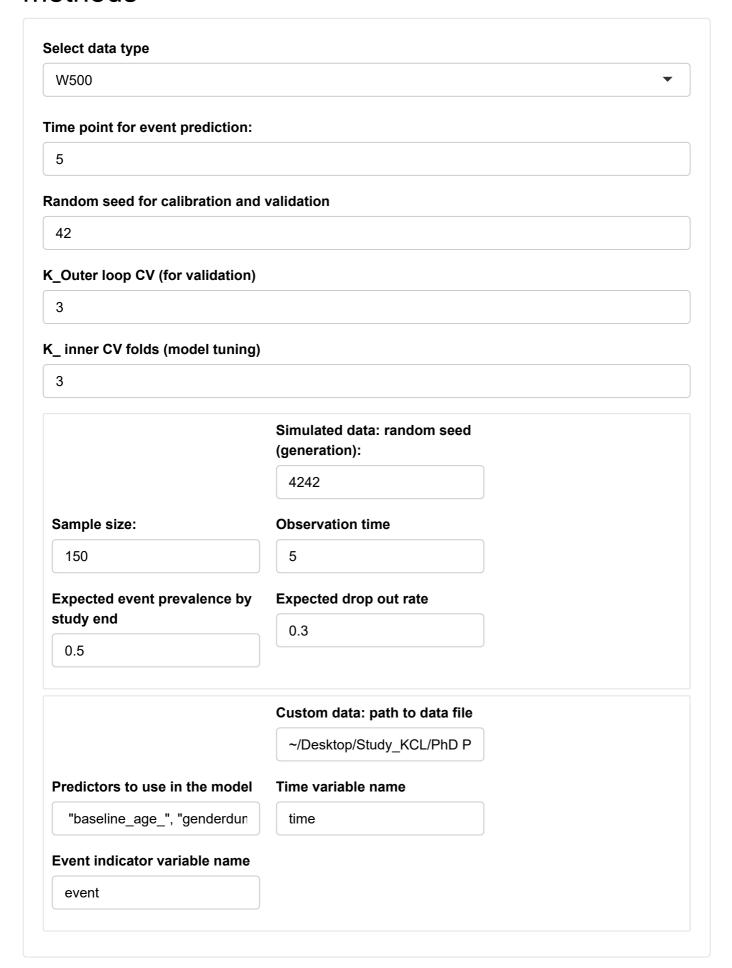
## Simulated examples for the survival ensemble methods



Showing 1 to 3 of 3 entries

Previous

1

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**SRF** Ens1: CoxPH->SRF Ens2: CoxPH in clusters Sample statistics CoxPH Ens3: extended CoxPH Summary Conclusions Internally cross-validated results: **Show** 10 entries Search: Calib\_slope \$ AUCROC \$ BS 🛊 BS\_scaled \$ C\_score \$ Calib\_alpha T 🛊 test 0.8373 0.2319 0.1816 0.7821 1.7767 0.0753 5 5 train 0.9039 0.1788 0.3731 0.8238 2.1936 0.0637 Showing 1 to 2 of 2 entries Previous 1 Next Internally cross-validated Test results for each CV fold: Show 10 ✓ entries Search: **AUCROC** \$ BS 🛊 BS\_scaled \$ C\_score \$ Calib\_slope \$ Calib\_alpha T 🛊 test.1 0.8723 0.3215 -0.0934 0.8065 2.3465 -0.0562 5 5 test.2 0.8136 0.1877 0.3262 0.7917 1.4341 0.1131 test.3 0.826 0.1865 0.312 0.7482 0.1691 5 1.5495

```
$test
 Τ
      AUCROC
                    BS BS_scaled
                                    C_score Calib_slope
1 5 0.8722601 0.3214817 -0.0933980 0.8064753
                                               2.346536
2 5 0.8135598 0.1877153 0.3261991 0.7917345
                                               1,434069
3 5 0.8260365 0.1864932 0.3119813 0.7482374
                                               1.549465
 Calib_alpha test cv_n
1 -0.0562467
   0.1130697
3 0.1691393 1
                     3
$train
 Τ
      AUCROC
                    BS BS_scaled C_score Calib_slope
1 5 0.8762396 0.2035782 0.2986905 0.8083743
2 5 0.9183884 0.1696934 0.3996468 0.8143104
                                              2.112169
3 5 0.9169287 0.1632106 0.4210484 0.8487737
                                              2.310226
 Calib_alpha test cv_n
1 0.05931673
2 0.06713708
                     2
                     3
3 0.06466375
                0
$testaverage
         Т
                AUCROC
                                BS
                                     BS scaled
 5.00000000 0.83728546 0.23189670 0.18159412
   C_score Calib_slope Calib_alpha
                                          test
 0.78214907 1.77668994 0.07532078 1.00000000
$trainaverage
         Τ
                AUCROC
                                BS
                                     BS_scaled
 5.00000000 0.90385225 0.17882740 0.37312860
   C_score Calib_slope Calib_alpha
                                          test
 0.82381948 2.19358982 0.06370585 0.00000000
$model list
$model_list[[1]]
$model_list[[1]]$beststats
   mtry nodesize nodedepth time
                                  AUCROC
V1
     3
             25
                       50 3.3 0.8042584 0.1854946
   BS_scaled C_score Calib_alpha Calib_slope
V1 0.2515693 0.747072 0.07469976
                                    1.414087
$model_list[[1]]$allstats
    mtry nodesize nodedepth time
                                   AUCROC
٧1
      4
              15
                        50 3.3 0.8036135 0.1827861
V2
      4
              20
                        50 3.3 0.8021343 0.1837010
V3
      4
              25
                        50 3.3 0.8050936 0.1828498
۷4
      4
              30
                        50 3.3 0.8040434 0.1838655
              25
                        50 3.3 0.8042584 0.1854946
V11
      3
V21
      5
              25
                        50 3.3 0.7956085 0.1858166
      7
V31
              25
                        50 3.3 0.7932647 0.1871167
V41
              25
                        50 3.3 0.7908765 0.1878931
     10
              25
                        50 3.3 0.7835970 0.1918826
V5
     15
   BS_scaled C_score Calib_alpha Calib_slope
V1 0.2624972 0.7478766 0.07073499
                                     1.1300125
V2 0.2588058 0.7493965 0.07188311
                                     1.1877798
```

