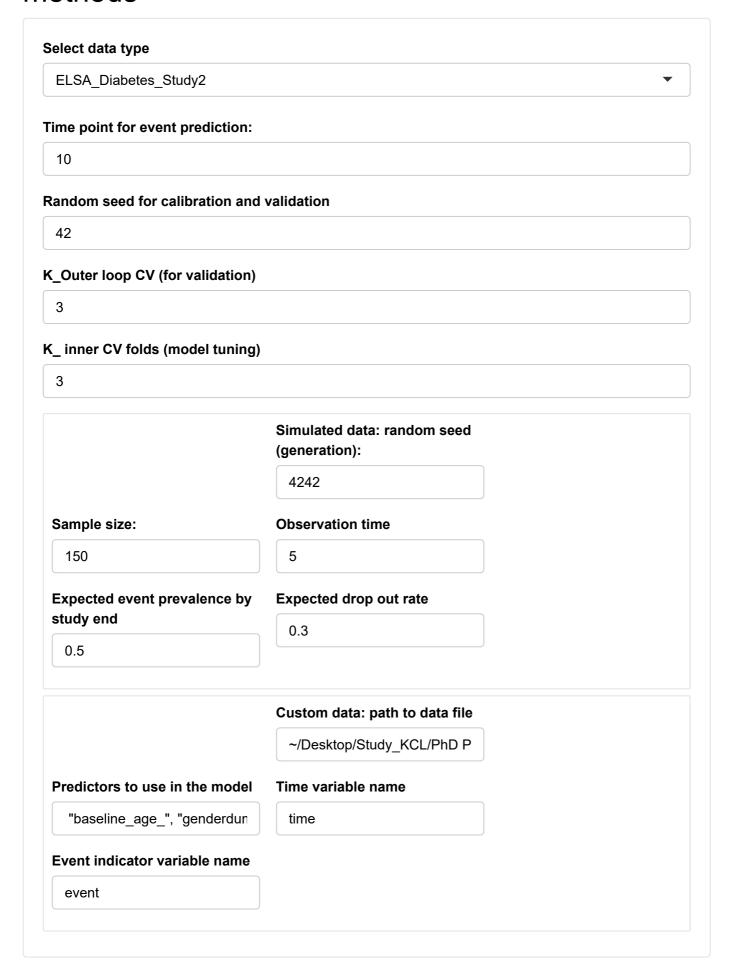
Simulated examples for the survival ensemble methods



Showing 1 to 3 of 3 entries

1

Next

Previous

SRF Ens1: CoxPH->SRF Ens2: CoxPH in clusters Sample statistics CoxPH Ens3: extended CoxPH Summary Conclusions Internally cross-validated results: Show 25 entries Search: BS 🛊 Calib_slope \$ AUCROC \$ BS_scaled \$ C_score \$ Calib_alpha T 🛊 test 0.7744 0.0775 0.0942 0.7442 1.2128 0.0401 10 train 0.9315 0.0631 0.262 0.8955 3.0708 0.0368 10 Showing 1 to 2 of 2 entries Previous 1 Next Internally cross-validated Test results for each CV fold: Show 25 ✓ entries Search: **AUCROC** \$ BS 🛊 BS_scaled \$ C_score \$ Calib_slope \$ Calib_alpha T 🛊 test.1 0.7663 0.0868 0.0948 0.7405 1.2223 0.0652 10 10 test.2 0.7611 0.0752 0.0824 0.7224 1.1853 0.033 test.3 0.7959 0.0707 0.1054 0.7699 1.2307 0.0221 10

