

Cox model

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This RMD file is corresponding to the Cox models in the main analysis following the second conceptual framework (the study is simplified as a cross-sectional study and time scale is time since entry)

Load data and packages

```
library(tidyverse)

## -- Attaching packages ----- tidyverse 1.3.1 --
## v ggplot2 3.3.5      v purrr 0.3.4
## v tibble 3.1.6       v dplyr 1.0.8
## v tidyr 1.2.0        v stringr 1.4.0
## v readr 2.1.2        v forcats 0.5.1

## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()     masks stats::lag()

library(survival)
dta <- readRDS("clean_data_0425.rds")

dta$fu_time <- ifelse(!is.na(dta$DIE_YR), dta$DIE_YR - dta$EXAM_YR,
                     dta$LAST_YR-dta$EXAM_YR)
# exclude people whose follow-up time is 0
dta <- dta %>% filter(fu_time > 0)
```

Main analysis

Model 1: Crude analysis

```
dta$B00ZE_ord <- dta$B00ZE_q %>% as.numeric()
# categorical version
fit1_q <- coxph(Surv(fu_time, cancer_death) ~ B00ZE_q, data = dta)
fit1_q %>% summary()

## Call:
## coxph(formula = Surv(fu_time, cancer_death) ~ B00ZE_q, data = dta)
##
##      n= 9190, number of events= 552
##
##              coef exp(coef) se(coef)      z Pr(>|z|)
```

```
## B00ZE_q(0,0.5] -0.05525    0.94625  0.15213 -0.363    0.716
## B00ZE_q(0.5,2] -0.17802    0.83693  0.12506 -1.423    0.155
## B00ZE_q(2,77]   0.14322    1.15398  0.09963  1.437    0.151
##
##               exp(coef) exp(-coef) lower .95 upper .95
## B00ZE_q(0,0.5]   0.9462    1.0568    0.7023    1.275
## B00ZE_q(0.5,2]   0.8369    1.1948    0.6550    1.069
## B00ZE_q(2,77]    1.1540    0.8666    0.9493    1.403
##
## Concordance= 0.532  (se = 0.012 )
## Likelihood ratio test= 6.57  on 3 df,   p=0.09
## Wald test              = 6.5  on 3 df,   p=0.09
## Score (logrank) test = 6.53  on 3 df,   p=0.09
# continuous version
fit1_con <- coxph(Surv(fu_time, cancer_death) ~ B00ZE, data = dta)
fit1_con %>% summary()
```

```
## Call:
## coxph(formula = Surv(fu_time, cancer_death) ~ B00ZE, data = dta)
##
##    n= 9190, number of events= 552
##
##               coef exp(coef) se(coef)      z Pr(>|z|)
## B00ZE 0.029295   1.029728 0.006454 4.539 5.66e-06 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
##               exp(coef) exp(-coef) lower .95 upper .95
## B00ZE      1.03      0.9711      1.017      1.043
##
## Concordance= 0.515  (se = 0.013 )
## Likelihood ratio test= 15.91  on 1 df,   p=7e-05
## Wald test              = 20.6  on 1 df,   p=6e-06
## Score (logrank) test = 20.32  on 1 df,   p=7e-06
```

```
# ordinal version for p-value for trend
fit1_ord <- coxph(Surv(fu_time, cancer_death) ~ B00ZE_ord, data = dta)
fit1_ord %>% summary()
```

```
## Call:
## coxph(formula = Surv(fu_time, cancer_death) ~ B00ZE_ord, data = dta)
##
##    n= 9190, number of events= 552
##
##               coef exp(coef) se(coef)      z Pr(>|z|)
## B00ZE_ord 0.03206   1.03258 0.03323 0.965    0.335
##
##               exp(coef) exp(-coef) lower .95 upper .95
## B00ZE_ord      1.033      0.9684      0.9675      1.102
##
## Concordance= 0.509  (se = 0.012 )
## Likelihood ratio test= 0.93  on 1 df,   p=0.3
## Wald test              = 0.93  on 1 df,   p=0.3
## Score (logrank) test = 0.93  on 1 df,   p=0.3
```

Model 2: Adjusting for age

```
# categorical version
fit2_q <- coxph(Surv(fu_time, cancer_death) ~ BOOZE_q + AGEYRS, data = dta)
fit2_q %>% summary()
```

```
## Call:
## coxph(formula = Surv(fu_time, cancer_death) ~ BOOZE_q + AGEYRS,
##       data = dta)
##
##      n= 9190, number of events= 552
##
##              coef exp(coef) se(coef)      z Pr(>|z|)
## BOOZE_q(0,0.5] 0.105880  1.111688 0.152333  0.695    0.487
## BOOZE_q(0.5,2] 0.123222  1.131135 0.125812  0.979    0.327
## BOOZE_q(2,77]  0.472044  1.603267 0.101003  4.674 2.96e-06 ***
## AGEYRS          0.070455  1.072996 0.004475 15.744 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
##              exp(coef) exp(-coef) lower .95 upper .95
## BOOZE_q(0,0.5]      1.112      0.8995      0.8247      1.498
## BOOZE_q(0.5,2]      1.131      0.8841      0.8839      1.447
## BOOZE_q(2,77]      1.603      0.6237      1.3153      1.954
## AGEYRS              1.073      0.9320      1.0636      1.082
##
## Concordance= 0.715 (se = 0.01 )
## Likelihood ratio test= 332.5 on 4 df,  p=<2e-16
## Wald test              = 255.1 on 4 df,  p=<2e-16
## Score (logrank) test = 291.1 on 4 df,  p=<2e-16
```

```
# continuous version
fit2_con <- coxph(Surv(fu_time, cancer_death) ~ BOOZE + AGEYRS, data = dta)
fit2_con %>% summary()
```

```
## Call:
## coxph(formula = Surv(fu_time, cancer_death) ~ BOOZE + AGEYRS,
##       data = dta)
##
##      n= 9190, number of events= 552
##
##              coef exp(coef) se(coef)      z Pr(>|z|)
## BOOZE  0.038740  1.039500 0.006129  6.321 2.6e-10 ***
## AGEYRS 0.069202  1.071653 0.004416 15.671 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
##              exp(coef) exp(-coef) lower .95 upper .95
## BOOZE      1.040      0.9620      1.027      1.052
## AGEYRS      1.072      0.9331      1.062      1.081
##
## Concordance= 0.719 (se = 0.01 )
## Likelihood ratio test= 339.9 on 2 df,  p=<2e-16
## Wald test              = 269.9 on 2 df,  p=<2e-16
## Score (logrank) test = 306.5 on 2 df,  p=<2e-16
```

```
# ordinal version for p-value for trend
fit2_ord <- coxph(Surv(fu_time, cancer_death) ~ BOOZE_ord + AGEYRS, data = dta)
fit2_ord %>% summary()
```

```
## Call:
## coxph(formula = Surv(fu_time, cancer_death) ~ BOOZE_ord + AGEYRS,
##       data = dta)
##
## n= 9190, number of events= 552
##
##              coef exp(coef) se(coef)      z Pr(>|z|)
## BOOZE_ord 0.14660   1.15789  0.03342  4.386 1.15e-05 ***
## AGEYRS    0.07063   1.07319  0.00447 15.800 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
##              exp(coef) exp(-coef) lower .95 upper .95
## BOOZE_ord      1.158      0.8636      1.084      1.236
## AGEYRS          1.073      0.9318      1.064      1.083
##
## Concordance= 0.715 (se = 0.01 )
## Likelihood ratio test= 329.9 on 2 df,  p=<2e-16
## Wald test              = 252.7 on 2 df,  p=<2e-16
## Score (logrank) test = 288.9 on 2 df,  p=<2e-16
```

