

Simulated examples for the survival ensemble methods

Select data type

ELSA_Diabetes_Study2

Time point for event prediction:

10

Random seed for calibration and validation

42

K_Outer loop CV (for validation)

3

K_inner CV folds (model tuning)

3

Simulated data: random seed (generation):

4242

Sample size:

150

Observation time

5

Expected event prevalence by study end

0.5

Expected drop out rate

0.3

Custom data: path to data file

~/Desktop/Study_KCL/PhD P

Predictors to use in the model

"baseline_age_", "genderdun

Time variable name

time

Event indicator variable name

event

Sample statistics

CoxPH

SRF

Ens1: CoxPH->SRF

Ens2: CoxPH in clusters

Ens3: extended CoxPH

Summary

Conclusions

Internally cross-validated results:

Show 25 ▾ entries

Search:

	AUCROC ▾	BS ▾	BS_scaled ▾	C_score ▾	Calib_slope ▾	Calib_alpha ▾	T ▾
test	0.7094	0.0861	-0.0074	0.683	0.3716	0.0455	10
train	0.8437	0.068	0.2058	0.8148	1.0203	0.0403	10

Showing 1 to 2 of 2 entries

Previous

1

Next

Internally cross-validated Test results for each CV fold:

Show 25 ▾ entries

Search:

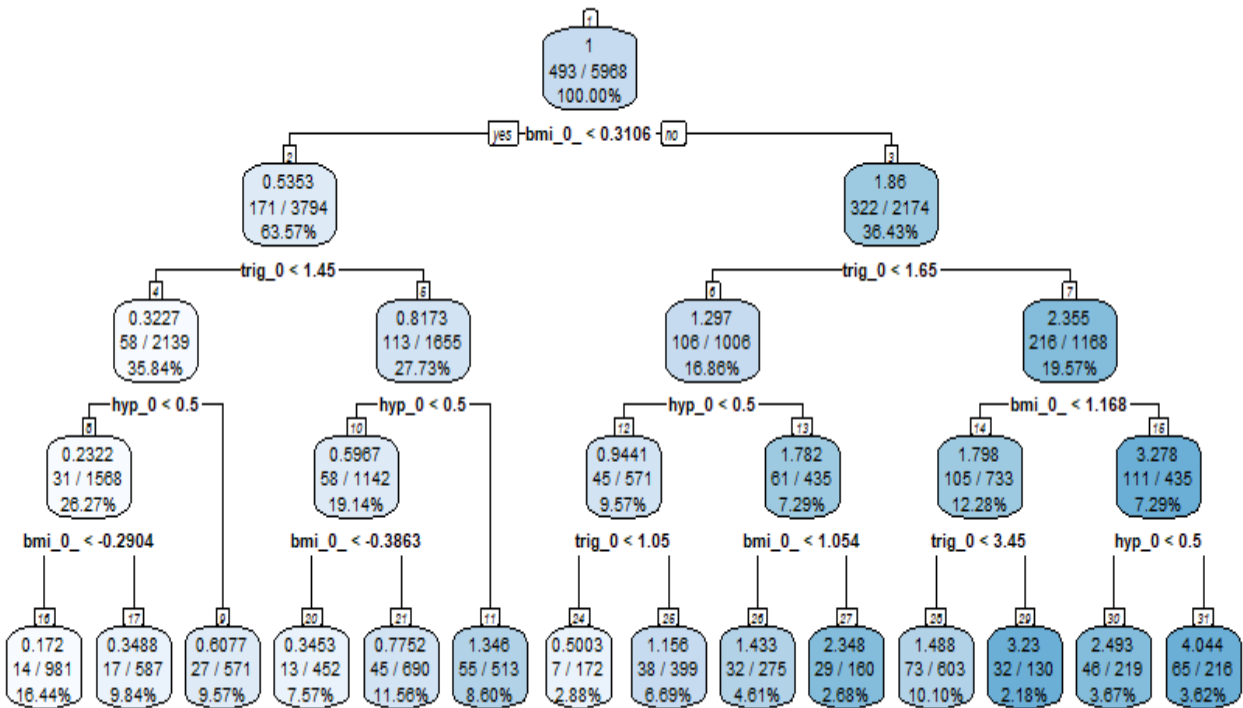
	AUCROC ▾	BS ▾	BS_scaled ▾	C_score ▾	Calib_slope ▾	Calib_alpha ▾	T ▾
test.1	0.7057	0.0941	0.0186	0.6762	0.4086	0.0776	10
test.2	0.7001	0.0868	-0.0599	0.6731	0.3609	0.0292	10
test.3	0.7222	0.0775	0.019	0.6995	0.3453	0.0296	10