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```
$test
  Τ
        AUCROC
                      BS BS_scaled
                                      C score
1 10 0.7662861 0.08675366 0.09483187 0.7404850
2 10 0.7610997 0.07517238 0.08243972 0.7223657
3 10 0.7959463 0.07070586 0.10541233 0.7698869
  Calib_slope Calib_alpha test cv_n
    1.222318 0.06523804
     1.185295 0.03300928
2
                            1
3
    1.230688 0.02211466
                            1
                                 3
$train
   Т
                      BS BS_scaled C_score Calib_slope
        AUCROC
1 10 0.9164321 0.0627577 0.2237111 0.8780920
                                               2.651160
2 10 0.9114885 0.0689563 0.2131885 0.8776219
                                               2.782932
3 10 0.9666902 0.0575690 0.3491984 0.9309040
                                               3.778457
  Calib_alpha test cv_n
1 0.03423430
2 0.04119281
                      2
                     3
3 0.03488901
                 0
$testaverage
          Т
                 AUCROC
                                BS
                                     BS scaled
10.00000000 0.77444405 0.07754397 0.09422797
    C_score Calib_slope Calib_alpha
                                          test
 0.74424587 1.21276679 0.04012066 1.00000000
$trainaverage
          Т
                 AUCROC
                                 BS
                                     BS_scaled
10.00000000 0.93153691 0.06309433 0.26203266
    C_score Calib_slope Calib_alpha
                                          test
 0.89553930 3.07084965 0.03677204 0.00000000
$model list
$model_list[[1]]
$model_list[[1]]$beststats
   mtry nodesize nodedepth time
                                  AUCROC
                                                  BS
V1
     3
                       50 8.9 0.7574443 0.06423064
    BS_scaled C_score Calib_alpha Calib_slope
V1 0.07790647 0.7316096 0.03298471
                                      1.167896
$model_list[[1]]$allstats
    mtry nodesize nodedepth time
                                   AUCROC
٧1
      5
               15
                        50 8.9 0.7524432 0.06409274
V2
       5
               20
                        50 8.9 0.7491188 0.06423795
٧3
       5
               25
                        50 8.9 0.7500564 0.06436698
٧4
       5
               30
                        50 8.9 0.7473205 0.06441021
۷5
       5
               35
                        50 8.9 0.7520207 0.06436109
       5
              40
۷6
                        50 8.9 0.7506937 0.06429571
V7
       5
              45
                        50 8.9 0.7555107 0.06418097
٧8
       5
              50
                        50 8.9 0.7526464 0.06418817
V11
       3
              45
                        50 8.9 0.7574443 0.06423064
V21
       5
              45
                        50 8.9 0.7555107 0.06418097
V31
      7
              45
                        50 8.9 0.7485881 0.06444108
              45
V41
                        50 8.9 0.7489870 0.06479384
      10
                        50 8.9 0.7411255 0.06515950
V51
              45
      15
```

