coxph(formula = Surv(time, death) ~ factor(sex), data = elsa_cf)
##
##      n= 9282, number of events= 1802
##
##              coef exp(coef)  se(coef)      z Pr(>|z|)
## factor(sex)1 -0.001538  0.998463  0.047146 -0.033   0.974
##
##              exp(coef) exp(-coef) lower .95 upper .95
## factor(sex)1    0.9985      1.002    0.9103    1.095
##
## Concordance= 0.501  (se = 0.006 )
## Rsquare= 0      (max possible= 0.97 )
## Likelihood ratio test= 0  on 1 df,   p=1
## Wald test         = 0  on 1 df,   p=1
## Score (logrank) test = 0  on 1 df,   p=1

# Sex + age
cox2 <- coxph(Surv(time, death)~ age1 + factor(sex), data=elsa_cf)
summary(cox2)

## Call:
## coxph(formula = Surv(time, death) ~ age1 + factor(sex), data = elsa_cf)
##

```

```
## n= 9282, number of events= 1802
##
##          coef exp(coef) se(coef)      z Pr(>|z|)
## age1      0.09048   1.09470  0.00271 33.381 < 2e-16 ***
## factor(sex)1 -0.40120   0.66952  0.04922 -8.152 3.59e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
##          exp(coef) exp(-coef) lower .95 upper .95
## age1            1.0947    0.9135    1.0889    1.1005
## factor(sex)1     0.6695    1.4936    0.6079    0.7373
##
## Concordance= 0.714 (se = 0.007 )
## Rsquare= 0.117 (max possible= 0.97 )
## Likelihood ratio test= 1151 on 2 df, p=<2e-16
## Wald test = 1116 on 2 df, p=<2e-16
## Score (logrank) test = 1248 on 2 df, p=<2e-16

# LR test: simply use anova of the two model fits
# Example to test the effect of adding CHD into the model
cox3 <- coxph(Surv(time, death)~ age1 + factor(sex) + factor(chd1), data=elsa_cf)
summary(cox3)
```

```
## Call:
## coxph(formula = Surv(time, death) ~ age1 + factor(sex) + factor(chd1),
##       data = elsa_cf)
##
## n= 9282, number of events= 1802
##
##          coef exp(coef) se(coef)      z Pr(>|z|)
## age1      0.087625  1.091579  0.002748 31.881 < 2e-16 ***
## factor(sex)1 -0.360043  0.697647  0.049519 -7.271 3.57e-13 ***
## factor(chd1)1  0.445509  1.561284  0.057317  7.773 7.69e-15 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
##          exp(coef) exp(-coef) lower .95 upper .95
## age1            1.0916    0.9161    1.0857    1.0975
## factor(sex)1     0.6976    1.4334    0.6331    0.7688
## factor(chd1)1    1.5613    0.6405    1.3954    1.7469
##
## Concordance= 0.721 (se = 0.007 )
## Rsquare= 0.122 (max possible= 0.97 )
## Likelihood ratio test= 1207 on 3 df, p=<2e-16
## Wald test = 1182 on 3 df, p=<2e-16
## Score (logrank) test = 1323 on 3 df, p=<2e-16
```

```
# likelihood ratio test:
anova(cox2,cox3)
```

```
## Analysis of Deviance Table
## Cox model: response is Surv(time, death)
## Model 1: ~ age1 + factor(sex)
```

```

## Model 2: ~ age1 + factor(sex) + factor(chd1)
## loglik Chisq Df P(>|Chi|)
## 1 -15700
## 2 -15672 55.767 1 8.159e-14 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

#### This returns a very small p-value, meaning that adding CHD to the model improves the likelihood.
## We can also simply see this in the p-value of the effect estimate of CHD in model 3

#####
## 2.2. COVARIATES : EXPLORATION AND SELECTION ##
#####

# Model 4 : all potential covariates included #
cigst1=factor(cigst1)
educ1=factor(educ1)
cox4 <- coxph(Surv(time, death) ~ age1 + sex + chd1 + cancer1+ educ1_ + cigst1_ + physinact1
              + alcohol1 , data=elsa_cf, method="breslow")
summary(cox4)

## Call:
## coxph(formula = Surv(time, death) ~ age1 + sex + chd1 + cancer1 +
##      educ1_ + cigst1_ + physinact1 + alcohol1, data = elsa_cf,
##      method = "breslow")
##
##      n= 9282, number of events= 1802
##
##              coef exp(coef)  se(coef)      z Pr(>|z|)
## age1          0.082631  1.086141  0.002843 29.070 < 2e-16 ***
## sex           -0.422418  0.655460  0.051610 -8.185 2.73e-16 ***
## chd1           0.325514  1.384743  0.057888  5.623 1.87e-08 ***
## cancer1        0.672062  1.958271  0.073436  9.152 < 2e-16 ***
## educ1_1        0.208939  1.232370  0.092687  2.254 0.024180 *
## educ1_2        0.323894  1.382501  0.087840  3.687 0.000227 ***
## cigst1_1       0.249609  1.283523  0.057761  4.321 1.55e-05 ***
## cigst1_2       0.656862  1.928731  0.070076  9.374 < 2e-16 ***
## physinact1     0.660724  1.936194  0.049777 13.274 < 2e-16 ***
## alcohol1       0.039824  1.040628  0.054498  0.731 0.464931
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
##              exp(coef) exp(-coef) lower .95 upper .95
## age1          1.0861      0.9207      1.0801      1.0922
## sex           0.6555      1.5256      0.5924      0.7252
## chd1          1.3847      0.7222      1.2362      1.5511
## cancer1       1.9583      0.5107      1.6958      2.2614
## educ1_1       1.2324      0.8114      1.0277      1.4779
## educ1_2       1.3825      0.7233      1.1638      1.6422
## cigst1_1      1.2835      0.7791      1.1461      1.4374
## cigst1_2      1.9287      0.5185      1.6812      2.2127
## physinact1    1.9362      0.5165      1.7562      2.1346

```

```
## alcohol1      1.0406      0.9610      0.9352      1.1579
##
## Concordance= 0.756 (se = 0.007 )
## Rsquare= 0.158 (max possible= 0.97 )
## Likelihood ratio test= 1595 on 10 df, p=<2e-16
## Wald test = 1572 on 10 df, p=<2e-16
## Score (logrank) test = 1759 on 10 df, p=<2e-16
```

```
### ALL COVARIATES ARE ASSOCIATED EXCEPT ALCOHOL ###
```

```
# --> Now that we have selected our set of covariates to include in the model, we want to test the
## crude and adjusted effect of cognitive function on mortality risk
```

```
#### MODEL 5 : COGNITIVE FUNCTION SCORE (CONTINUOUS)
```

```
cox5a <- coxph(Surv(time, death) ~ cf1 + age1 + sex , data=elsa_cf, method="breslow") #only adjusted
cox5b <- coxph(Surv(time, death) ~ cf1 + age1 + sex + chd1 + cancer1+ educ1_ + cigst1_ + physinact1
, data=elsa_cf, method="breslow") # adjusted for covariates
summary(cox5a)
```