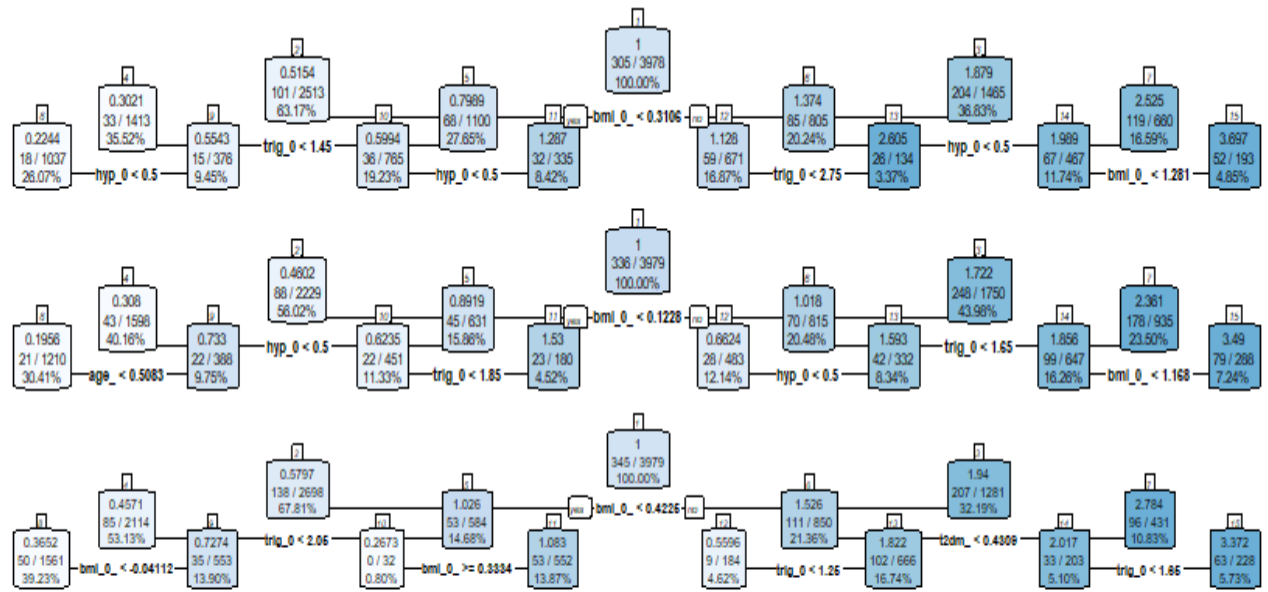
Showing 1 to 3 of 3 entries

Previous

1

Next





[[1]]