```
BS_scaled C_score Calib_alpha Calib_slope
V1 0.07988610 0.7266474 0.02987083
                                      0.8803919
V2 0.07780145 0.7218242 0.03128357
                                      0.8786895
V3 0.07594906 0.7251048 0.03141545
                                      0.8944504
V4 0.07532852 0.7219599 0.03232438 0.9219095
V5 0.07603363 0.7265248 0.03220964
                                      0.9408165
V6 0.07697233 0.7254902 0.03293138 0.9659345
V7 0.07861951 0.7299378 0.03293988 0.9939605
V8 0.07851616 0.7271501 0.03299204 0.9924619
V11 0.07790647 0.7316096 0.03298471 1.1678961
V21 0.07861951 0.7299378 0.03293988 0.9939605
V31 0.07488540 0.7229955 0.03249506 0.8995296
V41 0.06982110 0.7236501 0.03264760
                                      0.8408969
V51 0.06457173 0.7149470 0.03252309 0.7497635
$model_list[[1]]$model
                        Sample size: 3978
                   Number of deaths: 305
                    Number of trees: 500
          Forest terminal node size: 45
      Average no. of terminal nodes: 58.024
No. of variables tried at each split: 3
             Total no. of variables: 23
       Resampling used to grow trees: swor
   Resample size used to grow trees: 2514
                           Analysis: RSF
                             Family: surv
                     Splitting rule: logrank *random*
       Number of random split points: 50
                         (00B) CRPS: 0.04245173
   (OOB) Requested performance error: 0.26577737
$model_list[[1]]$vimp10
   cox_predict
                      bmi_0_
                                    trig_0
   0.055099492
                 0.031270070
                                0.013059578
        hyp_0
                       t2dm
                                       pc2_
  0.006160379
                 0.005029568
                                0.003850401
   wealth low
                               EduLevel low
                baseline hdl
   0.002880856
                 0.002607893
                                0.002207414
         pc4_ B_smokstatus_0
                                       age_
  0.001883513
                 0.001560937
                                0.001526236
 exercise_vig
                     B_dep_0
                                   stroke 0
   0.001193894
                 0.001184870
                                0.001144235
$model list[[1]]$model base
Call:
coxph(formula = as.formula(paste("Surv(df_train$time, df_train$event) ~",
   paste(predict.factors, collapse = "+"))), data = df_train,
   x = TRUE
                   coef exp(coef) se(coef)
sz20
              -0.028345 0.972053 0.055940 -0.507
pc1
               0.019860 1.020059 0.057520 0.345
              -0.091728 0.912353 0.057566 -1.593
pc2
pc3_
              -0.030055 0.970392 0.060139 -0.500
```

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```
0.146215 1.157445 0.071871 2.034
age
              -0.369663 0.690967 0.131234 -2.817
sex
bmi_0_
               0.469355 1.598963 0.055908 8.395
hyp_0
               0.561660 1.753581 0.120607 4.657
cvd 0
              -0.028051 0.972339 0.168995 -0.166
B_dep_0
              0.379309 1.461275 0.150855 2.514
               0.102940 1.108425 0.035682 2.885
trig_0
baseline_hdl -0.299819 0.740952 0.196535 -1.526
stroke_0
               0.330850 1.392151 0.303594 1.090
B_smokstatus_0 0.378792 1.460519 0.151321 2.503
exercise_light 0.007339 1.007366 0.245089 0.030
exercise_vig -0.343196 0.709499 0.145865 -2.353
EduLevel_low
               0.366422 1.442564 0.220926 1.659
EduLevel_med
               0.178474 1.195392 0.210247 0.849
wealth_med
               0.085987 1.089792 0.159795 0.538
wealth low
               0.230550 1.259292 0.162348 1.420
t2dm
               0.319483 1.376417 0.059816 5.341
                     р
               0.61237
sz20_
pc1
               0.72988
               0.11106
pc2
pc3_
               0.61724
               0.49605
pc4_
age_
               0.04191
sex
               0.00485
bmi_0_
               < 2e-16
hyp_0
              3.21e-06
cvd 0
               0.86817
B_dep_0
               0.01192
trig_0
               0.00392
baseline_hdl
               0.12713
stroke_0
               0.27581
B_smokstatus_0 0.01231
exercise_light 0.97611
exercise_vig
               0.01863
EduLevel_low
               0.09720
EduLevel_med
               0.39595
wealth_med
               0.59050
wealth low
               0.15558
t2dm
              9.24e-08
Likelihood ratio test=257.7 on 22 df, p=< 2.2e-16
n= 3978, number of events= 305
$model list[[2]]
$model_list[[2]]$beststats
   mtry nodesize nodedepth time
                                  AUCROC
                                               BS
V1
                       50
                             9 0.7767595 0.0720635
     3
   BS scaled C score Calib alpha Calib slope
V1 0.09193106 0.7486539
                         0.0366775
$model list[[2]]$allstats
   mtry nodesize nodedepth time
                                   AUCROC
                                                 BS
٧1
              15
                        50
                              9 0.7666048 0.07197338
```