```
## Call:
## coxph(formula = Surv(time, death) ~ cf1 + age1 + sex, data = elsa_cf,
## method = "breslow")
##
## n= 9282, number of events= 1802
##
##          coef exp(coef) se(coef)      z Pr(>|z|)
## cf1 -0.040372  0.960433  0.002225 -18.141 < 2e-16 ***
## age1  0.070877  1.073449  0.002888  24.541 < 2e-16 ***
## sex -0.365865  0.693597  0.049446  -7.399 1.37e-13 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
##          exp(coef) exp(-coef) lower .95 upper .95
## cf1      0.9604      1.0412      0.9563      0.9646
## age1     1.0734      0.9316      1.0674      1.0795
## sex      0.6936      1.4418      0.6295      0.7642
##
## Concordance= 0.745 (se = 0.007 )
## Rsquare= 0.147 (max possible= 0.97 )
## Likelihood ratio test= 1478 on 3 df, p=<2e-16
## Wald test = 1450 on 3 df, p=<2e-16
## Score (logrank) test = 1580 on 3 df, p=<2e-16
```

```
summary(cox5b)
```

```
## Call:
## coxph(formula = Surv(time, death) ~ cf1 + age1 + sex + chd1 +
## cancer1 + educ1_ + cigst1_ + physinact1, data = elsa_cf,
## method = "breslow")
##
## n= 9282, number of events= 1802
##
```

```
##           coef exp(coef) se(coef)      z Pr(>|z|)
## cf1        -0.033219  0.967326  0.002301 -14.435 < 2e-16 ***
## age1         0.068654  1.071066  0.002970  23.114 < 2e-16 ***
## sex         -0.390652  0.676615  0.051587  -7.573 3.66e-14 ***
## chd1         0.281896  1.325641  0.057918   4.867 1.13e-06 ***
## cancer1      0.680085  1.974045  0.073530   9.249 < 2e-16 ***
## educ1_1      0.126269  1.134587  0.092713   1.362  0.173
## educ1_2      0.125449  1.133657  0.087978   1.426  0.154
## cigst1_1     0.256305  1.292146  0.057536   4.455 8.40e-06 ***
## cigst1_2     0.622097  1.862829  0.070074   8.878 < 2e-16 ***
## physinact1   0.543259  1.721608  0.050373  10.785 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
##           exp(coef) exp(-coef) lower .95 upper .95
## cf1           0.9673    1.0338    0.9630    0.9717
## age1           1.0711    0.9336    1.0648    1.0773
## sex            0.6766    1.4779    0.6115    0.7486
## chd1           1.3256    0.7544    1.1834    1.4850
## cancer1        1.9740    0.5066    1.7091    2.2801
## educ1_1        1.1346    0.8814    0.9461    1.3607
## educ1_2        1.1337    0.8821    0.9541    1.3470
## cigst1_1       1.2921    0.7739    1.1543    1.4464
## cigst1_2       1.8628    0.5368    1.6238    2.1371
## physinact1     1.7216    0.5809    1.5598    1.9003
##
## Concordance= 0.771 (se = 0.007 )
## Rsquare= 0.176 (max possible= 0.97 )
## Likelihood ratio test= 1802 on 10 df,  p=<2e-16
## Wald test              = 1780 on 10 df,  p=<2e-16
## Score (logrank) test = 1976 on 10 df,  p=<2e-16
```

```
## --> In the multivariable-adjusted model (5b), the increase in 1 point of cognitive function score
## was associated with a 3% decrease in mortality risk
## Explanation: a hazard ratio of 0.97 means a  $0.97-1=-0.03 \times 100=-3\%$ 
```

```
## Note: all covariates remain significantly associated except for education
```

```
## We create a variable cf1/10 to interpret the HR as the reduction in risk for an increase
### of 10 points of score (instead of 1)
elsa_cf$cf1_10<-elsa_cf$cf1/10
summary(elsa_cf$cf1)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##      4.00  40.00   47.00   47.19  55.00   94.00
```

```
summary(elsa_cf$cf1_10)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##      0.400  4.000   4.700   4.719  5.500   9.400
```

```
cox5c <- coxph(Surv(time, death) ~ cf1_10 + age1 + sex + chd1 + cancer1+ educ1_ + cigst1_ + physinact1,
  data=elsa_cf, method="breslow") # adjusted for covariates
summary(cox5c)
```

```
## Call:
## coxph(formula = Surv(time, death) ~ cf1_10 + age1 + sex + chd1 +
## cancer1 + educ1_ + cigst1_ + physinact1, data = elsa_cf,
## method = "breslow")
##
## n= 9282, number of events= 1802
```

```
##
##      coef exp(coef) se(coef)      z Pr(>|z|)
## cf1_10   -0.33219   0.71735  0.02301 -14.435 < 2e-16 ***
## age1      0.06865   1.07107  0.00297  23.114 < 2e-16 ***
## sex      -0.39065   0.67661  0.05159  -7.573 3.66e-14 ***
## chd1      0.28190   1.32564  0.05792   4.867 1.13e-06 ***
## cancer1   0.68009   1.97405  0.07353   9.249 < 2e-16 ***
## educ1_1   0.12627   1.13459  0.09271   1.362  0.173
## educ1_2   0.12545   1.13366  0.08798   1.426  0.154
## cigst1_1  0.25631   1.29215  0.05754   4.455 8.40e-06 ***
## cigst1_2  0.62210   1.86283  0.07007   8.878 < 2e-16 ***
## physinact1 0.54326   1.72161  0.05037  10.785 < 2e-16 ***
## ---
```

```
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
```

```
##      exp(coef) exp(-coef) lower .95 upper .95
## cf1_10      0.7173    1.3940    0.6857    0.7504
## age1        1.0711    0.9336    1.0648    1.0773
## sex         0.6766    1.4779    0.6115    0.7486
## chd1        1.3256    0.7544    1.1834    1.4850
## cancer1     1.9740    0.5066    1.7091    2.2801
## educ1_1     1.1346    0.8814    0.9461    1.3607
## educ1_2     1.1337    0.8821    0.9541    1.3470
## cigst1_1    1.2921    0.7739    1.1543    1.4464
## cigst1_2    1.8628    0.5368    1.6238    2.1371
## physinact1  1.7216    0.5809    1.5598    1.9003
##
```

```
## Concordance= 0.771 (se = 0.007 )
## Rsquare= 0.176 (max possible= 0.97 )
## Likelihood ratio test= 1802 on 10 df, p=<2e-16
## Wald test = 1780 on 10 df, p=<2e-16
## Score (logrank) test = 1976 on 10 df, p=<2e-16
```

```
## --> In the multivariable-adjusted model (5b), the increase in 10 points of cognitive function score
## was associated with a 28 % decrease in mortality risk
```

```
# Finally, we also look at it across quintiles
# Divide cognitive function score (quintiles)
quantile(elsa_cf$cf1, prob=c(0.20, 0.40, 0.60, 0.80))
```

```
## 20% 40% 60% 80%
## 38 45 50 57
```

```

elsa_cf$q_cf1 <-cut(elsa_cf$cf1, breaks=c(0, 38, 45, 50, 57, 194))
cox5c <- coxph(Surv(time, death) ~ factor(q_cf1) + age1 + sex + chd1 + cancer1+ educ1_ + cigst1_ + ph
, data=elsa_cf, method="breslow") # adjusted for covariates
summary(cox5c)

```