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
sz20_	-0.008053	0.991980	0.122374	-0.066
pc1_	0.097947	1.102905	0.118013	0.830
pc2_	-0.032187	0.968325	0.121585	-0.265
pc3_	0.118377	1.125668	0.133055	0.890
pc4_	-0.367736	0.692300	0.124556	-2.952
age_	0.111906	1.118408	0.157924	0.709
sex	-0.302930	0.738651	0.283025	-1.070
bmi_0_	-0.207931	0.812263	0.512554	-0.406
hyp_0	0.366644	1.442885	0.241810	1.516
cvd_0	0.364959	1.440456	0.343885	1.061
B_dep_0	0.223598	1.250569	0.352808	0.634
trig_0	0.087144	1.091053	0.266859	0.327
baseline_hdl	-0.151595	0.859336	0.456105	-0.332
stroke_0	-0.804855	0.447153	1.029423	-0.782
B_smokstatus_0	0.483367	1.621525	0.313213	1.543
exercise_light	-0.986854	0.372748	0.738864	-1.336
exercise_vig	0.144168	1.155079	0.271139	0.532
EduLevel_low	0.376750	1.457540	0.405381	0.929
EduLevel_med	-0.162636	0.849901	0.390869	-0.416
wealth_med	-0.010185	0.989867	0.318103	-0.032
wealth_low	0.265036	1.303478	0.328702	0.806
t2dm_	0.483269	1.621367	0.130849	3.693

	p
sz20_	0.947533
pc1_	0.406556
pc2_	0.791216
pc3_	0.373638
pc4_	0.003153
age_	0.478568
sex	0.284470
bmi_0_	0.684981
hyp_0	0.129455
cvd_0	0.288561
B_dep_0	0.526232
trig_0	0.744006
baseline_hdl	0.739611
stroke_0	0.434302
B_smokstatus_0	0.122770
exercise_light	0.181668
exercise_vig	0.594924
EduLevel_low	0.352697
EduLevel_med	0.677346
wealth_med	0.974458
wealth_low	0.420064
t2dm_	0.000221

Likelihood ratio test=42.16 on 22 df, p=0.005983

n= 603, number of events= 73

[[2]]

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
sz20_	-0.11744	0.88920	0.18543	-0.633
pc1_	0.09317	1.09765	0.19407	0.480
pc2_	-0.15839	0.85351	0.19657	-0.806
pc3_	0.01636	1.01650	0.19478	0.084
pc4_	0.02574	1.02608	0.17936	0.144
age_	0.13261	1.14181	0.21201	0.625
sex	-0.84105	0.43126	0.47910	-1.755
bmi_0_	-0.15453	0.85682	0.93985	-0.164
hyp_0	0.00000	1.00000	0.00000	NaN
cvd_0	0.14997	1.16180	0.45835	0.327
B_dep_0	0.47221	1.60354	0.51915	0.910
trig_0	-0.76505	0.46531	0.69015	-1.109
baseline_hdl	0.98531	2.67863	0.68584	1.437
stroke_0	0.70960	2.03317	0.56656	1.252
B_smokstatus_0	-0.48746	0.61419	0.70237	-0.694
exercise_light	0.29509	1.34325	0.75419	0.391
exercise_vig	-0.77803	0.45931	0.53698	-1.449
EduLevel_low	-0.29657	0.74336	0.87444	-0.339
EduLevel_med	0.09916	1.10424	0.79509	0.125
wealth_med	-0.24342	0.78394	0.53938	-0.451
wealth_low	0.26183	1.29931	0.53120	0.493
t2dm_	0.62546	1.86911	0.20492	3.052

	p
sz20_	0.52654
pc1_	0.63118
pc2_	0.42037
pc3_	0.93306
pc4_	0.88588
age_	0.53165
sex	0.07918
bmi_0_	0.86940
hyp_0	NaN
cvd_0	0.74352
B_dep_0	0.36304
trig_0	0.26763
baseline_hdl	0.15082
stroke_0	0.21040
B_smokstatus_0	0.48767
exercise_light	0.69560
exercise_vig	0.14736
EduLevel_low	0.73449
EduLevel_med	0.90075
wealth_med	0.65178
wealth_low	0.62208
t2dm_	0.00227

Likelihood ratio test=25.23 on 22 df, p=0.2863

n= 275, number of events= 32

[[3]]