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1.2789207

V3 0.2622401 0.7492177 0.07434891

```
1.3479644
V4 0.2581421 0.7472508 0.07595773
V11 0.2515693 0.7470720 0.07469976
                                     1.4140874
V21 0.2502700 0.7474892 0.06849913
                                     1.1332781
                                     1.0277256
V31 0.2450242 0.7460885 0.06692381
V41 0.2418916 0.7424527 0.06683811
                                     0.9110972
V5 0.2257949 0.7379228 0.06700300
                                     0.8196784
$model_list[[1]]$model
                        Sample size: 333
                   Number of deaths: 148
                    Number of trees: 500
          Forest terminal node size: 25
      Average no. of terminal nodes: 10.412
No. of variables tried at each split: 3
             Total no. of variables: 17
       Resampling used to grow trees: swor
   Resample size used to grow trees: 210
                           Analysis: RSF
                             Family: surv
                     Splitting rule: logrank *random*
       Number of random split points: 50
                         (OOB) CRPS: 0.17290317
   (OOB) Requested performance error: 0.25534572
$model_list[[1]]$vimp10
              cox_predict
                                     chf
                                                   bmi
         age
 5.269667e-02 5.000389e-02 1.545431e-02 1.022894e-02
          hr
                       los
                                     sho
                                                diasbp
 6.643624e-03 3.182544e-03 9.647674e-04 6.115573e-04
       gender
                     miord
                                     afb
                                                mitype
 3.909241e-04 3.673733e-04 5.937008e-05 5.737802e-05
       y1997
                                   y1999
                       cvd
 1.582724e-05 -1.293352e-04 -1.377054e-04
$model_list[[1]]$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)
                                         Z
       0.045127 1.046160 0.008373 5.389 7.07e-08
age
gender -0.235510 0.790168 0.177216 -1.329 0.18387
hr
       0.011451 1.011516 0.003917 2.923 0.00347
       0.003548 1.003555 0.003670 0.967 0.33358
sysbp
diasbp -0.013238 0.986849 0.006395 -2.070 0.03845
bmi
       -0.047169 0.953926 0.019425 -2.428 0.01517
cvd
      -0.025967 0.974367 0.219483 -0.118 0.90582
afb
       0.046888 1.048005 0.224161 0.209 0.83431
sho
       0.942797 2.567151 0.366438 2.573 0.01009
chf
       0.518465 1.679448 0.189360 2.738 0.00618
av3
       0.550735 1.734527 0.490364 1.123 0.26139
miord
       0.031701 1.032208 0.189273 0.167 0.86699
mitype -0.028336 0.972062 0.230706 -0.123 0.90225
los
       -0.023947 0.976338 0.024218 -0.989 0.32277
```