```

## Call:
## coxph(formula = Surv(time, death) ~ factor(q_cf1) + age1 + sex +
##       chd1 + cancer1 + educ1_ + cigst1_ + physinact1, data = elsa_cf,
##       method = "breslow")
##
## n= 9282, number of events= 1802
##
##               coef exp(coef) se(coef)      z Pr(>|z|)
## factor(q_cf1)(38,45] -0.501086  0.605872  0.062803 -7.979 1.48e-15 ***
## factor(q_cf1)(45,50] -0.582811  0.558327  0.074667 -7.805 5.93e-15 ***
## factor(q_cf1)(50,57] -0.707011  0.493116  0.079303 -8.915 < 2e-16 ***
## factor(q_cf1)(57,194] -0.964249  0.381269  0.102172 -9.437 < 2e-16 ***
## age1                   0.071182  1.073777  0.002971 23.956 < 2e-16 ***
## sex                   -0.386460  0.679458  0.051581 -7.492 6.77e-14 ***
## chd1                   0.283569  1.327861  0.057900  4.898 9.70e-07 ***
## cancer1               0.674264  1.962587  0.073496  9.174 < 2e-16 ***
## educ1_1               0.146432  1.157696  0.092824  1.578  0.1147
## educ1_2               0.163159  1.177224  0.088153  1.851  0.0642 .
## cigst1_1              0.260279  1.297292  0.057576  4.521 6.16e-06 ***
## cigst1_2              0.635277  1.887544  0.070076  9.066 < 2e-16 ***
## physinact1            0.558993  1.748911  0.050271 11.120 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
##               exp(coef) exp(-coef) lower .95 upper .95
## factor(q_cf1)(38,45]    0.6059    1.6505    0.5357    0.6852
## factor(q_cf1)(45,50]    0.5583    1.7911    0.4823    0.6463
## factor(q_cf1)(50,57]    0.4931    2.0279    0.4221    0.5760
## factor(q_cf1)(57,194]   0.3813    2.6228    0.3121    0.4658
## age1                    1.0738    0.9313    1.0675    1.0800
## sex                     0.6795    1.4718    0.6141    0.7517
## chd1                    1.3279    0.7531    1.1854    1.4874
## cancer1                 1.9626    0.5095    1.6993    2.2667
## educ1_1                 1.1577    0.8638    0.9651    1.3887
## educ1_2                 1.1772    0.8495    0.9904    1.3992
## cigst1_1                1.2973    0.7708    1.1589    1.4523
## cigst1_2                1.8875    0.5298    1.6453    2.1654
## physinact1              1.7489    0.5718    1.5848    1.9300
##
## Concordance= 0.769 (se = 0.007 )
## Rsquare= 0.172 (max possible= 0.97 )
## Likelihood ratio test= 1757 on 13 df, p=<2e-16
## Wald test = 1728 on 13 df, p=<2e-16
## Score (logrank) test = 2004 on 13 df, p=<2e-16

```

```

#####
## 2.3. TESTING FOR INTERACTIONS ##
#####

```



```

# LR test: simply use anova of the two model fits
# Example to test the effect of an interaction between age and cognitive function
## !!! THE MODELS HAVE TO BE NESTED
cox6a <- coxph(Surv(time, death)~ age1 + factor(sex) + cf1, data=elsa_cf)
cox6b <- coxph(Surv(time, death)~ age1 + factor(sex) + cf1 + age1*cf1, data=elsa_cf)
summary(cox6a)

```

```

## Call:
## coxph(formula = Surv(time, death) ~ age1 + factor(sex) + cf1,
##       data = elsa_cf)
##
##      n= 9282, number of events= 1802
##
##              coef exp(coef)  se(coef)      z Pr(>|z|)
## age1          0.070934  1.073511  0.002888  24.559 < 2e-16 ***
## factor(sex)1 -0.366040  0.693475  0.049446  -7.403 1.33e-13 ***
## cf1           -0.040408  0.960397  0.002226 -18.155 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
##              exp(coef) exp(-coef) lower .95 upper .95
## age1              1.0735      0.9315      1.0675      1.0796
## factor(sex)1      0.6935      1.4420      0.6294      0.7640
## cf1               0.9604      1.0412      0.9562      0.9646
##
## Concordance= 0.745  (se = 0.007 )
## Rsquare= 0.147  (max possible= 0.97 )
## Likelihood ratio test= 1479  on 3 df,  p=<2e-16
## Wald test              = 1451  on 3 df,  p=<2e-16
## Score (logrank) test = 1582  on 3 df,  p=<2e-16

```

```
summary(cox6b)
```

```

## Call:
## coxph(formula = Surv(time, death) ~ age1 + factor(sex) + cf1 +
##       age1 * cf1, data = elsa_cf)
##
##      n= 9282, number of events= 1802
##
##              coef exp(coef)  se(coef)      z Pr(>|z|)
## age1          0.0268762  1.0272406  0.0096647   2.781  0.00542 **
## factor(sex)1 -0.3476461  0.7063488  0.0493710  -7.042 1.90e-12 ***
## cf1          -0.1169013  0.8896730  0.0162517  -7.193 6.33e-13 ***
## age1:cf1      0.0010884  1.0010890  0.0002293   4.747 2.06e-06 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
##              exp(coef) exp(-coef) lower .95 upper .95
## age1              1.0272      0.9735      1.0080      1.0469
## factor(sex)1      0.7063      1.4157      0.6412      0.7781
## cf1               0.8897      1.1240      0.8618      0.9185

```

```
## age1:cf1      1.0011      0.9989      1.0006      1.0015
##
## Concordance= 0.747 (se = 0.007 )
## Rsquare= 0.149 (max possible= 0.97 )
## Likelihood ratio test= 1503 on 4 df, p=<2e-16
## Wald test          = 1314 on 4 df, p=<2e-16
## Score (logrank) test = 1765 on 4 df, p=<2e-16
```

```
# likelihood ratio test:
anova(cox6a,cox6b)
```

```
## Analysis of Deviance Table
## Cox model: response is Surv(time, death)
## Model 1: ~ age1 + factor(sex) + cf1
## Model 2: ~ age1 + factor(sex) + cf1 + age1 * cf1
## loglik Chisq Df P(>|Chi|)
## 1 -15536
## 2 -15524 23.199 1 1.461e-06 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
##### Very small p-value: evidence of a potential interaction between cognitive function and age, i.e.
## the effect of cognitive function on mortality differs according to age
```

```
# Create agegr (age in categories) and stratify analysis
elsa_cf$agegr[elsa_cf$age1 < 60] <- 1
elsa_cf$agegr[elsa_cf$age1>=60 & elsa_cf$age1 < 70] <- 2
elsa_cf$agegr[elsa_cf$age1 >= 70] <- 3
```

```
#Create subsets#
elsaage1 <-subset(elsa_cf, elsa_cf$agegr==1)
elsaage2 <-subset(elsa_cf, elsa_cf$agegr==2)
elsaage3 <-subset(elsa_cf, elsa_cf$agegr==3)
```

```
#Run models in each strata#
coxf.age1 <- coxph(Surv(time, death) ~ cf1_10 + age1 + sex + chd1 + cancer1+ factor(educ1) + factor(cigst1),
                  data=elsaage1, method="breslow")
coxf.age2 <- coxph(Surv(time, death) ~ cf1_10 + age1 + sex + chd1 + cancer1+ factor(educ1) + factor(cigst1),
                  data=elsaage2, method="breslow")
coxf.age3 <- coxph(Surv(time, death) ~ cf1_10 + age1 + sex + chd1 + cancer1+ factor(educ1) + factor(cigst1),
                  data=elsaage3, method="breslow")
summary(coxf.age1)
```