Model 3: MV-adjusted analysis

```
# categorical version
fit3_q <- coxph(Surv(fu_time, cancer_death) ~ BOOZE_q + AGEYRS +
               SEX + as.factor(RACE) + GRADES + as.factor(SIZE) + as.factor(MARRY) + AVGSMD + as
               HTN_REP + RBC + DRMI + DIAB, data = dta)
fit3_q %>% summary()
```

```
## Call:
## coxph(formula = Surv(fu_time, cancer_death) ~ BOOZE_q + AGEYRS +
##       SEX + as.factor(RACE) + GRADES + as.factor(SIZE) + as.factor(MARRY) +
##       AVGSMD + as.factor(SMSA) + URBAN + HTN_REP + RBC + DRMI +
##       DIAB, data = dta)
##
## n= 9190, number of events= 552
##
##              coef exp(coef) se(coef)      z Pr(>|z|)
## BOOZE_q(0,0.5]    0.062381  1.064368  0.153886  0.405  0.6852
## BOOZE_q(0.5,2]   -0.022538  0.977714  0.130281 -0.173  0.8627
## BOOZE_q(2,77]    0.216482  1.241701  0.111603  1.940  0.0524 .
## AGEYRS            0.078181  1.081318  0.004890 15.988 < 2e-16 ***
## SEX              -0.533047  0.586814  0.101146 -5.270 1.36e-07 ***
## as.factor(RACE)2  -0.065050  0.937021  0.154848 -0.420  0.6744
## as.factor(RACE)3  -0.466283  0.627330  0.416720 -1.119  0.2632
## GRADES            -0.014233  0.985868  0.012522 -1.137  0.2557
## as.factor(SIZE)2   0.163250  1.177331  0.168056  0.971  0.3313
## as.factor(SIZE)3   0.353475  1.424007  0.166891  2.118  0.0342 *
## as.factor(SIZE)4   0.006977  1.007001  0.194056  0.036  0.9713
## as.factor(SIZE)5  -0.095880  0.908573  0.322831 -0.297  0.7665
```

```

## as.factor(SIZE)6    0.048661  1.049864  0.306838  0.159    0.8740
## as.factor(SIZE)7   -0.049226  0.951966  0.261103 -0.189    0.8505
## as.factor(SIZE)8    1.769602  5.868518  0.738832  2.395    0.0166 *
## as.factor(MARRY)3   0.024005  1.024296  0.129198  0.186    0.8526
## as.factor(MARRY)4   0.231038  1.259907  0.182961  1.263    0.2067
## as.factor(MARRY)5  -0.131390  0.876875  0.328963 -0.399    0.6896
## as.factor(MARRY)6  -0.157045  0.854666  0.216598 -0.725    0.4684
## as.factor(MARRY)8   0.086953  1.090846  0.712647  0.122    0.9029
## AVGSMK              0.025014  1.025329  0.002750  9.097    < 2e-16 ***
## as.factor(SMSA)2    -0.198918  0.819617  0.134333 -1.481    0.1387
## as.factor(SMSA)4    -0.077507  0.925421  0.226029 -0.343    0.7317
## URBAN               -1.775186  0.169452  0.745544 -2.381    0.0173 *
## HTN_REP             -0.053813  0.947609  0.091773 -0.586    0.5576
## RBC                 0.004711  1.004723  0.061802  0.076    0.9392
## DRMI                -0.095295  0.909104  0.171634 -0.555    0.5787
## DIAB                -0.118480  0.888269  0.185637 -0.638    0.5233
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
##               exp(coef) exp(-coef) lower .95 upper .95
## BOOZE_q(0,0.5]      1.0644      0.9395      0.7872      1.4391
## BOOZE_q(0.5,2]      0.9777      1.0228      0.7574      1.2621
## BOOZE_q(2,77]       1.2417      0.8053      0.9977      1.5453
## AGEYRS              1.0813      0.9248      1.0710      1.0917
## SEX                 0.5868      1.7041      0.4813      0.7155
## as.factor(RACE)2     0.9370      1.0672      0.6917      1.2693
## as.factor(RACE)3     0.6273      1.5941      0.2772      1.4197
## GRADES              0.9859      1.0143      0.9620      1.0104
## as.factor(SIZE)2     1.1773      0.8494      0.8469      1.6366
## as.factor(SIZE)3     1.4240      0.7022      1.0267      1.9750
## as.factor(SIZE)4     1.0070      0.9930      0.6884      1.4730
## as.factor(SIZE)5     0.9086      1.1006      0.4826      1.7106
## as.factor(SIZE)6     1.0499      0.9525      0.5754      1.9156
## as.factor(SIZE)7     0.9520      1.0505      0.5707      1.5881
## as.factor(SIZE)8     5.8685      0.1704      1.3792     24.9703
## as.factor(MARRY)3    1.0243      0.9763      0.7952      1.3195
## as.factor(MARRY)4    1.2599      0.7937      0.8802      1.8033
## as.factor(MARRY)5    0.8769      1.1404      0.4602      1.6709
## as.factor(MARRY)6    0.8547      1.1700      0.5590      1.3067
## as.factor(MARRY)8    1.0908      0.9167      0.2699      4.4093
## AVGSMK              1.0253      0.9753      1.0198      1.0309
## as.factor(SMSA)2     0.8196      1.2201      0.6299      1.0665
## as.factor(SMSA)4     0.9254      1.0806      0.5942      1.4412
## URBAN               0.1695      5.9014      0.0393      0.7306
## HTN_REP             0.9476      1.0553      0.7916      1.1343
## RBC                 1.0047      0.9953      0.8901      1.1341
## DRMI                0.9091      1.1000      0.6494      1.2726
## DIAB                0.8883      1.1258      0.6173      1.2781
##
## Concordance= 0.753 (se = 0.01 )
## Likelihood ratio test= 471.8 on 28 df,  p=<2e-16
## Wald test              = 402.1 on 28 df,  p=<2e-16
## Score (logrank) test = 428.7 on 28 df,  p=<2e-16

```

```

# continuous version
fit3_con <- coxph(Surv(fu_time, cancer_death) ~ BOOZE + AGEYRS +
                  SEX + as.factor(RACE) + GRADES + as.factor(SIZE)+ as.factor(MARRY) + AVGSMK + as.
                  HTN_REP + RBC + DRMI + DIAB, data = dta)
fit3_con %>% summary()

## Call:
## coxph(formula = Surv(fu_time, cancer_death) ~ BOOZE + AGEYRS +
##       SEX + as.factor(RACE) + GRADES + as.factor(SIZE) + as.factor(MARRY) +
##       AVGSMK + as.factor(SMSA) + URBAN + HTN_REP + RBC + DRMI +
##       DIAB, data = dta)
##
## n= 9190, number of events= 552
##
##              coef exp(coef)  se(coef)      z Pr(>|z|)
## BOOZE          0.025297  1.025620  0.007071   3.578 0.000347 ***
## AGEYRS          0.078123  1.081255  0.004866  16.056 < 2e-16 ***
## SEX           -0.509585  0.600745  0.100083  -5.092 3.55e-07 ***
## as.factor(RACE)2 -0.058130  0.943527  0.154639  -0.376 0.706985
## as.factor(RACE)3 -0.450551  0.637277  0.416578  -1.082 0.279451
## GRADES         -0.013815  0.986280  0.012375  -1.116 0.264277
## as.factor(SIZE)2  0.149436  1.161180  0.168014   0.889 0.373773
## as.factor(SIZE)3  0.343525  1.409909  0.166630   2.062 0.039245 *
## as.factor(SIZE)4  0.004909  1.004921  0.193847   0.025 0.979797
## as.factor(SIZE)5 -0.112964  0.893183  0.322738  -0.350 0.726326
## as.factor(SIZE)6  0.040942  1.041791  0.306999   0.133 0.893908
## as.factor(SIZE)7 -0.068461  0.933830  0.260874  -0.262 0.792990
## as.factor(SIZE)8  1.823447  6.193169  0.737773   2.472 0.013453 *
## as.factor(MARRY)3  0.016787  1.016929  0.129056   0.130 0.896505
## as.factor(MARRY)4  0.234022  1.263673  0.182686   1.281 0.200191
## as.factor(MARRY)5 -0.128498  0.879415  0.328669  -0.391 0.695822
## as.factor(MARRY)6 -0.159279  0.852758  0.216308  -0.736 0.461514
## as.factor(MARRY)8  0.095722  1.100453  0.712576   0.134 0.893140
## AVGSMK          0.024881  1.025193  0.002741   9.077 < 2e-16 ***
## as.factor(SMSA)2 -0.196148  0.821890  0.134305  -1.460 0.144162
## as.factor(SMSA)4 -0.059435  0.942297  0.225652  -0.263 0.792250
## URBAN          -1.844398  0.158121  0.744804  -2.476 0.013273 *
## HTN_REP        -0.057840  0.943801  0.091702  -0.631 0.528210
## RBC             0.009358  1.009401  0.061458   0.152 0.878984
## DRMI           -0.086226  0.917387  0.171515  -0.503 0.615154
## DIAB           -0.111931  0.894106  0.185001  -0.605 0.545160
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
##              exp(coef) exp(-coef) lower .95 upper .95
## BOOZE          1.0256    0.9750    1.01150    1.0399
## AGEYRS          1.0813    0.9249    1.07099    1.0916
## SEX             0.6007    1.6646    0.49374    0.7309
## as.factor(RACE)2  0.9435    1.0599    0.69683    1.2776
## as.factor(RACE)3  0.6373    1.5692    0.28167    1.4418
## GRADES          0.9863    1.0139    0.96265    1.0105
## as.factor(SIZE)2  1.1612    0.8612    0.83538    1.6140
## as.factor(SIZE)3  1.4099    0.7093    1.01708    1.9545
## as.factor(SIZE)4  1.0049    0.9951    0.68727    1.4694