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
sz20_	9.677e-01	2.632e+00	4.821e-01	2.007
pc1_	1.688e-01	1.184e+00	2.795e-01	0.604
pc2_	-1.880e-01	8.287e-01	3.353e-01	-0.561
pc3_	-1.400e-01	8.693e-01	2.575e-01	-0.544
pc4_	2.798e-01	1.323e+00	2.689e-01	1.041
age_	1.847e-02	1.019e+00	3.660e-01	0.050
sex	-3.625e-01	6.959e-01	6.486e-01	-0.559
bmi_0_	-1.066e+00	3.445e-01	8.482e-01	-1.256
hyp_0	0.000e+00	1.000e+00	0.000e+00	NaN
cvd_0	-6.725e-01	5.104e-01	1.120e+00	-0.600
B_dep_0	7.457e-01	2.108e+00	7.557e-01	0.987
trig_0	8.488e-02	1.089e+00	3.038e-01	0.279
baseline_hdl	-2.264e+00	1.040e-01	1.155e+00	-1.960
stroke_0	1.772e+00	5.885e+00	1.304e+00	1.359
B_smokstatus_0	-1.053e+00	3.491e-01	8.479e-01	-1.241
exercise_light	1.462e+00	4.316e+00	9.435e-01	1.550
exercise_vig	9.117e-01	2.488e+00	7.060e-01	1.291
EduLevel_low	1.879e+01	1.441e+08	5.651e+03	0.003
EduLevel_med	1.818e+01	7.837e+07	5.651e+03	0.003
wealth_med	1.271e-01	1.135e+00	9.222e-01	0.138
wealth_low	6.426e-01	1.901e+00	9.065e-01	0.709
t2dm_	1.852e-01	1.203e+00	3.145e-01	0.589

	p
sz20_	0.0447
pc1_	0.5459
pc2_	0.5751
pc3_	0.5865
pc4_	0.2980
age_	0.9598
sex	0.5762
bmi_0_	0.2090
hyp_0	NaN
cvd_0	0.5483
B_dep_0	0.3238
trig_0	0.7799
baseline_hdl	0.0500
stroke_0	0.1741
B_smokstatus_0	0.2145
exercise_light	0.1212
exercise_vig	0.1966
EduLevel_low	0.9973
EduLevel_med	0.9974
wealth_med	0.8904
wealth_low	0.4784
t2dm_	0.5560

Likelihood ratio test=28.01 on 22 df, p=0.1753

n= 452, number of events= 13

[[4]]

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
sz20_	-3.789e-04	9.996e-01	2.720e-01	-0.001
pc1_	2.973e-01	1.346e+00	2.703e-01	1.100
pc2_	1.306e-01	1.140e+00	2.767e-01	0.472
pc3_	1.780e-01	1.195e+00	3.111e-01	0.572
pc4_	1.614e-01	1.175e+00	2.338e-01	0.690
age_	8.306e-01	2.295e+00	3.081e-01	2.696
sex	-4.381e-01	6.452e-01	6.313e-01	-0.694
bmi_0_	-4.597e-01	6.315e-01	6.666e-01	-0.690
hyp_0	0.000e+00	1.000e+00	0.000e+00	NaN
cvd_0	-1.852e+01	9.066e-09	7.774e+03	-0.002
B_dep_0	-1.841e+01	1.016e-08	7.210e+03	-0.003
trig_0	-3.149e-01	7.298e-01	1.200e+00	-0.263
baseline_hdl	-9.260e-02	9.116e-01	6.964e-01	-0.133
stroke_0	-1.874e+01	7.295e-09	3.630e+04	-0.001
B_smokstatus_0	1.050e+00	2.857e+00	6.607e-01	1.589
exercise_light	-1.961e+01	3.032e-09	1.403e+04	-0.001
exercise_vig	-9.296e-01	3.947e-01	6.760e-01	-1.375
EduLevel_low	-4.299e-01	6.506e-01	9.714e-01	-0.443
EduLevel_med	1.173e-02	1.012e+00	8.562e-01	0.014
wealth_med	-6.187e-01	5.387e-01	8.392e-01	-0.737
wealth_low	8.242e-01	2.280e+00	6.407e-01	1.286
t2dm_	3.402e-01	1.405e+00	3.003e-01	1.133