```
## Call:
## coxph(formula = Surv(time, death) ~ cf1_10 + age1 + sex + chd1 +
##       cancer1 + factor(educ1) + factor(cigst1) + physinact1, data = elsaage1,
##       method = "breslow")
##
## n= 3481, number of events= 298
##
##               coef exp(coef) se(coef)      z Pr(>|z|)
## cf1_10        -0.469458  0.625341  0.051903 -9.045 < 2e-16 ***
```

```
## age1          -0.007814  0.992217  0.025216 -0.310  0.756669
## sex           -0.229694  0.794777  0.127200 -1.806  0.070953 .
## chd1          0.635006  1.887033  0.150090  4.231  2.33e-05 ***
## cancer1       1.308207  3.699536  0.178544  7.327  2.35e-13 ***
## factor(educ1)1 0.614994  1.849646  0.238515  2.578  0.009925 **
## factor(educ1)2 0.620125  1.859161  0.230376  2.692  0.007107 **
## factor(cigst1)1 0.373330  1.452564  0.152088  2.455  0.014100 *
## factor(cigst1)2 0.603552  1.828603  0.163898  3.682  0.000231 ***
## physinact1     0.718401  2.051150  0.123450  5.819  5.91e-09 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
##               exp(coef) exp(-coef) lower .95 upper .95
## cf1_10         0.6253      1.5991    0.5649    0.6923
## age1           0.9922      1.0078    0.9444    1.0425
## sex            0.7948      1.2582    0.6194    1.0198
## chd1           1.8870      0.5299    1.4061    2.5324
## cancer1        3.6995      0.2703    2.6072    5.2496
## factor(educ1)1  1.8496      0.5406    1.1589    2.9520
## factor(educ1)2  1.8592      0.5379    1.1836    2.9202
## factor(cigst1)1  1.4526      0.6884    1.0781    1.9570
## factor(cigst1)2  1.8286      0.5469    1.3262    2.5213
## physinact1      2.0512      0.4875    1.6103    2.6126
##
## Concordance= 0.749 (se = 0.017 )
## Rsquare= 0.077 (max possible= 0.751 )
## Likelihood ratio test= 277.9 on 10 df,  p=<2e-16
## Wald test           = 308.6 on 10 df,  p=<2e-16
## Score (logrank) test = 338 on 10 df,  p=<2e-16
```

```
summary(coxf.age2)
```

```
## Call:
## coxph(formula = Surv(time, death) ~ cf1_10 + age1 + sex + chd1 +
##       cancer1 + factor(educ1) + factor(cigst1) + physinact1, data = elsaage2,
##       method = "breslow")
##
## n= 3121, number of events= 446
##
##               coef exp(coef) se(coef)      z Pr(>|z|)
## cf1_10        -0.31501  0.72979  0.04528 -6.956 3.50e-12 ***
## age1           0.08483  1.08853  0.01735  4.888 1.02e-06 ***
## sex           -0.40844  0.66469  0.09968 -4.098 4.17e-05 ***
## chd1           0.46562  1.59300  0.11473  4.058 4.94e-05 ***
## cancer1        0.96378  2.62159  0.13692  7.039 1.93e-12 ***
## factor(educ1)1 -0.03338  0.96717  0.17337 -0.193  0.847
## factor(educ1)2 -0.07842  0.92458  0.16081 -0.488  0.626
## factor(cigst1)1 0.50469  1.65647  0.12615  4.001 6.32e-05 ***
## factor(cigst1)2 0.85938  2.36169  0.13955  6.158 7.36e-10 ***
## physinact1      0.71370  2.04153  0.09947  7.175 7.24e-13 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
##               exp(coef) exp(-coef) lower .95 upper .95
```

```
## cf1_10          0.7298      1.3703      0.6678      0.7975
## age1            1.0885      0.9187      1.0521      1.1262
## sex             0.6647      1.5045      0.5467      0.8081
## chd1            1.5930      0.6277      1.2722      1.9947
## cancer1         2.6216      0.3814      2.0046      3.4285
## factor(educ1)1  0.9672      1.0339      0.6885      1.3586
## factor(educ1)2  0.9246      1.0816      0.6746      1.2671
## factor(cigst1)1 1.6565      0.6037      1.2936      2.1211
## factor(cigst1)2 2.3617      0.4234      1.7965      3.1046
## physinact1      2.0415      0.4898      1.6799      2.4810
##
## Concordance= 0.72 (se = 0.014 )
## Rsquare= 0.094 (max possible= 0.898 )
## Likelihood ratio test= 309.3 on 10 df, p=<2e-16
## Wald test          = 317.9 on 10 df, p=<2e-16
## Score (logrank) test = 340.1 on 10 df, p=<2e-16
```

```
summary(coxf.age3)
```

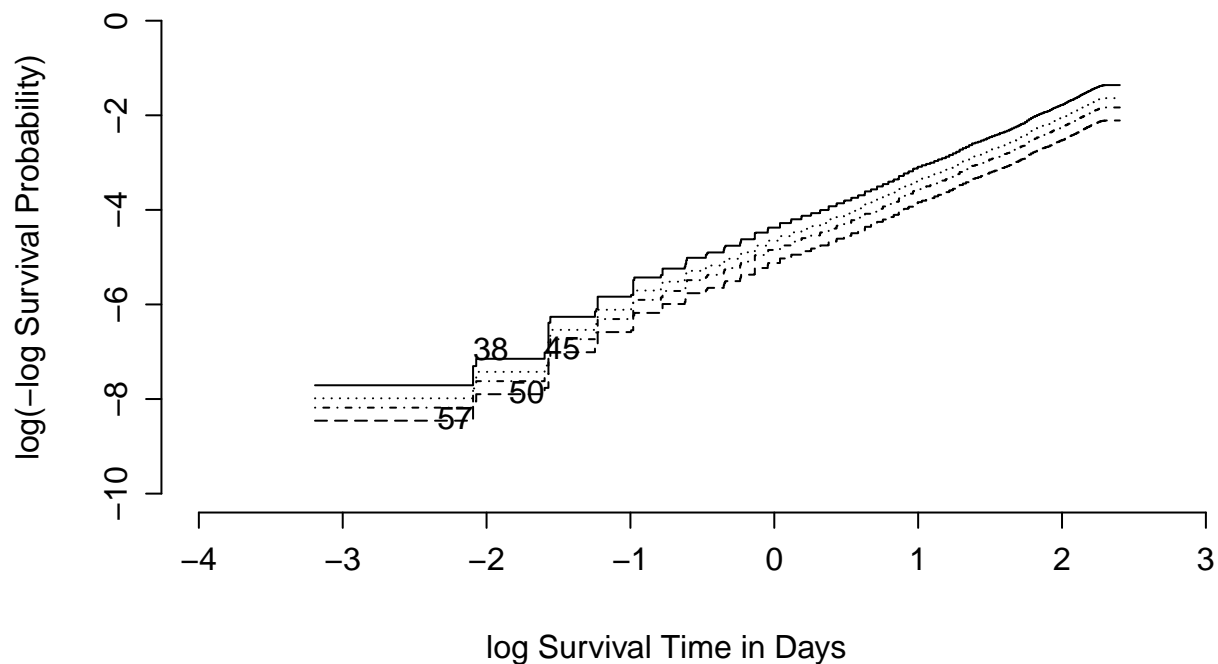
```
## Call:
## coxph(formula = Surv(time, death) ~ cf1_10 + age1 + sex + chd1 +
##       cancer1 + factor(educ1) + factor(cigst1) + physinact1, data = elsaage3,
##       method = "breslow")
##
## n= 2680, number of events= 1058
##
##              coef exp(coef) se(coef)      z Pr(>|z|)
## cf1_10        -0.275977  0.758830  0.030958 -8.915 < 2e-16 ***
## age1           0.070384  1.072920  0.007206  9.767 < 2e-16 ***
## sex           -0.426267  0.652942  0.071158 -5.990 2.09e-09 ***
## chd1           0.123884  1.131884  0.074506  1.663  0.0964 .
## cancer1        0.431101  1.538950  0.100310  4.298 1.73e-05 ***
## factor(educ1)1  0.003740  1.003747  0.124312  0.030  0.9760
## factor(educ1)2  0.012809  1.012892  0.118391  0.108  0.9138
## factor(cigst1)1 0.129655  1.138436  0.072112  1.798  0.0722 .
## factor(cigst1)2 0.543218  1.721538  0.097486  5.572 2.51e-08 ***
## physinact1      0.421812  1.524721  0.065467  6.443 1.17e-10 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
##              exp(coef) exp(-coef) lower .95 upper .95
## cf1_10          0.7588      1.3178      0.7142      0.8063
## age1            1.0729      0.9320      1.0579      1.0882
## sex             0.6529      1.5315      0.5679      0.7507
## chd1            1.1319      0.8835      0.9781      1.3099
## cancer1         1.5390      0.6498      1.2643      1.8733
## factor(educ1)1  1.0037      0.9963      0.7867      1.2807
## factor(educ1)2  1.0129      0.9873      0.8031      1.2774
## factor(cigst1)1 1.1384      0.8784      0.9884      1.3113
## factor(cigst1)2 1.7215      0.5809      1.4221      2.0840
## physinact1      1.5247      0.6559      1.3411      1.7335
##
## Concordance= 0.667 (se = 0.009 )
## Rsquare= 0.133 (max possible= 0.998 )
```