```

```
## as.factor(SIZE)5      0.8932      1.1196      0.47449      1.6813
## as.factor(SIZE)6      1.0418      0.9599      0.57077      1.9015
## as.factor(SIZE)7      0.9338      1.0709      0.56003      1.5571
## as.factor(SIZE)8      6.1932      0.1615      1.45854     26.2970
## as.factor(MARRY)3     1.0169      0.9834      0.78966      1.3096
## as.factor(MARRY)4     1.2637      0.7913      0.88335      1.8077
## as.factor(MARRY)5     0.8794      1.1371      0.46178      1.6748
## as.factor(MARRY)6     0.8528      1.1727      0.55809      1.3030
## as.factor(MARRY)8     1.1005      0.9087      0.27229      4.4475
## AVGSMK                1.0252      0.9754      1.01970      1.0307
## as.factor(SMSA)2      0.8219      1.2167      0.63167      1.0694
## as.factor(SMSA)4      0.9423      1.0612      0.60550      1.4664
## URBAN                 0.1581      6.3243      0.03673      0.6807
## HTN_REP               0.9438      1.0595      0.78854      1.1296
## RBC                   1.0094      0.9907      0.89485      1.1386
## DRMI                  0.9174      1.0901      0.65548      1.2839
## DIAB                  0.8941      1.1184      0.62218      1.2849
```

```
##
## Concordance= 0.755 (se = 0.01 )
## Likelihood ratio test= 477.7 on 26 df, p=<2e-16
## Wald test              = 411.4 on 26 df, p=<2e-16
## Score (logrank) test = 438.2 on 26 df, p=<2e-16
```

```
# ordinal version for p-value for trend
```

```
fit3_ord <- coxph(Surv(fu_time, cancer_death) ~ BOOZE_ord + AGEYRS +
                  SEX + as.factor(RACE) + GRADES + as.factor(SIZE) + as.factor(MARRY) + AVGSMK + as
                  HTN_REP + RBC + DRMI + DIAB, data = dta)
fit3_ord %>% summary()
```

```
## Call:
## coxph(formula = Surv(fu_time, cancer_death) ~ BOOZE_ord + AGEYRS +
##       SEX + as.factor(RACE) + GRADES + as.factor(SIZE) + as.factor(MARRY) +
##       AVGSMK + as.factor(SMSA) + URBAN + HTN_REP + RBC + DRMI +
##       DIAB, data = dta)
##
## n= 9190, number of events= 552
##
##              coef exp(coef) se(coef)      z Pr(>|z|)
## BOOZE_ord      0.062151  1.064122  0.036847  1.687  0.0917 .
## AGEYRS          0.078365  1.081517  0.004886 16.039 < 2e-16 ***
## SEX            -0.538485  0.583632  0.100911 -5.336 9.49e-08 ***
## as.factor(RACE)2 -0.066634  0.935537  0.154802 -0.430  0.6669
## as.factor(RACE)3 -0.457329  0.632972  0.416624 -1.098  0.2723
## GRADES          -0.013534  0.986557  0.012497 -1.083  0.2788
## as.factor(SIZE)2  0.167929  1.182852  0.168016  0.999  0.3176
## as.factor(SIZE)3  0.360193  1.433606  0.166776  2.160  0.0308 *
## as.factor(SIZE)4  0.014382  1.014486  0.193901  0.074  0.9409
## as.factor(SIZE)5 -0.105015  0.900311  0.322890 -0.325  0.7450
## as.factor(SIZE)6  0.042319  1.043227  0.307031  0.138  0.8904
## as.factor(SIZE)7 -0.050068  0.951165  0.261082 -0.192  0.8479
## as.factor(SIZE)8  1.793805  6.012284  0.738198  2.430  0.0151 *
## as.factor(MARRY)3  0.025838  1.026174  0.129184  0.200  0.8415
## as.factor(MARRY)4  0.238907  1.269860  0.182786  1.307  0.1912
## as.factor(MARRY)5 -0.127184  0.880572  0.328802 -0.387  0.6989
## as.factor(MARRY)6 -0.159988  0.852154  0.216474 -0.739  0.4599
```

```
## as.factor(MARRY)8 0.086470 1.090319 0.712627 0.121 0.9034
## AVGSMK 0.024900 1.025212 0.002740 9.087 < 2e-16 ***
## as.factor(SMSA)2 -0.192976 0.824501 0.134187 -1.438 0.1504
## as.factor(SMSA)4 -0.064359 0.937668 0.225840 -0.285 0.7757
## URBAN -1.800171 0.165271 0.744872 -2.417 0.0157 *
## HTN_REP -0.051133 0.950153 0.091712 -0.558 0.5772
## RBC 0.005406 1.005421 0.061781 0.088 0.9303
## DRMI -0.105163 0.900178 0.171519 -0.613 0.5398
## DIAB -0.111261 0.894706 0.185551 -0.600 0.5488
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## exp(coef) exp(-coef) lower .95 upper .95
## BOOZE_ord 1.0641 0.9397 0.98998 1.1438
## AGEYRS 1.0815 0.9246 1.07121 1.0919
## SEX 0.5836 1.7134 0.47890 0.7113
## as.factor(RACE)2 0.9355 1.0689 0.69071 1.2672
## as.factor(RACE)3 0.6330 1.5798 0.27974 1.4322
## GRADES 0.9866 1.0136 0.96269 1.0110
## as.factor(SIZE)2 1.1829 0.8454 0.85097 1.6442
## as.factor(SIZE)3 1.4336 0.6975 1.03388 1.9879
## as.factor(SIZE)4 1.0145 0.9857 0.69374 1.4835
## as.factor(SIZE)5 0.9003 1.1107 0.47813 1.6953
## as.factor(SIZE)6 1.0432 0.9586 0.57152 1.9042
## as.factor(SIZE)7 0.9512 1.0513 0.57019 1.5867
## as.factor(SIZE)8 6.0123 0.1663 1.41477 25.5502
## as.factor(MARRY)3 1.0262 0.9745 0.79663 1.3219
## as.factor(MARRY)4 1.2699 0.7875 0.88750 1.8170
## as.factor(MARRY)5 0.8806 1.1356 0.46226 1.6774
## as.factor(MARRY)6 0.8522 1.1735 0.55751 1.3025
## as.factor(MARRY)8 1.0903 0.9172 0.26975 4.4070
## AVGSMK 1.0252 0.9754 1.01972 1.0307
## as.factor(SMSA)2 0.8245 1.2129 0.63383 1.0725
## as.factor(SMSA)4 0.9377 1.0665 0.60230 1.4598
## URBAN 0.1653 6.0507 0.03838 0.7116
## HTN_REP 0.9502 1.0525 0.79383 1.1373
## RBC 1.0054 0.9946 0.89076 1.1348
## DRMI 0.9002 1.1109 0.64318 1.2599
## DIAB 0.8947 1.1177 0.62193 1.2871
##
## Concordance= 0.753 (se = 0.01 )
## Likelihood ratio test= 469.8 on 26 df, p=<2e-16
## Wald test = 401.1 on 26 df, p=<2e-16
## Score (logrank) test = 427.2 on 26 df, p=<2e-16
```

Exploring effect modification by sex

Sex interaction terms in the Cox model

```
dta$SEX <- if_else(dta$SEX == 1, "Male", "Female")

# categorical version
fit4_q <- coxph(Surv(fu_time, cancer_death) ~ BOOZE_q + BOOZE_q:SEX + AGEYRS +
```



```

SEX + as.factor(RACE) + GRADES + as.factor(MARRY) + as.factor(SIZE) + AVGSMK + as
HTN_REP + RBC + DRMI + DIAB, data = dta)
fit4_q %>% summary()

```

```

## Call:
## coxph(formula = Surv(fu_time, cancer_death) ~ BOOZE_q + BOOZE_q:SEX +
## AGEYRS + SEX + as.factor(RACE) + GRADES + as.factor(MARRY) +
## as.factor(SIZE) + AVGSMK + as.factor(SMSA) + URBAN + HTN_REP +
## RBC + DRMI + DIAB, data = dta)
##
## n= 9190, number of events= 552
##
##               coef exp(coef) se(coef)      z Pr(>|z|)
## BOOZE_q(0,0.5]    0.302619  1.353399  0.207658  1.457   0.1450
## BOOZE_q(0.5,2]    0.286139  1.331278  0.190068  1.505   0.1322
## BOOZE_q(2,77]     0.463128  1.589037  0.183904  2.518   0.0118 *
## AGEYRS            0.078086  1.081215  0.004893 15.959 < 2e-16 ***
## SEXMale           0.783906  2.190009  0.139428  5.622 1.88e-08 ***
## as.factor(RACE)2  -0.066640  0.935532  0.154850 -0.430   0.6669
## as.factor(RACE)3  -0.467228  0.626737  0.416778 -1.121   0.2623
## GRADES            -0.014812  0.985297  0.012532 -1.182   0.2372
## as.factor(MARRY)3  0.030567  1.031039  0.129223  0.237   0.8130
## as.factor(MARRY)4  0.225202  1.252575  0.183173  1.229   0.2189
## as.factor(MARRY)5 -0.134948  0.873761  0.328997 -0.410   0.6817
## as.factor(MARRY)6 -0.156478  0.855150  0.216471 -0.723   0.4698
## as.factor(MARRY)8  0.023324  1.023598  0.713275  0.033   0.9739
## as.factor(SIZE)2   0.167515  1.182363  0.168075  0.997   0.3189
## as.factor(SIZE)3   0.361694  1.435759  0.166878  2.167   0.0302 *
## as.factor(SIZE)4   0.014946  1.015058  0.194024  0.077   0.9386
## as.factor(SIZE)5  -0.105367  0.899994  0.323012 -0.326   0.7443
## as.factor(SIZE)6   0.071535  1.074155  0.307017  0.233   0.8158
## as.factor(SIZE)7  -0.056982  0.944611  0.260965 -0.218   0.8272
## as.factor(SIZE)8   1.756008  5.789279  0.738936  2.376   0.0175 *
## AVGSMK            0.025052  1.025369  0.002761  9.073 < 2e-16 ***
## as.factor(SMSA)2  -0.198166  0.820234  0.134216 -1.476   0.1398
## as.factor(SMSA)4  -0.067451  0.934773  0.225930 -0.299   0.7653
## URBAN             -1.766685  0.170899  0.745551 -2.370   0.0178 *
## HTN_REP           -0.043962  0.956990  0.091853 -0.479   0.6322
## RBC                0.009946  1.009996  0.061852  0.161   0.8722
## DRMI              -0.090652  0.913335  0.171547 -0.528   0.5972
## DIAB              -0.112334  0.893746  0.185810 -0.605   0.5455
## BOOZE_q(0,0.5]:SEXMale -0.493872  0.610259  0.307189 -1.608   0.1079
## BOOZE_q(0.5,2]:SEXMale -0.555087  0.574023  0.252243 -2.201   0.0278 *
## BOOZE_q(2,77]:SEXMale -0.405682  0.666522  0.218054 -1.860   0.0628 .
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
##               exp(coef) exp(-coef) lower .95 upper .95
## BOOZE_q(0,0.5]    1.3534    0.7389   0.90088   2.0332
## BOOZE_q(0.5,2]    1.3313    0.7512   0.91724   1.9322
## BOOZE_q(2,77]     1.5890    0.6293   1.10814   2.2786
## AGEYRS            1.0812    0.9249   1.07090   1.0916
## SEXMale           2.1900    0.4566   1.66634   2.8782
## as.factor(RACE)2   0.9355    1.0689   0.69064   1.2673