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```
y1997 -0.205430 0.814297 0.245965 -0.835 0.40360
y1999 -0.185491 0.830697 0.230437 -0.805 0.42085
Likelihood ratio test=135.4 on 16 df, p=< 2.2e-16
n= 333, number of events= 148
$model_list[[2]]
$model_list[[2]]$beststats
   mtry nodesize nodedepth time
                                  AUCROC
                                                BS
                       50 2.9 0.8485461 0.1693326
٧1
     3
             30
              C_score Calib_alpha Calib_slope
   BS_scaled
V1 0.3298849 0.7695282 0.03475621
                                      1.61813
$model_list[[2]]$allstats
   mtry nodesize nodedepth time
                                   AUCROC
٧1
              15
                        50 2.9 0.8470288 0.1645384
V2
              20
                        50 2.9 0.8463550 0.1658128
V3
      4
              25
                        50 2.9 0.8463719 0.1663581
              30
                        50 2.9 0.8476221 0.1672425
V4
V11
      3
              30
                        50 2.9 0.8485461 0.1693326
V21
      5
              30
                        50 2.9 0.8468629 0.1658770
      7
                        50 2.9 0.8442001 0.1647465
V31
              30
V41
     10
              30
                        50 2.9 0.8432322 0.1638070
                        50 2.9 0.8365888 0.1630288
V5
     15
              30
   BS_scaled C_score Calib_alpha Calib_slope
V1 0.3488577 0.7701725 0.02977282
                                     1.2259684
V2 0.3438143 0.7692060 0.03035507
                                     1.2738710
V3 0.3416565 0.7705825 0.03308607 1.3450685
V4 0.3381564 0.7715198 0.03512416 1.4405591
V11 0.3298849 0.7695282 0.03475621
                                     1.6181298
V21 0.3435602 0.7697332 0.03223757
                                     1.3411137
V31 0.3480343 0.7693524 0.03321698 1.1814584
V41 0.3517520 0.7656621 0.03162826 1.0670551
V5 0.3548316 0.7616202 0.03361119 0.9803958
$model_list[[2]]$model
                        Sample size: 334
                   Number of deaths: 146
                    Number of trees: 500
          Forest terminal node size: 30
      Average no. of terminal nodes: 8.546
No. of variables tried at each split: 3
             Total no. of variables: 17
       Resampling used to grow trees: swor
    Resample size used to grow trees: 211
                           Analysis: RSF
                             Family: surv
                     Splitting rule: logrank *random*
      Number of random split points: 50
                         (OOB) CRPS: 0.15583045
   (OOB) Requested performance error: 0.22954033
$model_list[[2]]$vimp10
 cox_predict
                       age
                                     chf
                                                   sho
```

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```
8.027918e-02 3.106505e-02 1.965666e-02 9.984943e-03
         bmi
                        hr
                                     afb
                                                 sysbp
 7.124829e-03 4.445464e-03 1.601209e-03 1.182673e-03
         los
                    diasbp
                                   y1997
                                                gender
 1.154096e-03 1.151968e-03 7.810680e-04 5.613834e-04
       y1999
                       cvd
                                  mitype
5.823523e-05 -1.111680e-04 -1.658726e-04
$model_list[[2]]$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)
                                            Z
age
       0.0427255 1.0436514 0.0082711 5.166 2.40e-07
gender -0.4636530 0.6289817 0.1845558 -2.512 0.01200
hr
       0.0105162 1.0105717 0.0040092 2.623 0.00871
sysbp -0.0016290 0.9983723 0.0036701 -0.444 0.65715
diasbp -0.0124902 0.9875874 0.0060900 -2.051 0.04027
       -0.0628271 0.9391058 0.0221249 -2.840 0.00452
bmi
cvd
       0.0550500 1.0565934 0.2185633 0.252 0.80114
      -0.0564619 0.9451025 0.2190806 -0.258 0.79662
afb
sho
      1.5720199 4.8163671 0.3514348 4.473 7.71e-06
       0.8723982 2.3926420 0.2024632 4.309 1.64e-05
chf
      -0.0124767 0.9876008 0.6267103 -0.020 0.98412
av3
miord -0.0127836 0.9872978 0.1875898 -0.068 0.94567
mitype -0.2589697 0.7718464 0.2389325 -1.084 0.27843
       0.0008633 1.0008637 0.0176832 0.049 0.96106
los
y1997 -0.6730080 0.5101717 0.2518453 -2.672 0.00753
     -0.3506881 0.7042033 0.2262112 -1.550 0.12108