```
## Likelihood ratio test= 383.2 on 10 df, p=<2e-16
## Wald test = 382.6 on 10 df, p=<2e-16
## Score (logrank) test = 390.3 on 10 df, p=<2e-16
```

```
#####
## 3. CHECKING PROPORTIONAL HAZARDS ASSUMPTION ##
#####
```

```
# Plot 1- log (-log survival)
coxa <- cph(Surv(time, death) ~ cf1+age1 + sex + chd1,
            data=elsa_cf, x=TRUE,y=TRUE, method="breslow")

survplot(coxa, cf1=c(38,45,50,57), age1=mean(age1), sex=0, chd1=0,
          logt=TRUE, loglog=TRUE, xlim=c(-4, 2.5), ylim=c(-10, 0))
```

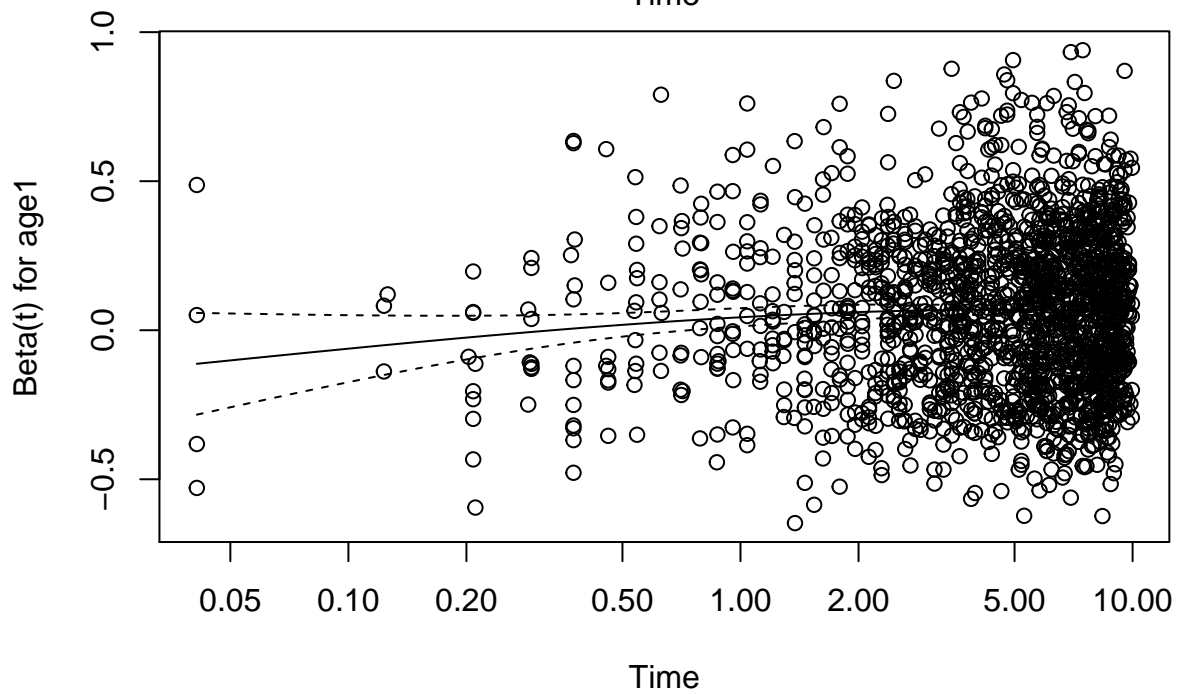
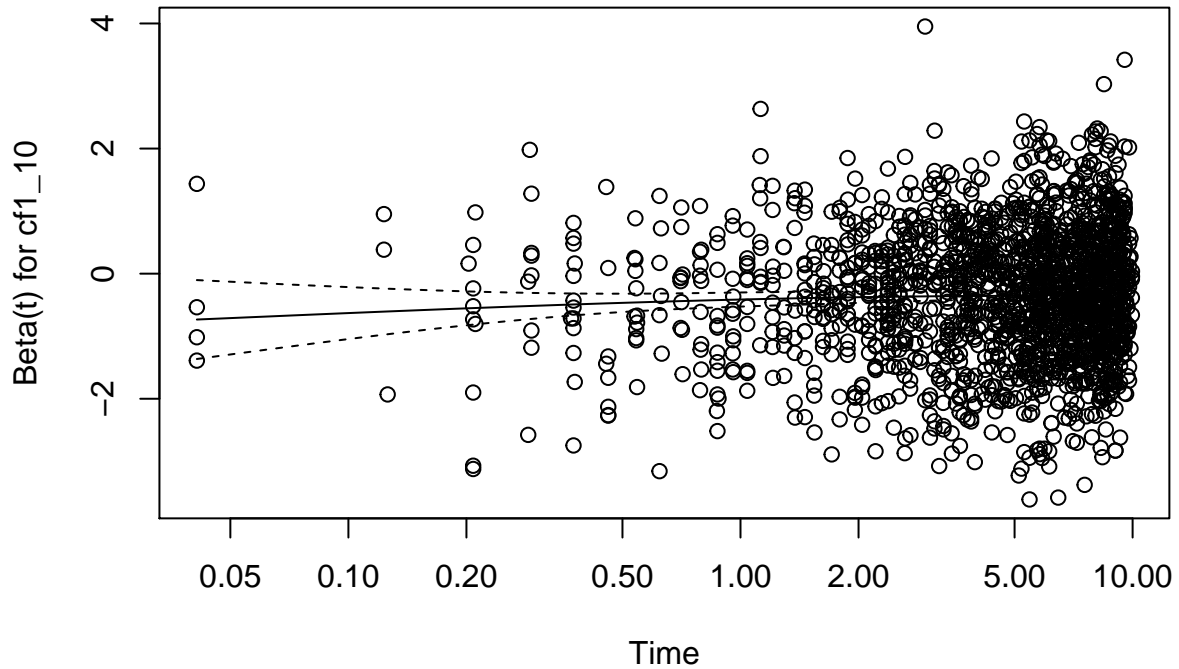


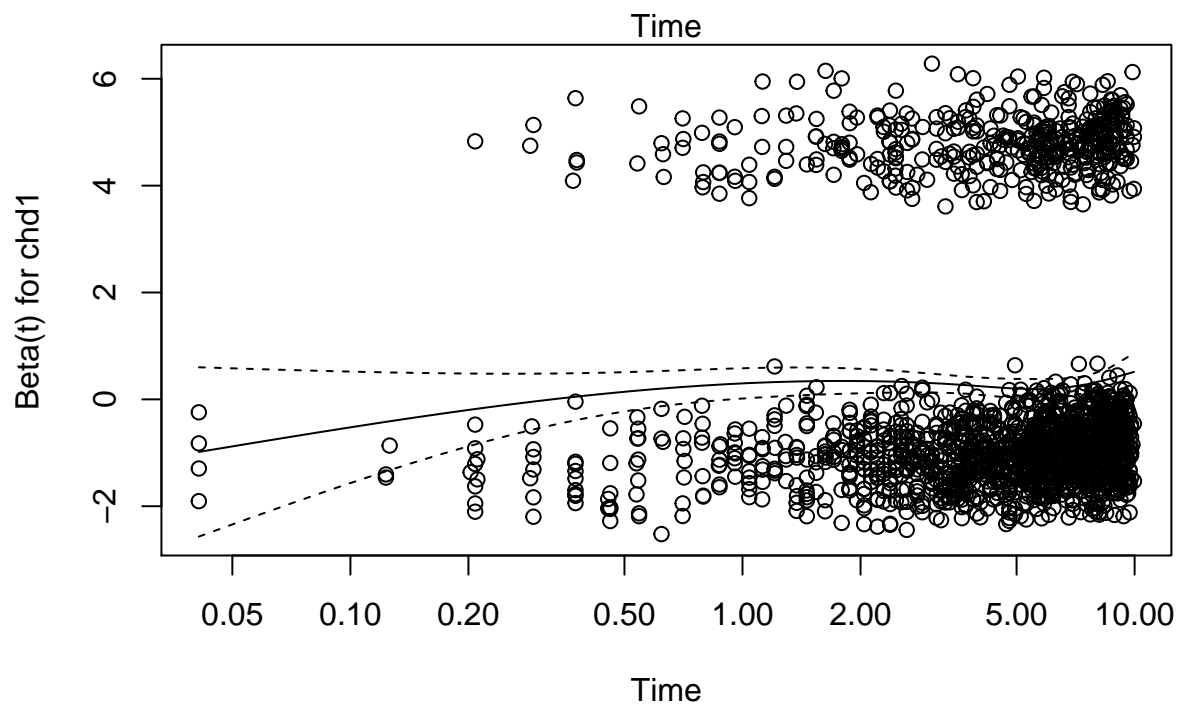
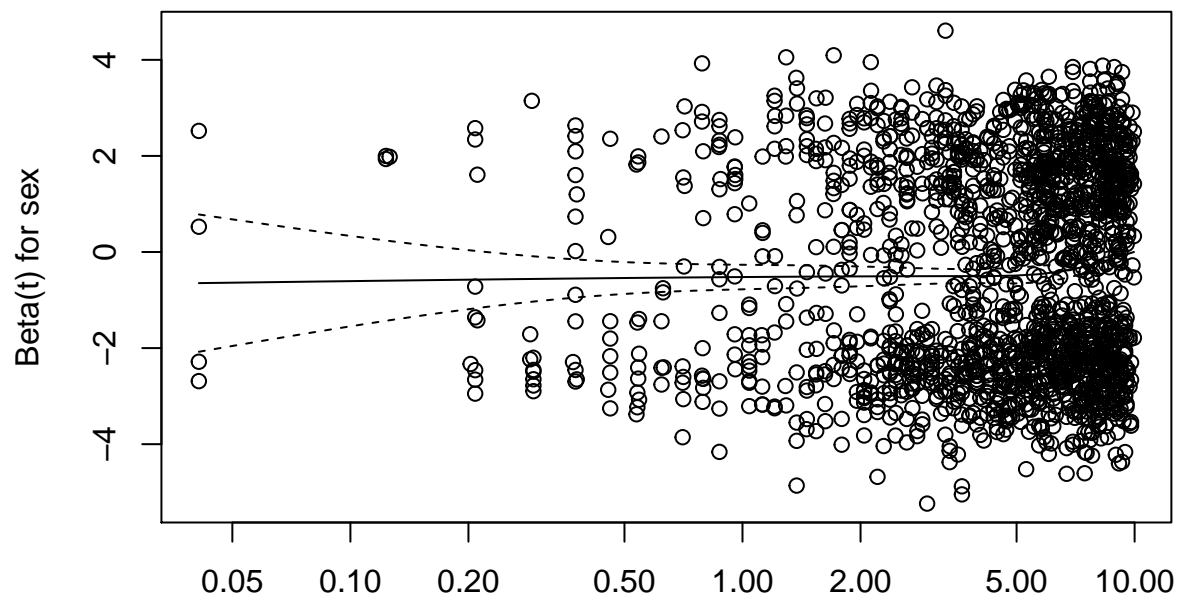
```
####
#Create a time-dependent model and test for interactions with time for all covariates
time.dep <- coxph (Surv(time, death)~ cf1_10 + age1 + sex + chd1 + cancer1 + physinact1 + educ1_ + cig
                strata(agegr), data=elsa_cf, method="breslow", na.action=na.exclude)