```

```
## as.factor(RACE)3      0.6267      1.5956      0.27690      1.4186
## GRADES                0.9853      1.0149      0.96139      1.0098
## as.factor(MARRY)3     1.0310      0.9699      0.80035      1.3282
## as.factor(MARRY)4     1.2526      0.7984      0.87476      1.7936
## as.factor(MARRY)5     0.8738      1.1445      0.45851      1.6651
## as.factor(MARRY)6     0.8551      1.1694      0.55948      1.3071
## as.factor(MARRY)8     1.0236      0.9769      0.25292      4.1426
## as.factor(SIZE)2      1.1824      0.8458      0.85052      1.6437
## as.factor(SIZE)3      1.4358      0.6965      1.03522      1.9913
## as.factor(SIZE)4      1.0151      0.9852      0.69396      1.4847
## as.factor(SIZE)5      0.9000      1.1111      0.47785      1.6951
## as.factor(SIZE)6      1.0742      0.9310      0.58848      1.9607
## as.factor(SIZE)7      0.9446      1.0586      0.56640      1.5754
## as.factor(SIZE)8      5.7893      0.1727      1.36032      24.6381
## AVGSMK                1.0254      0.9753      1.01983      1.0309
## as.factor(SMSA)2      0.8202      1.2192      0.63051      1.0670
## as.factor(SMSA)4      0.9348      1.0698      0.60034      1.4555
## URBAN                  0.1709      5.8514      0.03964      0.7368
## HTN_REP                0.9570      1.0449      0.79932      1.1458
## RBC                    1.0100      0.9901      0.89469      1.1402
## DRMI                   0.9133      1.0949      0.65254      1.2784
## DIAB                   0.8937      1.1189      0.62094      1.2864
## BOOZE_q(0,0.5]:SEXMale 0.6103      1.6386      0.33422      1.1143
## BOOZE_q(0.5,2]:SEXMale 0.5740      1.7421      0.35012      0.9411
## BOOZE_q(2,77]:SEXMale  0.6665      1.5003      0.43472      1.0219
```

```
##
## Concordance= 0.754 (se = 0.01 )
## Likelihood ratio test= 478.9 on 31 df, p=<2e-16
## Wald test              = 408.7 on 31 df, p=<2e-16
## Score (logrank) test = 437.8 on 31 df, p=<2e-16
```

```
coef <- coef(fit4_q)[c(3, 31)] %>% as.matrix(ncol = 1)
vcov <- vcov(fit4_q)[c(3, 31), c(3, 31)]
var <- t(c(1, 1)) %*% vcov %*% c(1, 1)
sd <- sqrt(var)
mean_male <- coef[2,] + coef[1,]
exp(mean_male)
```

```
## BOOZE_q(2,77]:SEXMale
## 1.059128
```

```
c(mean_male - 1.96*sd, mean_male+ 1.96*sd) %>% exp()
```

```
## [1] 0.8187146 1.3701391
```

```
# Continuous version
```

```
fit4_con <- coxph(Surv(fu_time, cancer_death) ~ BOOZE + BOOZE:SEX + AGEYRS +
                SEX + as.factor(RACE) + GRADES + as.factor(MARRY) + as.factor(SIZE) + AVGSMK + as
                HTN_REP + RBC + DRMI + DIAB, data = dta)
fit4_con %>% summary()
```

```
## Call:
```

```
## coxph(formula = Surv(fu_time, cancer_death) ~ BOOZE + BOOZE:SEX +
## AGEYRS + SEX + as.factor(RACE) + GRADES + as.factor(MARRY) +
## as.factor(SIZE) + AVGSMK + as.factor(SMSA) + URBAN + HTN_REP +
## RBC + DRMI + DIAB, data = dta)
```

```

##
##   n= 9190, number of events= 552
##
##               coef exp(coef)   se(coef)      z Pr(>|z|)
## BOOZE          0.030800  1.031279  0.019053  1.617  0.1060
## AGEYRS          0.078126  1.081259  0.004866 16.054 < 2e-16 ***
## SEXMale         0.522100  1.685564  0.108219  4.825  1.4e-06 ***
## as.factor(RACE)2 -0.058404  0.943269  0.154645 -0.378  0.7057
## as.factor(RACE)3 -0.449974  0.637645  0.416591 -1.080  0.2801
## GRADES          -0.014027  0.986071  0.012392 -1.132  0.2577
## as.factor(MARRY)3  0.018414  1.018584  0.129168  0.143  0.8866
## as.factor(MARRY)4  0.234040  1.263695  0.182684  1.281  0.2002
## as.factor(MARRY)5 -0.127087  0.880657  0.328696 -0.387  0.6990
## as.factor(MARRY)6 -0.159358  0.852691  0.216306 -0.737  0.4613
## as.factor(MARRY)8  0.094250  1.098834  0.712587  0.132  0.8948
## as.factor(SIZE)2   0.150162  1.162023  0.168022  0.894  0.3715
## as.factor(SIZE)3   0.345427  1.412592  0.166742  2.072  0.0383 *
## as.factor(SIZE)4   0.007147  1.007173  0.193984  0.037  0.9706
## as.factor(SIZE)5  -0.113895  0.892352  0.322744 -0.353  0.7242
## as.factor(SIZE)6   0.041269  1.042133  0.306998  0.134  0.8931
## as.factor(SIZE)7  -0.069331  0.933017  0.260922 -0.266  0.7905
## as.factor(SIZE)8   1.825465  6.205679  0.737802  2.474  0.0134 *
## AVGSMK          0.024850  1.025161  0.002744  9.057 < 2e-16 ***
## as.factor(SMSA)2  -0.196562  0.821551  0.134291 -1.464  0.1433
## as.factor(SMSA)4  -0.056827  0.944757  0.225800 -0.252  0.8013
## URBAN           -1.846863  0.157731  0.744872 -2.479  0.0132 *
## HTN_REP         -0.056597  0.944975  0.091777 -0.617  0.5374
## RBC              0.009765  1.009813  0.061460  0.159  0.8738
## DRMI            -0.086738  0.916918  0.171529 -0.506  0.6131
## DIAB            -0.110376  0.895497  0.185070 -0.596  0.5509
## BOOZE:SEXMale    -0.006266  0.993753  0.020302 -0.309  0.7576
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
##               exp(coef) exp(-coef) lower .95 upper .95
## BOOZE          1.0313    0.9697    0.99348    1.0705
## AGEYRS          1.0813    0.9248    1.07099    1.0916
## SEXMale         1.6856    0.5933    1.36342    2.0838
## as.factor(RACE)2  0.9433    1.0601    0.69663    1.2772
## as.factor(RACE)3  0.6376    1.5683    0.28182    1.4427
## GRADES          0.9861    1.0141    0.96241    1.0103
## as.factor(MARRY)3  1.0186    0.9818    0.79077    1.3120
## as.factor(MARRY)4  1.2637    0.7913    0.88337    1.8078
## as.factor(MARRY)5  0.8807    1.1355    0.46240    1.6772
## as.factor(MARRY)6  0.8527    1.1728    0.55805    1.3029
## as.factor(MARRY)8  1.0988    0.9101    0.27188    4.4411
## as.factor(SIZE)2   1.1620    0.8606    0.83598    1.6152
## as.factor(SIZE)3   1.4126    0.7079    1.01879    1.9586
## as.factor(SIZE)4   1.0072    0.9929    0.68863    1.4731
## as.factor(SIZE)5   0.8924    1.1206    0.47404    1.6798
## as.factor(SIZE)6   1.0421    0.9596    0.57096    1.9021
## as.factor(SIZE)7   0.9330    1.0718    0.55949    1.5559
## as.factor(SIZE)8   6.2057    0.1611    1.46141    26.3516
## AVGSMK          1.0252    0.9755    1.01966    1.0307

```

```
## as.factor(SMSA)2      0.8216      1.2172      0.63143      1.0689
## as.factor(SMSA)4      0.9448      1.0585      0.60690      1.4707
## URBAN                 0.1577      6.3399      0.03663      0.6791
## HTN_REP               0.9450      1.0582      0.78940      1.1312
## RBC                   1.0098      0.9903      0.89521      1.1391
## DRMI                  0.9169      1.0906      0.65513      1.2833
## DIAB                  0.8955      1.1167      0.62306      1.2871
## BOOZE:SEXMale         0.9938      1.0063      0.95499      1.0341
```

```
##
## Concordance= 0.755 (se = 0.01 )
## Likelihood ratio test= 477.8 on 27 df, p=<2e-16
## Wald test              = 411 on 27 df, p=<2e-16
## Score (logrank) test = 438.4 on 27 df, p=<2e-16
```

```
coef <- coef(fit4_con)[c(1, 27)] %>% as.matrix(ncol = 1)
vcov <- vcov(fit4_con)[c(1, 27), c(1, 27)]
var <- t(c(1, 1)) %*% vcov %*% c(1, 1)
sd <- sqrt(var)
mean_male <- coef[2,] + coef[1,]
exp(mean_male)
```

```
## BOOZE:SEXMale
##      1.024837
```

```
c(mean_male - 1.96*sd, mean_male+ 1.96*sd) %>% exp()
```

```
## [1] 1.009782 1.040117
```