	p
sz20_	0.99889
pc1_	0.27135
pc2_	0.63683
pc3_	0.56713
pc4_	0.49008
age_	0.00702
sex	0.48763
bmi_0_	0.49046
hyp_0	NaN
cvd_0	0.99810
B_dep_0	0.99796
trig_0	0.79293
baseline_hdl	0.89422
stroke_0	0.99959
B_smokstatus_0	0.11205
exercise_light	0.99888
exercise_vig	0.16912
EduLevel_low	0.65812
EduLevel_med	0.98907
wealth_med	0.46098
wealth_low	0.19832
t2dm_	0.25728

Likelihood ratio test=29.83 on 22 df, p=0.1227

n= 981, number of events= 14

[[5]]

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
sz20_	-1.860e-01	8.303e-01	1.847e-01	-1.007
pc1_	1.935e-01	1.213e+00	1.991e-01	0.972
pc2_	-4.800e-02	9.531e-01	1.874e-01	-0.256
pc3_	-5.137e-02	9.499e-01	2.158e-01	-0.238
pc4_	1.463e-01	1.158e+00	1.931e-01	0.758
age_	3.253e-01	1.385e+00	2.486e-01	1.309
sex	-1.559e+00	2.104e-01	5.320e-01	-2.930
bmi_0_	-1.884e-01	8.283e-01	4.875e-01	-0.387
hyp_0	0.000e+00	1.000e+00	0.000e+00	NaN
cvd_0	2.113e-01	1.235e+00	5.321e-01	0.397
B_dep_0	5.751e-01	1.777e+00	6.890e-01	0.835
trig_0	-1.634e-01	8.492e-01	8.739e-01	-0.187
baseline_hdl	-6.537e-01	5.201e-01	6.896e-01	-0.948
stroke_0	1.555e+00	4.736e+00	6.338e-01	2.454
B_smokstatus_0	-1.020e-01	9.031e-01	7.092e-01	-0.144
exercise_light	-1.622e+01	9.065e-08	3.674e+03	-0.004
exercise_vig	1.112e-02	1.011e+00	4.665e-01	0.024
EduLevel_low	8.845e-01	2.422e+00	7.468e-01	1.184
EduLevel_med	6.189e-01	1.857e+00	6.899e-01	0.897
wealth_med	-1.154e-01	8.910e-01	4.803e-01	-0.240
wealth_low	-1.254e+00	2.855e-01	6.301e-01	-1.990
t2dm_	4.320e-01	1.540e+00	2.425e-01	1.781

	p
sz20_	0.31408
pc1_	0.33122
pc2_	0.79786
pc3_	0.81184
pc4_	0.44870
age_	0.19065
sex	0.00339
bmi_0_	0.69908
hyp_0	NaN
cvd_0	0.69127
B_dep_0	0.40386
trig_0	0.85164
baseline_hdl	0.34316
stroke_0	0.01414
B_smokstatus_0	0.88569
exercise_light	0.99648
exercise_vig	0.98098
EduLevel_low	0.23627
EduLevel_med	0.36973
wealth_med	0.81009
wealth_low	0.04664
t2dm_	0.07487

Likelihood ratio test=35.54 on 22 df, p=0.03402
 n= 571, number of events= 27

[[6]]