-0.039656 0.961120 0.058256 -0.681

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```
V2
       5
               20
                         50
                               9 0.7658599 0.07206223
V3
       5
               25
                         50
                               9 0.7683058 0.07192512
٧4
       5
               30
                         50
                               9 0.7684578 0.07202294
V5
       5
               35
                         50
                               9 0.7709751 0.07206851
V6
       5
               40
                         50
                               9 0.7704632 0.07188872
V7
       5
               45
                         50
                               9 0.7699974 0.07217563
٧8
       5
                         50
                               9 0.7725238 0.07190498
               50
V11
       3
               50
                         50
                               9 0.7767595 0.07206350
V21
       5
               50
                         50
                               9 0.7725238 0.07190498
V31
       7
               50
                         50
                              9 0.7663602 0.07230879
V41
                         50
                               9 0.7655590 0.07240470
      10
               50
V51
               50
                         50
                               9 0.7641787 0.07271219
      15
                C_score Calib_alpha Calib_slope
     BS_scaled
V1 0.09306665 0.7393086 0.03336042
                                       0.8960075
V2 0.09194706 0.7408122 0.03378378
                                       0.9455051
V3 0.09367479 0.7411962 0.03407004
                                       0.9853218
V4 0.09244218 0.7411933 0.03564269
                                      1.0008196
V5 0.09186795 0.7436081 0.03528510
                                      1.0327777
V6 0.09413353 0.7427847 0.03601608
                                       1.0560081
V7 0.09051817 0.7450198 0.03575406
                                      1.0713556
V8 0.09392858 0.7464099 0.03617636
                                       1.0998736
V11 0.09193106 0.7486539 0.03667750
                                      1.3093224
V21 0.09392858 0.7464099 0.03617636
                                       1.0998736
V31 0.08884024 0.7395762 0.03551413
                                       0.9842565
V41 0.08763164 0.7375523 0.03555250
                                       0.9118257
V51 0.08375706 0.7355275 0.03591630
                                       0.8406177
$model_list[[2]]$model
                         Sample size: 3979
                    Number of deaths: 336
                     Number of trees: 500
           Forest terminal node size: 50
       Average no. of terminal nodes: 52.166
No. of variables tried at each split: 3
              Total no. of variables: 23
       Resampling used to grow trees: swor
    Resample size used to grow trees: 2515
                            Analysis: RSF
```

Family: surv

Splitting rule: logrank *random*

Number of random split points: 50

(OOB) CRPS: 0.04534768

(OOB) Requested performance error: 0.25188944

\$model_list[[2]]\$vimp10