Stratified Cox model by sex, and adding an interaction term by SEX and alcohol consumption

```
# Categorical version
fit5_q <- coxph(Surv(fu_time, cancer_death) ~ BOOZE_q + strata(SEX) + BOOZE_q:strata(SEX) + AGEYRS + as.factor(RACE) + GRADES + as.factor(MARRY) + as.factor(SIZE) + AVGSMA + as.factor(SMSA) + URBAN + HTN_REP + RBC + DRMI + DIAB, data = dta)
fit5_q %>% summary()
```

```
## Call:
## coxph(formula = Surv(fu_time, cancer_death) ~ BOOZE_q + strata(SEX) + BOOZE_q:strata(SEX) + AGEYRS + as.factor(RACE) + GRADES + as.factor(MARRY) + as.factor(SIZE) + AVGSMA + as.factor(SMSA) + URBAN + HTN_REP + RBC + DRMI + DIAB, data = dta)
##
## n= 9190, number of events= 552
##
##              coef exp(coef) se(coef)      z Pr(>|z|)
## BOOZE_q(0,0.5]    0.300374  1.350364  0.207677  1.446  0.1481
## BOOZE_q(0.5,2]    0.280642  1.323980  0.190103  1.476  0.1399
## BOOZE_q(2,77]     0.460000  1.584074  0.183936  2.501  0.0124
## AGEYRS            0.077855  1.080966  0.004891 15.918 <2e-16
## as.factor(RACE)2   -0.066293  0.935857  0.154866 -0.428  0.6686
## as.factor(RACE)3   -0.466262  0.627343  0.416794 -1.119  0.2633
## GRADES            -0.014681  0.985426  0.012538 -1.171  0.2416
## as.factor(MARRY)3   0.032210  1.032734  0.129181  0.249  0.8031
## as.factor(MARRY)4   0.223209  1.250082  0.183193  1.218  0.2231
## as.factor(MARRY)5  -0.133737  0.874820  0.328993 -0.407  0.6844
```

```

## as.factor(MARRY)6      -0.158490  0.853432  0.216451 -0.732  0.4640
## as.factor(MARRY)8      0.028605  1.029018  0.713344  0.040  0.9680
## as.factor(SIZE)2       0.166094  1.180684  0.168084  0.988  0.3231
## as.factor(SIZE)3       0.362502  1.436921  0.166874  2.172  0.0298
## as.factor(SIZE)4       0.016400  1.016535  0.194023  0.085  0.9326
## as.factor(SIZE)5      -0.104615  0.900671  0.323020 -0.324  0.7460
## as.factor(SIZE)6       0.072781  1.075495  0.307076  0.237  0.8126
## as.factor(SIZE)7      -0.056924  0.944666  0.260976 -0.218  0.8273
## as.factor(SIZE)8       1.724272  5.608434  0.739043  2.333  0.0196
## AVGSMK                 0.024968  1.025282  0.002762  9.039  <2e-16
## as.factor(SMSA)2      -0.196934  0.821245  0.134203 -1.467  0.1423
## as.factor(SMSA)4      -0.069065  0.933266  0.225912 -0.306  0.7598
## URBAN                  -1.733064  0.176742  0.745653 -2.324  0.0201
## HTN_REP                -0.044148  0.956812  0.091822 -0.481  0.6307
## RBC                    0.012230  1.012305  0.061772  0.198  0.8431
## DRMI                   -0.094820  0.909537  0.171556 -0.553  0.5805
## DIAB                   -0.113823  0.892416  0.185809 -0.613  0.5402
## BOOZE_q(0,0.5]:strata(SEX)Male -0.491957  0.611428  0.307200 -1.601  0.1093
## BOOZE_q(0.5,2]:strata(SEX)Male -0.546088  0.579211  0.252285 -2.165  0.0304
## BOOZE_q(2,77]:strata(SEX)Male -0.398870  0.671078  0.218088 -1.829  0.0674
##
## BOOZE_q(0,0.5]
## BOOZE_q(0.5,2]
## BOOZE_q(2,77]          *
## AGEYRS                 ***
## as.factor(RACE)2
## as.factor(RACE)3
## GRADES
## as.factor(MARRY)3
## as.factor(MARRY)4
## as.factor(MARRY)5
## as.factor(MARRY)6
## as.factor(MARRY)8
## as.factor(SIZE)2
## as.factor(SIZE)3          *
## as.factor(SIZE)4
## as.factor(SIZE)5
## as.factor(SIZE)6
## as.factor(SIZE)7
## as.factor(SIZE)8          *
## AVGSMK                 ***
## as.factor(SMSA)2
## as.factor(SMSA)4
## URBAN                  *
## HTN_REP
## RBC
## DRMI
## DIAB
## BOOZE_q(0,0.5]:strata(SEX)Male
## BOOZE_q(0.5,2]:strata(SEX)Male *
## BOOZE_q(2,77]:strata(SEX)Male .
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##

```

	exp(coef)	exp(-coef)	lower .95	upper .95
##				
## BOOZE_q(0,0.5]	1.3504	0.7405	0.89883	2.0287
## BOOZE_q(0.5,2]	1.3240	0.7553	0.91215	1.9218
## BOOZE_q(2,77]	1.5841	0.6313	1.10461	2.2717
## AGEYRS	1.0810	0.9251	1.07065	1.0914
## as.factor(RACE)2	0.9359	1.0685	0.69086	1.2677
## as.factor(RACE)3	0.6273	1.5940	0.27716	1.4200
## GRADES	0.9854	1.0148	0.96151	1.0099
## as.factor(MARRY)3	1.0327	0.9683	0.80173	1.3303
## as.factor(MARRY)4	1.2501	0.7999	0.87298	1.7901
## as.factor(MARRY)5	0.8748	1.1431	0.45907	1.6671
## as.factor(MARRY)6	0.8534	1.1717	0.55837	1.3044
## as.factor(MARRY)8	1.0290	0.9718	0.25423	4.1651
## as.factor(SIZE)2	1.1807	0.8470	0.84930	1.6414
## as.factor(SIZE)3	1.4369	0.6959	1.03607	1.9929
## as.factor(SIZE)4	1.0165	0.9837	0.69498	1.4869
## as.factor(SIZE)5	0.9007	1.1103	0.47820	1.6964
## as.factor(SIZE)6	1.0755	0.9298	0.58915	1.9633
## as.factor(SIZE)7	0.9447	1.0586	0.56642	1.5755
## as.factor(SIZE)8	5.6084	0.1783	1.31755	23.8735
## AVGSMK	1.0253	0.9753	1.01975	1.0308
## as.factor(SMSA)2	0.8212	1.2177	0.63130	1.0683
## as.factor(SMSA)4	0.9333	1.0715	0.59939	1.4531
## URBAN	0.1767	5.6580	0.04099	0.7622
## HTN_REP	0.9568	1.0451	0.79922	1.1455
## RBC	1.0123	0.9878	0.89687	1.1426
## DRMI	0.9095	1.0995	0.64982	1.2731
## DIAB	0.8924	1.1206	0.62002	1.2845
## BOOZE_q(0,0.5]:strata(SEX)Male	0.6114	1.6355	0.33485	1.1164
## BOOZE_q(0.5,2]:strata(SEX)Male	0.5792	1.7265	0.35326	0.9497
## BOOZE_q(2,77]:strata(SEX)Male	0.6711	1.4901	0.43766	1.0290
##				

```
## Concordance= 0.738 (se = 0.01 )
## Likelihood ratio test= 429.9 on 30 df, p=<2e-16
## Wald test = 353.9 on 30 df, p=<2e-16
## Score (logrank) test = 383.4 on 30 df, p=<2e-16
```

```
coef <- coef(fit5_q)[c(3, 30)] %>% as.matrix(ncol = 1)
vcov <- vcov(fit5_q)[c(3, 30), c(3, 30)]
var <- t(c(1, 1)) %*% vcov %*% c(1, 1)
sd <- sqrt(var)
mean_male <- coef[2,] + coef[1,]
exp(mean_male)
```

```
## BOOZE_q(2,77]:strata(SEX)Male
## 1.063037
```

```
c(mean_male - 1.96*sd, mean_male+ 1.96*sd) %>% exp()
```

```
## [1] 0.8217616 1.3751522
```

```
# Continuous version
```

```
fit5_cont <- coxph(Surv(fu_time, cancer_death) ~ BOOZE + strata(SEX) + BOOZE:strata(SEX) + AGEYRS + as
HTN_REP + RBC + DRMI + DIAB, data = dta)
fit5_cont %>% summary()
```

```
## Call:
## coxph(formula = Surv(fu_time, cancer_death) ~ BOOZE + strata(SEX) +
##      BOOZE:strata(SEX) + AGEYRS + as.factor(RACE) + GRADES + as.factor(MARRY) +
##      as.factor(SIZE) + AVGSMK + as.factor(SMSA) + URBAN + HTN_REP +
##      RBC + DRMI + DIAB, data = dta)
##
##      n= 9190, number of events= 552
##
##              coef exp(coef)  se(coef)      z Pr(>|z|)
## BOOZE          0.030383  1.030849  0.019053  1.595  0.1108
## AGEYRS          0.077876  1.080989  0.004865 16.009 <2e-16 ***
## as.factor(RACE)2 -0.058446  0.943229  0.154663 -0.378  0.7055
## as.factor(RACE)3 -0.449498  0.637948  0.416599 -1.079  0.2806
## GRADES          -0.013846  0.986249  0.012398 -1.117  0.2641
## as.factor(MARRY)3  0.020074  1.020277  0.129126  0.155  0.8765
## as.factor(MARRY)4  0.231805  1.260874  0.182702  1.269  0.2045
## as.factor(MARRY)5 -0.125861  0.881738  0.328694 -0.383  0.7018
## as.factor(MARRY)6 -0.161532  0.850839  0.216294 -0.747  0.4552
## as.factor(MARRY)8  0.099779  1.104927  0.712624  0.140  0.8886
## as.factor(SIZE)2   0.148899  1.160556  0.168026  0.886  0.3755
## as.factor(SIZE)3   0.345704  1.412985  0.166733  2.073  0.0381 *
## as.factor(SIZE)4   0.008757  1.008795  0.193967  0.045  0.9640
## as.factor(SIZE)5  -0.113205  0.892968  0.322759 -0.351  0.7258
## as.factor(SIZE)6   0.042694  1.043619  0.307055  0.139  0.8894
## as.factor(SIZE)7  -0.069086  0.933246  0.260948 -0.265  0.7912
## as.factor(SIZE)8   1.794065  6.013851  0.737925  2.431  0.0150 *
## AVGSMK          0.024763  1.025072  0.002745  9.020 <2e-16 ***
## as.factor(SMSA)2  -0.195066  0.822781  0.134280 -1.453  0.1463
## as.factor(SMSA)4  -0.058536  0.943144  0.225792 -0.259  0.7954
## URBAN           -1.813968  0.163006  0.744990 -2.435  0.0149 *
## HTN_REP         -0.056634  0.944940  0.091743 -0.617  0.5370
## RBC              0.012133  1.012207  0.061377  0.198  0.8433
## DRMI            -0.090597  0.913386  0.171533 -0.528  0.5974
## DIAB            -0.111247  0.894718  0.185062 -0.601  0.5478
## BOOZE:strata(SEX)Male -0.005884  0.994133  0.020297 -0.290  0.7719
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
##              exp(coef) exp(-coef) lower .95 upper .95
## BOOZE          1.0308      0.9701   0.99306   1.070
## AGEYRS          1.0810      0.9251   1.07073   1.091
## as.factor(RACE)2  0.9432      1.0602   0.69658   1.277
## as.factor(RACE)3  0.6379      1.5675   0.28195   1.443
## GRADES          0.9862      1.0139   0.96257   1.011
## as.factor(MARRY)3  1.0203      0.9801   0.79215   1.314
## as.factor(MARRY)4  1.2609      0.7931   0.88136   1.804
## as.factor(MARRY)5  0.8817      1.1341   0.46297   1.679
## as.factor(MARRY)6  0.8508      1.1753   0.55685   1.300
## as.factor(MARRY)8  1.1049      0.9050   0.27337   4.466
## as.factor(SIZE)2   1.1606      0.8617   0.83491   1.613
## as.factor(SIZE)3   1.4130      0.7077   1.01909   1.959
## as.factor(SIZE)4   1.0088      0.9913   0.68976   1.475
## as.factor(SIZE)5   0.8930      1.1199   0.47435   1.681
## as.factor(SIZE)6   1.0436      0.9582   0.57171   1.905
```

```
## as.factor(SIZE)7      0.9332      1.0715      0.55960      1.556
## as.factor(SIZE)8      6.0139      0.1663      1.41589     25.543
## AVGSMK                1.0251      0.9755      1.01957      1.031
## as.factor(SMSA)2      0.8228      1.2154      0.63239      1.070
## as.factor(SMSA)4      0.9431      1.0603      0.60588      1.468
## URBAN                 0.1630      6.1347      0.03785      0.702
## HTN_REP               0.9449      1.0583      0.78943      1.131
## RBC                   1.0122      0.9879      0.89748      1.142
## DRMI                  0.9134      1.0948      0.65260      1.278
## DIAB                  0.8947      1.1177      0.62253      1.286
## BOOZE:strata(SEX)Male 0.9941      1.0059      0.95536      1.034
##
## Concordance= 0.739 (se = 0.01 )
## Likelihood ratio test= 428.9 on 26 df, p=<2e-16
## Wald test              = 355.5 on 26 df, p=<2e-16
## Score (logrank) test = 384.2 on 26 df, p=<2e-16
coef <- coef(fit5_cont)[c(1, 26)] %>% as.matrix(ncol = 1)
vcov <- vcov(fit5_cont)[c(1, 26), c(1, 26)]
var <- t(c(1, 1)) %*% vcov %*% c(1, 1)
sd <- sqrt(var)
mean_male <- coef[2,] + coef[1,]
exp(mean_male)
```

```
## BOOZE:strata(SEX)Male
##                1.024801
c(mean_male - 1.96*sd, mean_male+ 1.96*sd) %>% exp()

## [1] 1.009775 1.040050
```