y1999
Likelihood ratio test=172.6 on 16 df, p=< 2.2e-16
n= 334, number of events= 146
$model_list[[3]]
$model_list[[3]]$beststats
   mtry nodesize nodedepth time
                                 AUCROC
                                               BS
V1
                       50 2.2 0.838926 0.1551005
     3
             15
  BS scaled
              C_score Calib_alpha Calib_slope
V1 0.3236312 0.7741169 0.0307184
                                     1.328401
$model_list[[3]]$allstats
   mtry nodesize nodedepth time
                                   AUCROC
٧1
                        50 2.2 0.8378846 0.1534363
              15
V2
      4
              20
                        50 2.2 0.8357532 0.1542568
٧3
      4
              25
                        50 2.2 0.8357496 0.1553463
٧4
              30
                        50 2.2 0.8356515 0.1560454
V11
      3
              15
                        50 2.2 0.8389260 0.1551005
V21
      5
              15
                        50 2.2 0.8362526 0.1541165
V31
      7
              15
                        50 2.2 0.8346919 0.1552187
V41
     10
              15
                        50 2.2 0.8308972 0.1577378
V5
     15
              15
                        50 2.2 0.8215753 0.1628212
   BS scaled
              C_score Calib_alpha Calib_slope
V1 0.3308887 0.7733148 0.03260733
                                     1.2015905
```

```
V2 0.3273104 0.7714638 0.03589434
                                    1.2563099
V3 0.3225595 0.7712479 0.03674847
                                    1.3048902
V4 0.3195106 0.7707543 0.03668933
                                    1.3801161
V11 0.3236312 0.7741169 0.03071840
                                    1.3284013
V21 0.3279225 0.7727904 0.02938729
                                   1.1057494
V31 0.3231157 0.7749807 0.02703298
                                    1.0236751
V41 0.3121303 0.7722659 0.02596653
                                    0.9024273
V5 0.2899623 0.7666821 0.02500830
                                    0.8013229
$model_list[[3]]$model
                       Sample size: 333
                   Number of deaths: 136
                    Number of trees: 500
          Forest terminal node size: 15
      Average no. of terminal nodes: 16.426
No. of variables tried at each split: 3
             Total no. of variables: 17
      Resampling used to grow trees: swor
   Resample size used to grow trees: 210
                          Analysis: RSF
                            Family: surv
                    Splitting rule: logrank *random*
      Number of random split points: 50
                         (OOB) CRPS: 0.15684567
  (OOB) Requested performance error: 0.22876109
$model_list[[3]]$vimp10
 cox predict
                    age
                                 chf
0.0610714613 0.0404930999 0.0177301208 0.0092150952
                    los
                              gender
                                          diasbp
0.0052743281 \ 0.0043323313 \ 0.0018941710 \ 0.0014710194
        sho
                  y1997
                               sysbp
                                           miord
0.0013158923 0.0005718562 0.0003803418 0.0003494961
        cvd
                 mitype
0.0003476008 0.0002391834 0.0001877771
$model_list[[3]]$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)
                                          Ζ
age
       0.0537301 1.0551998 0.0095672 5.616 1.95e-08
gender -0.2346093  0.7908797  0.1822319 -1.287  0.197947
       0.0112430 1.0113064 0.0038730 2.903 0.003697
hr
       0.0012513 1.0012521 0.0036191 0.346 0.729524
sysbp
diasbp -0.0129867 0.9870973 0.0059917 -2.167 0.030202
bmi
      -0.0371948 0.9634884 0.0207787 -1.790 0.073447
cvd
      afb
       0.2506873 1.2849082 0.2343887 1.070 0.284828
sho
       1.4600417 4.3061390 0.3960059 3.687 0.000227
chf
       0.8340849 2.3027060 0.1887787 4.418 9.95e-06
av3
      miord
       0.1944781 1.2146769 0.1852333 1.050 0.293760
```

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```
mitype -0.4134881 0.6613394 0.2645408 -1.563 0.118043
los 0.0009399 1.0009404 0.0200538 0.047 0.962617
y1997 -0.4720247 0.6237381 0.2512653 -1.879 0.060300
y1999 -0.4348502 0.6473617 0.2340114 -1.858 0.063134
```

Likelihood ratio test=166.9 on 16 df, p=< 2.2e-16 n= 333, number of events= 136

## \$time

Time difference of 6.509842 secs