```
cox predict
                     bmi 0
                                     trig 0
0.0636945012
               0.0263490307
                               0.0171049348
       hyp_0
                                      t2dm
                       age_
0.0088539436
               0.0063876320
                               0.0056133904
baseline hdl
                   stroke 0
                                 wealth low
0.0046737470
               0.0018113059
                               0.0017268723
        pc1_
               EduLevel_low
                                       pc4_
0.0016132373
               0.0015386472
                               0.0012770149
                       pc2_ exercise_light
exercise vig
0.0011574051
               0.0008165957
                               0.0006028412
```

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```
$model list[[2]]$model base
coxph(formula = as.formula(paste("Surv(df_train$time, df_train$event) ~",
   paste(predict.factors, collapse = "+"))), data = df_train,
   x = TRUE
                 coef exp(coef) se(coef)
              0.05845
                        1.06019 0.05648 1.035
sz20_
pc1_
              0.02100
                       1.02122 0.05473 0.384
                        0.98684 0.05520 -0.240
pc2_
             -0.01324
              0.03302 1.03358 0.05921 0.558
pc3_
              0.03321 1.03376 0.05220 0.636
pc4_
age_
              sex
             -0.27238
                        0.76156 0.12402 -2.196
bmi_0_
             0.47992 1.61595 0.04985 9.628
hyp_0
              0.56260
                      1.75523 0.11439 4.918
cvd 0
             -0.04503 0.95597 0.15915 -0.283
B_dep_0
              0.32119
                        1.37877 0.14796 2.171
trig_0
              0.10805 1.11410 0.03174 3.404
baseline hdl -0.60433
                       0.54644 0.19215 -3.145
stroke 0
              0.56474 1.75900 0.25133 2.247
B_smokstatus_0 0.36836
                        1.44537 0.14856 2.480
exercise_light 0.02430
                        1.02459 0.23387 0.104
exercise vig -0.21397
                        0.80738 0.13363 -1.601
EduLevel_low 0.42431 1.52853 0.22276 1.905
EduLevel_med
              0.34439
                        1.41113 0.21110 1.631
wealth med
          -0.00133
                        0.99867 0.14948 -0.009
wealth low
              0.17185
                        1.18750 0.15294 1.124
              0.31351
                        1.36823 0.05610 5.589
t2dm_
sz20
             0.300705
pc1_
             0.701234
pc2_
             0.810360
             0.577049
pc3_
             0.524662
pc4_
             0.000362
age_
             0.028072
sex
bmi_0_
              < 2e-16
             8.73e-07
hyp_0
cvd_0
             0.777231
B_dep_0
             0.029943
trig 0
             0.000663
baseline_hdl
             0.001661
stroke_0
             0.024641
B smokstatus 0 0.013154
exercise light 0.917258
exercise vig
             0.109347
EduLevel low
             0.056809
EduLevel med
             0.102802
wealth med
             0.992903
wealth low
             0.261150
t2dm_
             2.28e-08
```

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Likelihood ratio test=310.2 on 22 df, p=< 2.2e-16

n= 3979, number of events= 336

```
$model list[[3]]
$model_list[[3]]$beststats
   mtry nodesize nodedepth time
                                  AUCROC
                                                 BS
V2
                       50
                             9 0.7410925 0.07398025
             20
    BS_scaled C_score Calib_alpha Calib_slope
V2 0.06943004 0.713259 0.03310082
$model_list[[3]]$allstats
    mtry nodesize nodedepth time
                                   AUCROC
٧1
      5
              15
                        50
                              9 0.7366901 0.07429957
V2
       5
              20
                        50
                              9 0.7410925 0.07398025
V3
       5
              25
                        50
                            9 0.7351740 0.07433120
V4
      5
              30
                        50
                              9 0.7342701 0.07434394
V5
       5
              35
                        50
                             9 0.7371426 0.07407061
۷6
      5
              40
                        50
                              9 0.7373241 0.07401269
      5
              45
V7
                        50
                            9 0.7355204 0.07417644
٧8
      5
              50
                        50
                              9 0.7362332 0.07414613
      3
              20
                        50
                            9 0.7360386 0.07394098
V11
V21
      5
              20
                        50
                              9 0.7410925 0.07398025
V31
      7
              20
                        50
                           9 0.7280921 0.07482518
V41
     10
              20
                        50
                            9 0.7261060 0.07517868
V51
     15
              20
                        50
                              9 0.7230698 0.07613958
     BS scaled C score Calib alpha Calib slope
V1 0.06541346 0.7104335 0.03245704
                                     0.8315036
V2 0.06943004 0.7132590 0.03310082 0.8649443
V3 0.06501550 0.7086842 0.03378657 0.8635963
V4 0.06485528 0.7086244 0.03429826 0.8854501
V5 0.06829343 0.7103689 0.03465459 0.9081545
V6 0.06902200 0.7130094 0.03513163 0.9337602
V7 0.06696220 0.7101800 0.03543247 0.9356905
V8 0.06734341 0.7103670 0.03576064 0.9589342
V11 0.06992399 0.7096533 0.03384507 0.9621704
V21 0.06943004 0.7132590 0.03310082 0.8649443
V31 0.05880193 0.7028878 0.03256426
                                      0.7591329
V41 0.05435536 0.7011319 0.03224581
                                      0.7023665
V51 0.04226864 0.6961665 0.03079759
                                      0.6366920
$model_list[[3]]$model
                        Sample size: 3979
                   Number of deaths: 345
                     Number of trees: 500
           Forest terminal node size: 20
       Average no. of terminal nodes: 107.028
No. of variables tried at each split: 5
             Total no. of variables: 23
       Resampling used to grow trees: swor
    Resample size used to grow trees: 2515
                           Analysis: RSF
                             Family: surv
                     Splitting rule: logrank *random*
       Number of random split points: 50
                         (OOB) CRPS: 0.04793599
```