time.dep.zph <- cox.zph(time.dep, transform = 'log')
time.dep.zph
```

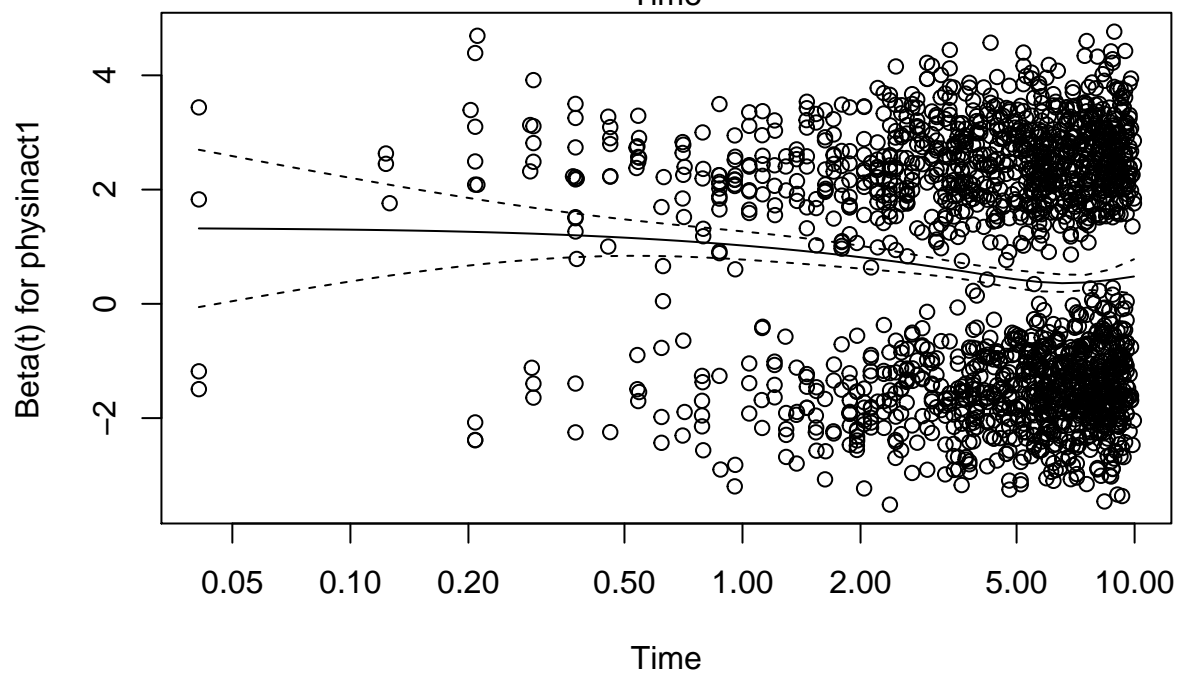
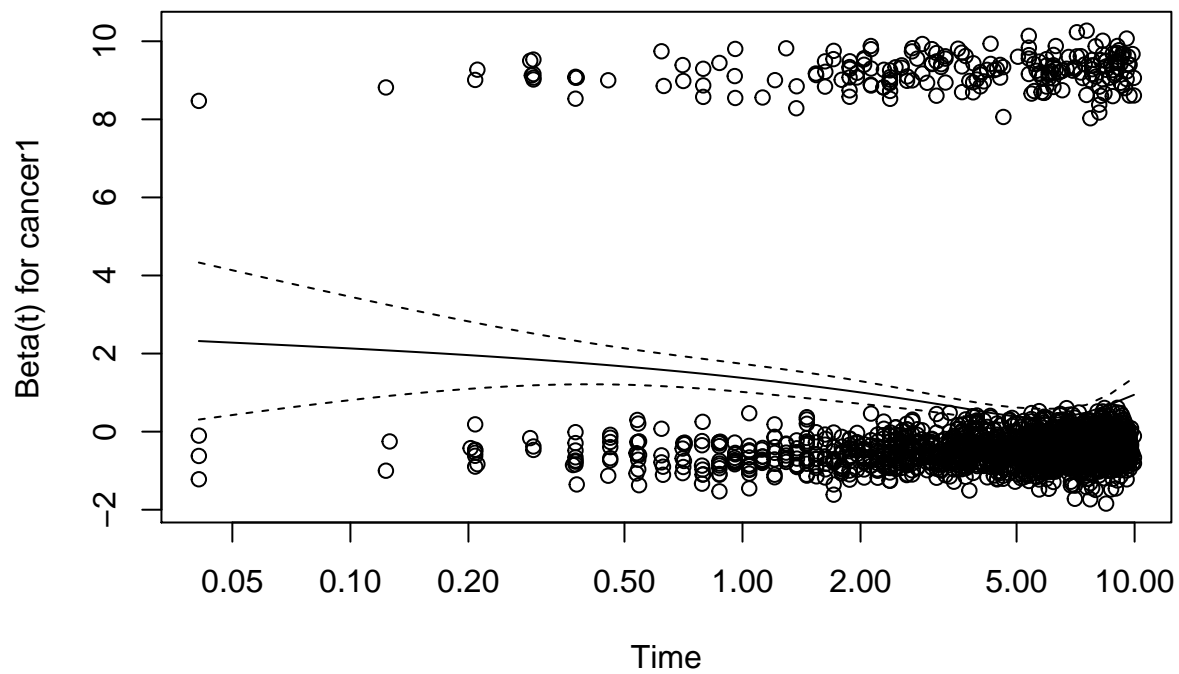
	rho	chisq	p
cf1_10	0.05421	6.09538	1.36e-02
age1	0.04702	4.15477	4.15e-02
sex	0.04764	4.12605	4.22e-02
chd1	0.02271	0.95357	3.29e-01
cancer1	-0.08866	14.43839	1.45e-04
physinact1	-0.09891	17.91530	2.31e-05
educ1_1	0.00666	0.08061	7.76e-01
educ1_2	0.01800	0.58677	4.44e-01