Subgroup analysis

```
# categorical version
fit6_q_m <- coxph(Surv(fu_time, cancer_death) ~ BOOZE_q + AGEYRS + as.factor(RACE) + GRADES + as.factor(
  HTN_REP + RBC + DRMI + DIAB, data = dta[dta$SEX == "Male",])
fit6_q_m %>% summary()

## Call:
## coxph(formula = Surv(fu_time, cancer_death) ~ BOOZE_q + AGEYRS +
##       as.factor(RACE) + GRADES + as.factor(MARRY) + as.factor(SIZE) +
##       AVGSMK + as.factor(SMSA) + URBAN + HTN_REP + RBC + DRMI +
##       DIAB, data = dta[dta$SEX == "Male", ])
##
##      n= 4314, number of events= 329
##
##              coef exp(coef) se(coef)      z Pr(>|z|)
## BOOZE_q(0,0.5] -0.119329  0.887516  0.229390 -0.520  0.6029
## BOOZE_q(0.5,2] -0.219839  0.802648  0.174367 -1.261  0.2074
## BOOZE_q(2,77]  0.136033  1.145720  0.135379  1.005  0.3150
## AGEYRS         0.088349  1.092370  0.006724 13.140 < 2e-16 ***
## as.factor(RACE)2  0.014507  1.014613  0.190425  0.076  0.9393
## as.factor(RACE)3 -0.752925  0.470987  0.589271 -1.278  0.2013
## GRADES         -0.033238  0.967308  0.015547 -2.138  0.0325 *
## as.factor(MARRY)3 -0.040408  0.960397  0.227778 -0.177  0.8592
```



```

## as.factor(MARRY)4  0.114662  1.121494  0.271474  0.422  0.6728
## as.factor(MARRY)5 -0.269597  0.763687  0.424699 -0.635  0.5256
## as.factor(MARRY)6  0.083859  1.087475  0.264647  0.317  0.7513
## as.factor(MARRY)8 -0.065804  0.936315  1.015043 -0.065  0.9483
## as.factor(SIZE)2   0.219714  1.245721  0.223155  0.985  0.3248
## as.factor(SIZE)3   0.471345  1.602148  0.214932  2.193  0.0283 *
## as.factor(SIZE)4  -0.008270  0.991764  0.256648 -0.032  0.9743
## as.factor(SIZE)5  -0.032918  0.967618  0.417673 -0.079  0.9372
## as.factor(SIZE)6  -0.582896  0.558279  0.493430 -1.181  0.2375
## as.factor(SIZE)7  -0.066800  0.935382  0.356182 -0.188  0.8512
## as.factor(SIZE)8   1.639899  5.154651  0.768539  2.134  0.0329 *
## AVGSMDK           0.024089  1.024381  0.003367  7.155 8.35e-13 ***
## as.factor(SMSA)2  -0.173825  0.840444  0.176303 -0.986  0.3242
## as.factor(SMSA)4   0.213062  1.237462  0.318649  0.669  0.5037
## URBAN              -1.956302  0.141380  0.789058 -2.479  0.0132 *
## HTN_REP            -0.110896  0.895032  0.122633 -0.904  0.3658
## RBC                -0.118771  0.888011  0.094023 -1.263  0.2065
## DRMI               -0.049896  0.951328  0.193459 -0.258  0.7965
## DIAB               -0.044193  0.956769  0.238210 -0.186  0.8528
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
##
##               exp(coef) exp(-coef) lower .95 upper .95
## BOOZE_q(0,0.5]      0.8875      1.1267   0.56613   1.3913
## BOOZE_q(0.5,2]      0.8026      1.2459   0.57030   1.1297
## BOOZE_q(2,77]       1.1457      0.8728   0.87871   1.4939
## AGEYRS              1.0924      0.9154   1.07807   1.1069
## as.factor(RACE)2     1.0146      0.9856   0.69857   1.4736
## as.factor(RACE)3     0.4710      2.1232   0.14840   1.4949
## GRADES              0.9673      1.0338   0.93828   0.9972
## as.factor(MARRY)3    0.9604      1.0412   0.61456   1.5008
## as.factor(MARRY)4    1.1215      0.8917   0.65875   1.9093
## as.factor(MARRY)5    0.7637      1.3094   0.33221   1.7556
## as.factor(MARRY)6    1.0875      0.9196   0.64737   1.8268
## as.factor(MARRY)8    0.9363      1.0680   0.12806   6.8459
## as.factor(SIZE)2     1.2457      0.8027   0.80440   1.9292
## as.factor(SIZE)3     1.6021      0.6242   1.05136   2.4415
## as.factor(SIZE)4     0.9918      1.0083   0.59972   1.6401
## as.factor(SIZE)5     0.9676      1.0335   0.42676   2.1940
## as.factor(SIZE)6     0.5583      1.7912   0.21225   1.4685
## as.factor(SIZE)7     0.9354      1.0691   0.46538   1.8801
## as.factor(SIZE)8     5.1547      0.1940   1.14292  23.2478
## AVGSMDK             1.0244      0.9762   1.01764   1.0312
## as.factor(SMSA)2     0.8404      1.1898   0.59489   1.1873
## as.factor(SMSA)4     1.2375      0.8081   0.66267   2.3108
## URBAN                0.1414      7.0731   0.03011   0.6638
## HTN_REP              0.8950      1.1173   0.70381   1.1382
## RBC                  0.8880      1.1261   0.73856   1.0677
## DRMI                 0.9513      1.0512   0.65112   1.3900
## DIAB                 0.9568      1.0452   0.59985   1.5261
##
## Concordance= 0.772 (se = 0.012 )
## Likelihood ratio test= 333 on 27 df,  p=<2e-16
## Wald test              = 262.3 on 27 df,  p=<2e-16