(OOB) Requested performance error: 0.29073272

```
$model list[[3]]$vimp10
  cox predict
                      bmi 0
                                    trig 0
 0.0633668227
                0.0279939763
                              0.0125464277
                       t2dm_
         age_
                                      pc2_
                0.0037898724
 0.0078066203
                              0.0026323100
        hyp_0
                        pc4_
                              EduLevel_low
 0.0024262964
                0.0014300096
                              0.0010435545
      B_dep_0
               exercise_vig
                                  stroke_0
 0.0008701044
              0.0008426293
                              0.0007508497
exercise_light
                       sz20_
                                     cvd_0
 0.0004282929
                0.0002451051
                              0.0001607238
$model_list[[3]]$model_base
Call:
coxph(formula = as.formula(paste("Surv(df_train$time, df_train$event) ~",
   paste(predict.factors, collapse = "+"))), data = df_train,
   x = TRUE
                   coef exp(coef) se(coef)
               0.065039 1.067201 0.055387 1.174
sz20
pc1
               0.032623 1.033161 0.054653 0.597
              -0.089742 0.914167 0.053964 -1.663
pc2_
pc3_
               0.016462 1.016598 0.054782 0.300
              -0.042009 0.958861 0.053633 -0.783
pc4
               0.142562 1.153225 0.068243 2.089
age_
              -0.315088 0.729725 0.124046 -2.540
sex
bmi 0
              0.486410 1.626467 0.048283 10.074
hyp_0
               0.424529 1.528870 0.112253 3.782
cvd_0
              -0.006792 0.993231 0.161008 -0.042
              0.234661 1.264480 0.148876 1.576
B_dep_0
              0.117352 1.124515 0.040517 2.896
trig_0
baseline_hdl -0.433751 0.648073 0.189558 -2.288
               0.472919 1.604671 0.282917 1.672
stroke_0
B_smokstatus_0 0.387311 1.473014 0.141095 2.745
exercise light -0.129075 0.878908 0.240804 -0.536
exercise_vig -0.186913 0.829516 0.129140 -1.447
EduLevel_low
               0.391485 1.479176 0.214204 1.828
EduLevel med
               0.333466 1.395797 0.201573 1.654
wealth med
              -0.038004 0.962709 0.143598 -0.265
wealth_low
               0.097583 1.102503 0.148756 0.656
t2dm_
               0.294049 1.341849 0.056125 5.239
sz20
              0.240290
pc1
              0.550563
pc2
              0.096314
pc3
              0.763797
              0.433472
pc4
age_
              0.036703
              0.011082
sex
bmi 0
               < 2e-16
hyp 0
              0.000156
cvd_0
              0.966352
B dep 0
              0.114976
trig_0
              0.003775
baseline_hdl
              0.022124
```

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```
stroke_0 0.094607

B_smokstatus_0 0.006051

exercise_light 0.591947

exercise_vig 0.147794

EduLevel_low 0.067605

EduLevel_med 0.098063

wealth_med 0.791276

wealth_low 0.511828

t2dm_ 1.61e-07
```

Likelihood ratio test=257.4 on 22 df, p=< 2.2e-16 n= 3979, number of events= 345

\$time

Time difference of 4.668681 mins