```
## cigst1_1    -0.00198  0.00708 9.33e-01
## cigst1_2    -0.01444  0.37023 5.43e-01
## GLOBAL      NA 49.74832 2.97e-07
```

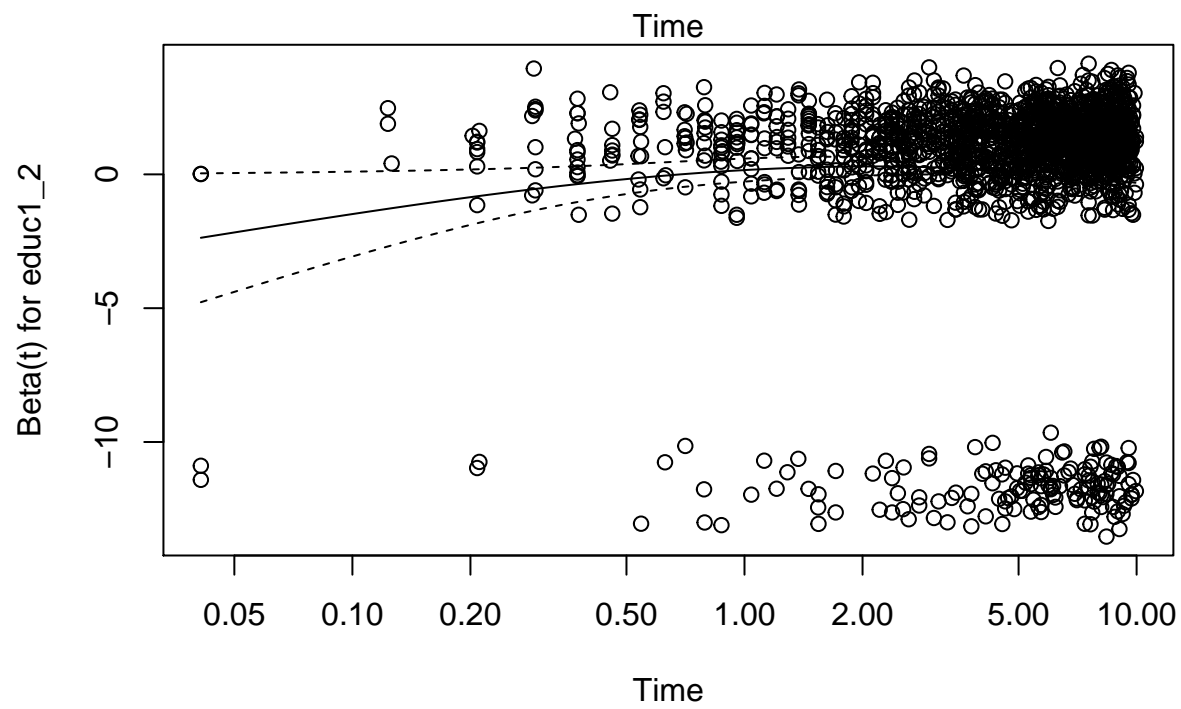
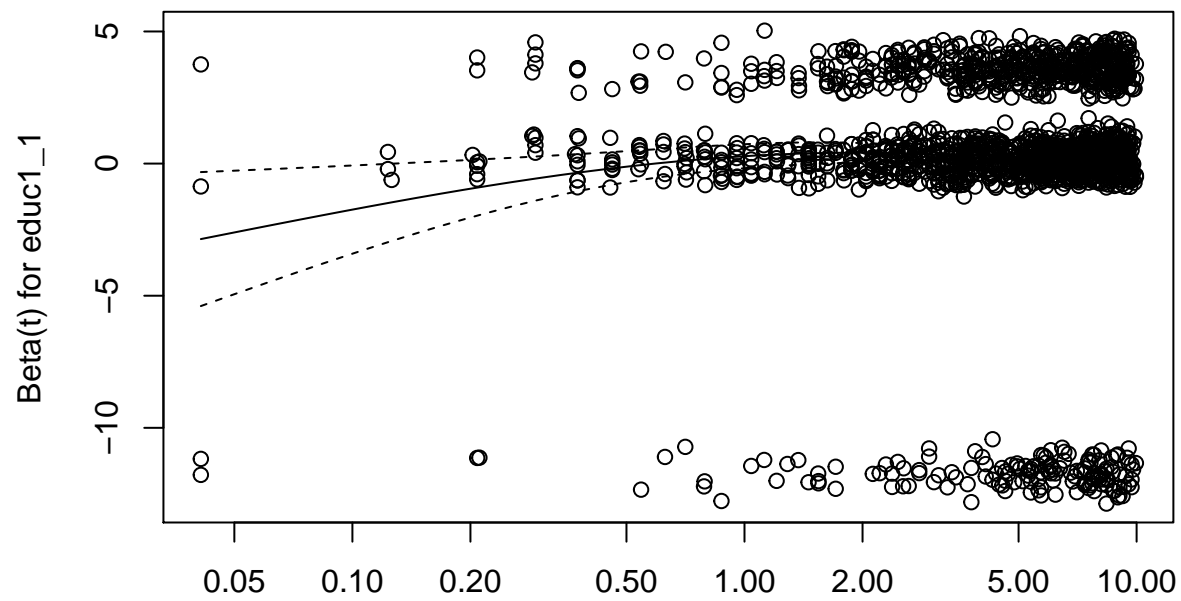
*# plots of residuals for all predictors (use arrows to go from one plot to the other)*  
`plot(time.dep.zph)`

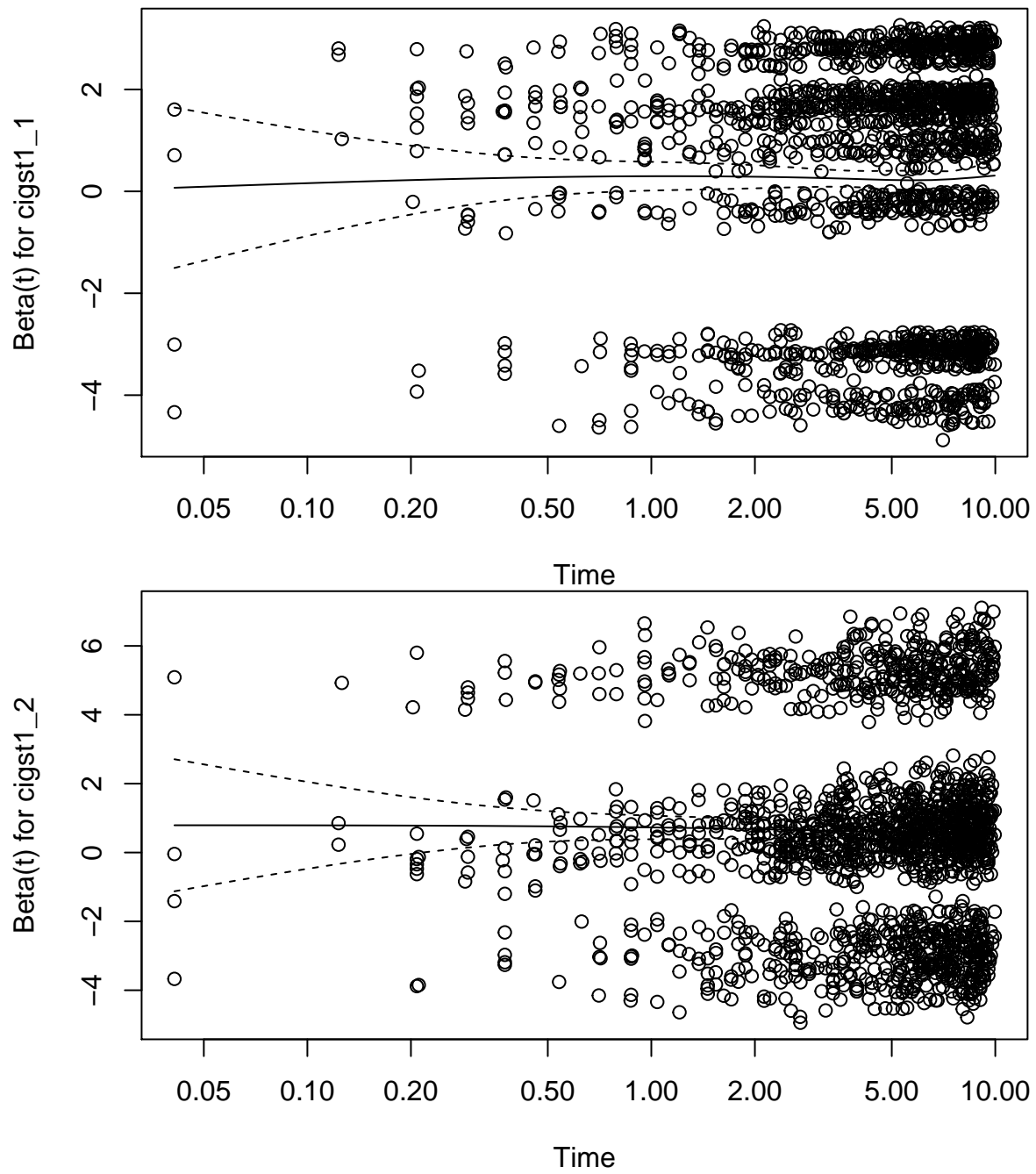












```
#plots for the predictor number in bracket. In the model, cf1_10 is the first predictor so we are plott
plot(time.dep.zph[1])
abline(h=0, lty=3)
```