```

```
## Score (logrank) test = 294.4 on 27 df, p=<2e-16
fit6_q_f <- coxph(Surv(fu_time, cancer_death) ~ BOOZE_q + AGEYRS + as.factor(RACE) + GRADES + as.factor(
  HTN_REP + RBC + DRMI + DIAB, data = dta[dta$SEX == "Female",])

## Warning in coxph.fit(X, Y, istrat, offset, init, control, weights = weights, :
## Loglik converged before variable 19,23 ; coefficient may be infinite.
fit6_q_f %>% summary()

## Call:
## coxph(formula = Surv(fu_time, cancer_death) ~ BOOZE_q + AGEYRS +
##       as.factor(RACE) + GRADES + as.factor(MARRY) + as.factor(SIZE) +
##       AVGSМК + as.factor(SMSA) + URBAN + HTN_REP + RBC + DRMI +
##       DIAB, data = dta[dta$SEX == "Female", ])
##
## n= 4876, number of events= 223
##
##              coef exp(coef) se(coef)      z Pr(>|z|)
## BOOZE_q(0,0.5]  2.346e-01  1.264e+00  2.101e-01  1.117  0.2641
## BOOZE_q(0.5,2]  1.785e-01  1.195e+00  1.959e-01  0.911  0.3621
## BOOZE_q(2,77]  3.028e-01  1.354e+00  1.940e-01  1.561  0.1186
## AGEYRS          6.231e-02  1.064e+00  7.383e-03  8.440 < 2e-16 ***
## as.factor(RACE)2 -3.387e-01  7.127e-01  2.776e-01 -1.220  0.2224
## as.factor(RACE)3 -1.188e-02  9.882e-01  5.903e-01 -0.020  0.9839
## GRADES          3.107e-02  1.032e+00  2.223e-02  1.398  0.1622
## as.factor(MARRY)3  1.736e-01  1.190e+00  1.624e-01  1.068  0.2854
## as.factor(MARRY)4  2.914e-01  1.338e+00  2.519e-01  1.157  0.2473
## as.factor(MARRY)5  1.062e-01  1.112e+00  5.257e-01  0.202  0.8400
## as.factor(MARRY)6 -5.246e-01  5.918e-01  3.937e-01 -1.332  0.1828
## as.factor(MARRY)8  5.692e-02  1.059e+00  1.010e+00  0.056  0.9551
## as.factor(SIZE)2   7.061e-02  1.073e+00  2.568e-01  0.275  0.7834
## as.factor(SIZE)3   1.723e-01  1.188e+00  2.672e-01  0.645  0.5190
## as.factor(SIZE)4   1.997e-02  1.020e+00  2.987e-01  0.067  0.9467
## as.factor(SIZE)5  -4.074e-01  6.654e-01  5.554e-01 -0.734  0.4632
## as.factor(SIZE)6   6.107e-01  1.842e+00  4.059e-01  1.505  0.1324
## as.factor(SIZE)7  -1.391e-01  8.701e-01  3.996e-01 -0.348  0.7277
## as.factor(SIZE)8  -1.101e+01  1.656e-05  1.123e+03 -0.010  0.9922
## AVGSМК          2.856e-02  1.029e+00  5.019e-03  5.691 1.27e-08 ***
## as.factor(SMSA)2  -2.101e-01  8.105e-01  2.083e-01 -1.009  0.3131
## as.factor(SMSA)4  -4.197e-01  6.572e-01  3.284e-01 -1.278  0.2012
## URBAN           1.138e+01  8.721e+04  1.123e+03  0.010  0.9919
## HTN_REP         1.116e-01  1.118e+00  1.408e-01  0.793  0.4281
## RBC             1.427e-01  1.153e+00  7.628e-02  1.871  0.0613 .
## DRMI           -3.667e-01  6.930e-01  3.878e-01 -0.946  0.3443
## DIAB           -2.153e-01  8.063e-01  3.017e-01 -0.713  0.4756
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
##              exp(coef) exp(-coef) lower .95 upper .95
## BOOZE_q(0,0.5]  1.264e+00  7.909e-01  0.8377  1.909
## BOOZE_q(0.5,2]  1.195e+00  8.365e-01  0.8143  1.755
## BOOZE_q(2,77]  1.354e+00  7.388e-01  0.9255  1.980
## AGEYRS          1.064e+00  9.396e-01  1.0490  1.080
## as.factor(RACE)2  7.127e-01  1.403e+00  0.4137  1.228
```

```

## as.factor(RACE)3 9.882e-01 1.012e+00 0.3107 3.143
## GRADES 1.032e+00 9.694e-01 0.9876 1.077
## as.factor(MARRY)3 1.190e+00 8.407e-01 0.8652 1.636
## as.factor(MARRY)4 1.338e+00 7.472e-01 0.8169 2.193
## as.factor(MARRY)5 1.112e+00 8.993e-01 0.3969 3.116
## as.factor(MARRY)6 5.918e-01 1.690e+00 0.2735 1.280
## as.factor(MARRY)8 1.059e+00 9.447e-01 0.1461 7.668
## as.factor(SIZE)2 1.073e+00 9.318e-01 0.6487 1.775
## as.factor(SIZE)3 1.188e+00 8.417e-01 0.7037 2.006
## as.factor(SIZE)4 1.020e+00 9.802e-01 0.5681 1.832
## as.factor(SIZE)5 6.654e-01 1.503e+00 0.2241 1.976
## as.factor(SIZE)6 1.842e+00 5.430e-01 0.8313 4.080
## as.factor(SIZE)7 8.701e-01 1.149e+00 0.3976 1.904
## as.factor(SIZE)8 1.656e-05 6.040e+04 0.0000 Inf
## AVGSMK 1.029e+00 9.718e-01 1.0189 1.039
## as.factor(SMSA)2 8.105e-01 1.234e+00 0.5388 1.219
## as.factor(SMSA)4 6.572e-01 1.522e+00 0.3453 1.251
## URBAN 8.721e+04 1.147e-05 0.0000 Inf
## HTN_REP 1.118e+00 8.944e-01 0.8484 1.473
## RBC 1.153e+00 8.670e-01 0.9932 1.339
## DRMI 6.930e-01 1.443e+00 0.3241 1.482
## DIAB 8.063e-01 1.240e+00 0.4463 1.457
##
## Concordance= 0.717 (se = 0.016 )
## Likelihood ratio test= 141.4 on 27 df, p=<2e-16
## Wald test = 124.5 on 27 df, p=2e-14
## Score (logrank) test = 132.7 on 27 df, p=6e-16

z <- (coef(fit6_q_f)[3] - coef(fit6_q_m)[3])/sqrt(vcov(fit6_q_f)[3,3] + vcov(fit6_q_m)[3,3])
1 - pnorm(abs(z))

## BOOZE_q(2,77]
## 0.2404372

# continuous version
fit6_cont_m <- coxph(Surv(fu_time, cancer_death) ~ BOOZE + AGEYRS + as.factor(RACE) + GRADES + as.factor(MARRY) +
HTN_REP + RBC + DRMI + DIAB, data = dta[dta$SEX == "Male",])
fit6_cont_m %>% summary()

## Call:
## coxph(formula = Surv(fu_time, cancer_death) ~ BOOZE + AGEYRS +
## as.factor(RACE) + GRADES + as.factor(MARRY) + as.factor(SIZE) +
## AVGSMK + as.factor(SMSA) + URBAN + HTN_REP + RBC + DRMI +
## DIAB, data = dta[dta$SEX == "Male", ])
##
## n= 4314, number of events= 329
##
## coef exp(coef) se(coef) z Pr(>|z|)
## BOOZE 0.027028 1.027396 0.007505 3.601 0.000316 ***
## AGEYRS 0.089025 1.093108 0.006717 13.254 < 2e-16 ***
## as.factor(RACE)2 0.032191 1.032715 0.190032 0.169 0.865485
## as.factor(RACE)3 -0.711462 0.490926 0.589152 -1.208 0.227200
## GRADES -0.034877 0.965724 0.015428 -2.261 0.023780 *
## as.factor(MARRY)3 -0.071713 0.930798 0.228175 -0.314 0.753303
## as.factor(MARRY)4 0.124238 1.132285 0.270507 0.459 0.646034

```

```

## as.factor(MARRY)5 -0.253290 0.776243 0.423828 -0.598 0.550091
## as.factor(MARRY)6 0.083721 1.087326 0.264406 0.317 0.751517
## as.factor(MARRY)8 0.021485 1.021718 1.014861 0.021 0.983110
## as.factor(SIZE)2 0.200740 1.222307 0.223233 0.899 0.368525
## as.factor(SIZE)3 0.467983 1.596770 0.215059 2.176 0.029550 *
## as.factor(SIZE)4 0.002177 1.002180 0.256571 0.008 0.993229
## as.factor(SIZE)5 -0.068238 0.934038 0.417371 -0.163 0.870129
## as.factor(SIZE)6 -0.612824 0.541819 0.493718 -1.241 0.214516
## as.factor(SIZE)7 -0.081216 0.921994 0.355378 -0.229 0.819230
## as.factor(SIZE)8 1.760029 5.812603 0.767788 2.292 0.021886 *
## AVGSMK 0.023583 1.023863 0.003348 7.044 1.87e-12 ***
## as.factor(SMSA)2 -0.164745 0.848110 0.176423 -0.934 0.350402
## as.factor(SMSA)4 0.265983 1.304713 0.318330 0.836 0.403404
## URBAN -2.085212 0.124281 0.788630 -2.644 0.008191 **
## HTN_REP -0.113310 0.892873 0.122585 -0.924 0.355308
## RBC -0.105150 0.900190 0.093337 -1.127 0.259927
## DRMI -0.050182 0.951056 0.193419 -0.259 0.795288
## DIAB -0.014320 0.985783 0.237405 -0.060 0.951903
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## exp(coef) exp(-coef) lower .95 upper .95
## BOOZE 1.0274 0.9733 1.01239 1.0426
## AGEYRS 1.0931 0.9148 1.07881 1.1076
## as.factor(RACE)2 1.0327 0.9683 0.71158 1.4988
## as.factor(RACE)3 0.4909 2.0370 0.15471 1.5578
## GRADES 0.9657 1.0355 0.93696 0.9954
## as.factor(MARRY)3 0.9308 1.0743 0.59516 1.4557
## as.factor(MARRY)4 1.1323 0.8832 0.66635 1.9240
## as.factor(MARRY)5 0.7762 1.2883 0.33825 1.7814
## as.factor(MARRY)6 1.0873 0.9197 0.64759 1.8257
## as.factor(MARRY)8 1.0217 0.9787 0.13979 7.4676
## as.factor(SIZE)2 1.2223 0.8181 0.78916 1.8932
## as.factor(SIZE)3 1.5968 0.6263 1.04757 2.4339
## as.factor(SIZE)4 1.0022 0.9978 0.60611 1.6571
## as.factor(SIZE)5 0.9340 1.0706 0.41219 2.1166
## as.factor(SIZE)6 0.5418 1.8456 0.20587 1.4260
## as.factor(SIZE)7 0.9220 1.0846 0.45944 1.8502
## as.factor(SIZE)8 5.8126 0.1720 1.29071 26.1766
## AVGSMK 1.0239 0.9767 1.01717 1.0306
## as.factor(SMSA)2 0.8481 1.1791 0.60018 1.1985
## as.factor(SMSA)4 1.3047 0.7665 0.69912 2.4349
## URBAN 0.1243 8.0463 0.02649 0.5830
## HTN_REP 0.8929 1.1200 0.70218 1.1354
## RBC 0.9002 1.1109 0.74970 1.0809
## DRMI 0.9511 1.0515 0.65098 1.3895
## DIAB 0.9858 1.0144 0.61902 1.5699
##
## Concordance= 0.775 (se = 0.011 )
## Likelihood ratio test= 338.7 on 25 df, p=<2e-16
## Wald test = 269.3 on 25 df, p=<2e-16
## Score (logrank) test = 301 on 25 df, p=<2e-16

```

```
fit6_cont_f <- coxph(Surv(fu_time, cancer_death) ~ BOOZE + AGEYRS + as.factor(RACE) + GRADES + as.factor(
  HTN_REP + RBC + DRMI + DIAB, data = dta[dta$SEX == "Female",])
```

```
## Warning in coxph.fit(X, Y, istrat, offset, init, control, weights = weights, :
## Loglik converged before variable 17,21 ; coefficient may be infinite.
```

```
fit6_cont_f %>% summary()
```

```
## Call:
```

```
## coxph(formula = Surv(fu_time, cancer_death) ~ BOOZE + AGEYRS +
##       as.factor(RACE) + GRADES + as.factor(MARRY) + as.factor(SIZE) +
##       AVGSMK + as.factor(SMSA) + URBAN + HTN_REP + RBC + DRMI +
##       DIAB, data = dta[dta$SEX == "Female", ])
```

```
##
```

```
## n= 4876, number of events= 223
```

```
##
```

	coef	exp(coef)	se(coef)	z	Pr(> z)
## BOOZE	1.422e-02	1.014e+00	2.099e-02	0.677	0.4981
## AGEYRS	6.111e-02	1.063e+00	7.301e-03	8.370	< 2e-16 ***
## as.factor(RACE)2	-3.515e-01	7.037e-01	2.771e-01	-1.268	0.2047
## as.factor(RACE)3	-3.810e-02	9.626e-01	5.899e-01	-0.065	0.9485
## GRADES	3.648e-02	1.037e+00	2.195e-02	1.662	0.0965 .
## as.factor(MARRY)3	1.812e-01	1.199e+00	1.624e-01	1.116	0.2643
## as.factor(MARRY)4	2.890e-01	1.335e+00	2.519e-01	1.148	0.2511
## as.factor(MARRY)5	1.023e-01	1.108e+00	5.251e-01	0.195	0.8455
## as.factor(MARRY)6	-5.407e-01	5.824e-01	3.937e-01	-1.373	0.1697
## as.factor(MARRY)8	7.725e-02	1.080e+00	1.010e+00	0.077	0.9390
## as.factor(SIZE)2	5.776e-02	1.059e+00	2.565e-01	0.225	0.8218
## as.factor(SIZE)3	1.403e-01	1.151e+00	2.664e-01	0.527	0.5985
## as.factor(SIZE)4	-1.325e-02	9.868e-01	2.982e-01	-0.044	0.9646
## as.factor(SIZE)5	-3.917e-01	6.759e-01	5.549e-01	-0.706	0.4803
## as.factor(SIZE)6	5.865e-01	1.798e+00	4.052e-01	1.448	0.1477
## as.factor(SIZE)7	-1.499e-01	8.608e-01	3.995e-01	-0.375	0.7076
## as.factor(SIZE)8	-1.109e+01	1.534e+05	1.125e+03	-0.010	0.9921
## AVGSMK	2.910e-02	1.030e+00	4.956e-03	5.871	4.33e-09 ***
## as.factor(SMSA)2	-2.106e-01	8.101e-01	2.085e-01	-1.010	0.3125
## as.factor(SMSA)4	-4.605e-01	6.309e-01	3.278e-01	-1.405	0.1600
## URBAN	1.143e+01	9.180e+04	1.125e+03	0.010	0.9919
## HTN_REP	1.046e-01	1.110e+00	1.407e-01	0.744	0.4572
## RBC	1.351e-01	1.145e+00	7.611e-02	1.776	0.0758 .
## DRMI	-3.575e-01	6.994e-01	3.875e-01	-0.922	0.3563
## DIAB	-2.536e-01	7.760e-01	3.007e-01	-0.843	0.3990

```
## ---
```

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

```
##
```

	exp(coef)	exp(-coef)	lower .95	upper .95
## BOOZE	1.014e+00	9.859e-01	0.9734	1.057
## AGEYRS	1.063e+00	9.407e-01	1.0479	1.078
## as.factor(RACE)2	7.037e-01	1.421e+00	0.4088	1.211
## as.factor(RACE)3	9.626e-01	1.039e+00	0.3029	3.059
## GRADES	1.037e+00	9.642e-01	0.9935	1.083
## as.factor(MARRY)3	1.199e+00	8.342e-01	0.8720	1.648
## as.factor(MARRY)4	1.335e+00	7.490e-01	0.8150	2.187
## as.factor(MARRY)5	1.108e+00	9.027e-01	0.3958	3.100

```
## as.factor(MARRY)6 5.824e-01 1.717e+00 0.2692 1.260
## as.factor(MARRY)8 1.080e+00 9.257e-01 0.1493 7.819
## as.factor(SIZE)2 1.059e+00 9.439e-01 0.6409 1.751
## as.factor(SIZE)3 1.151e+00 8.691e-01 0.6826 1.939
## as.factor(SIZE)4 9.868e-01 1.013e+00 0.5501 1.770
## as.factor(SIZE)5 6.759e-01 1.480e+00 0.2278 2.006
## as.factor(SIZE)6 1.798e+00 5.563e-01 0.8125 3.977
## as.factor(SIZE)7 8.608e-01 1.162e+00 0.3934 1.884
## as.factor(SIZE)8 1.534e-05 6.520e+04 0.0000 Inf
## AVGSMK 1.030e+00 9.713e-01 1.0196 1.040
## as.factor(SMSA)2 8.101e-01 1.234e+00 0.5384 1.219
## as.factor(SMSA)4 6.309e-01 1.585e+00 0.3319 1.199
## URBAN 9.180e+04 1.089e-05 0.0000 Inf
## HTN_REP 1.110e+00 9.006e-01 0.8426 1.463
## RBC 1.145e+00 8.736e-01 0.9861 1.329
## DRMI 6.994e-01 1.430e+00 0.3272 1.495
## DIAB 7.760e-01 1.289e+00 0.4304 1.399
##
## Concordance= 0.716 (se = 0.016 )
## Likelihood ratio test= 138.9 on 25 df, p=<2e-16
## Wald test = 123.3 on 25 df, p=6e-15
## Score (logrank) test = 131.1 on 25 df, p=2e-16

z <- (coef(fit6_cont_f)[1] - coef(fit6_cont_m)[1])/sqrt(vcov(fit6_cont_f)[1,1] + vcov(fit6_cont_m)[1,1])
1 - pnorm(z)

## BOOZE
## 0.7171777
```

Checking PH-assumption

```
# categorical version
fit3_q <- coxph(Surv(fu_time, cancer_death) ~ BOOZE_q + AGEYRS +
                SEX + as.factor(RACE) + GRADES + as.factor(SIZE) + as.factor(MARRY) + AVGSMK + as
                HTN_REP + RBC + DRMI + DIAB, data = dta)
cox.zph(fit3_q)

##          chisq df      p
## BOOZE_q    2.92259  3 0.404
## AGEYRS      0.99688  1 0.318
## SEX         3.50873  1 0.061
## as.factor(RACE) 0.06268  2 0.969
## GRADES      2.06122  1 0.151
## as.factor(SIZE) 3.88985  7 0.792
## as.factor(MARRY) 5.51469  5 0.356
## AVGSMK      0.00156  1 0.968
## as.factor(SMSA) 0.40469  2 0.817
## URBAN       0.27417  1 0.601
## HTN_REP     0.25422  1 0.614
## RBC         0.40691  1 0.524
## DRMI        0.00317  1 0.955
## DIAB        3.01364  1 0.083
## GLOBAL     24.54144 28 0.653
```

```
# continuous version
fit3_con <- coxph(Surv(fu_time, cancer_death) ~ BOOZE + AGEYRS +
                  SEX + as.factor(RACE) + GRADES + as.factor(SIZE) + as.factor(MARRY) + AVGSMK + as
                  HTN_REP + RBC + DRMI + DIAB, data = dta)
cox.zph(fit3_con)

##               chisq df      p
## BOOZE          5.34e-01  1 0.465
## AGEYRS          1.01e+00  1 0.315
## SEX             3.41e+00  1 0.065
## as.factor(RACE)  5.92e-02  2 0.971
## GRADES           2.05e+00  1 0.152
## as.factor(SIZE)  3.85e+00  7 0.797
## as.factor(MARRY) 5.46e+00  5 0.362
## AVGSMK           2.26e-05  1 0.996
## as.factor(SMSA)  4.17e-01  2 0.812
## URBAN            2.58e-01  1 0.611
## HTN_REP          2.58e-01  1 0.612
## RBC              3.77e-01  1 0.539
## DRMI             2.93e-03  1 0.957
## DIAB             3.05e+00  1 0.081
## GLOBAL           2.23e+01 26 0.672
```

Exploring Non-linearity

```
library(splines)
fit_nl <- coxph(Surv(fu_time, cancer_death) ~ ns(BOOZE, knots = 3) + AGEYRS +
                SEX + as.factor(RACE) + GRADES + as.factor(SIZE) + as.factor(MARRY) + AVGSMK + as
                HTN_REP + RBC + DRMI + DIAB, data = dta)
x <- seq(0,21, 0.5)
loghr <- ns(x, knots = 3) %*% coef(fit_nl)[c(1,2)]
vcov <- vcov(fit_nl)[c(1,2), c(1,2)]
var <- diag(ns(x, knots = 3) %*% vcov(fit_nl)[c(1,2), c(1,2)] %*% t(ns(x, knots = 3)))
se <- sqrt(var)
df <- tibble(x = x,
             y = loghr,
             ymin = loghr - 1.96*se,
             ymax = loghr + 1.96*se)
ggplot(df) +
  geom_line(aes(x = x, y = y), color = "red", linetype = "dashed")+
  geom_ribbon(aes(x = x, y = y, ymin = ymin, ymax = ymax),
            alpha = 0.5, fill = "#DB7093") +
  theme_bw() +
  labs(x = "Alcohol consumption (drinks/week)",
       y = "log(HR)")
```