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
sz20_	-0.011274	0.988789	0.139791	-0.081
pc1_	-0.154996	0.856419	0.141965	-1.092
pc2_	-0.099009	0.905734	0.135994	-0.728
pc3_	-0.074706	0.928016	0.162616	-0.459
pc4_	0.083769	1.087378	0.145118	0.577
age_	0.394060	1.482989	0.174110	2.263
sex	0.139304	1.149474	0.311587	0.447
bmi_0_	0.636972	1.890747	0.409179	1.557
hyp_0	0.000000	1.000000	0.000000	NaN
cvd_0	-0.013386	0.986703	0.372868	-0.036
B_dep_0	0.243176	1.275293	0.363770	0.668
trig_0	0.021343	1.021572	0.113353	0.188
baseline_hdl	-1.165005	0.311921	0.535653	-2.175
stroke_0	0.282131	1.325952	0.497262	0.567
B_smokstatus_0	0.869997	2.386904	0.331669	2.623
exercise_light	-1.090029	0.336207	1.036140	-1.052
exercise_vig	0.183608	1.201544	0.317898	0.578
EduLevel_low	0.511165	1.667233	0.777506	0.657
EduLevel_med	0.990390	2.692285	0.759421	1.304
wealth_med	0.009735	1.009782	0.402627	0.024
wealth_low	0.412730	1.510937	0.380883	1.084
t2dm_	0.153435	1.165832	0.134674	1.139

	p
sz20_	0.93572
pc1_	0.27493
pc2_	0.46659
pc3_	0.64594
pc4_	0.56377
age_	0.02362
sex	0.65482
bmi_0_	0.11954
hyp_0	NaN
cvd_0	0.97136
B_dep_0	0.50382
trig_0	0.85065
baseline_hdl	0.02964
stroke_0	0.57046
B_smokstatus_0	0.00871
exercise_light	0.29280
exercise_vig	0.56356
EduLevel_low	0.51090
EduLevel_med	0.19219
wealth_med	0.98071
wealth_low	0.27854
t2dm_	0.25458

Likelihood ratio test=40.12 on 22 df, p=0.01046

n= 513, number of events= 55

[[7]]

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
sz20_	0.096124	1.100895	0.166732	0.577
pc1_	0.295299	1.343527	0.146267	2.019
pc2_	-0.033456	0.967098	0.156191	-0.214
pc3_	-0.008621	0.991416	0.158802	-0.054
pc4_	0.065802	1.068015	0.136352	0.483
age_	0.273777	1.314921	0.172266	1.589
sex	0.462846	1.588588	0.350273	1.321
bmi_0_	0.411092	1.508464	0.745288	0.552
hyp_0	0.000000	1.000000	0.000000	NaN
cvd_0	0.058590	1.060341	0.443243	0.132
B_dep_0	0.310140	1.363616	0.419457	0.739
trig_0	0.157123	1.170140	0.123973	1.267
baseline_hdl	-0.123254	0.884039	0.538927	-0.229
stroke_0	1.179539	3.252873	1.157620	1.019
B_smokstatus_0	-0.110199	0.895655	0.418682	-0.263
exercise_light	0.294174	1.342017	0.649044	0.453
exercise_vig	-0.120314	0.886642	0.357205	-0.337
EduLevel_low	0.713879	2.041896	0.605558	1.179
EduLevel_med	0.237398	1.267945	0.575619	0.412
wealth_med	-0.304955	0.737156	0.395771	-0.771
wealth_low	0.014439	1.014544	0.392022	0.037
t2dm_	0.358255	1.430830	0.164680	2.175

	p
sz20_	0.5643
pc1_	0.0435
pc2_	0.8304
pc3_	0.9567
pc4_	0.6294
age_	0.1120
sex	0.1864
bmi_0_	0.5812
hyp_0	NaN
cvd_0	0.8948
B_dep_0	0.4597
trig_0	0.2050
baseline_hdl	0.8191
stroke_0	0.3082
B_smokstatus_0	0.7924
exercise_light	0.6504
exercise_vig	0.7363
EduLevel_low	0.2384
EduLevel_med	0.6800
wealth_med	0.4410
wealth_low	0.9706
t2dm_	0.0296

Likelihood ratio test=29.98 on 22 df, p=0.1188

n= 690, number of events= 45

[[8]]

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	p
sz20_	0.25237	1.28708	0.13378	1.886	0.0592
pc1_	-0.16601	0.84704	0.14299	-1.161	0.2456
pc2_	0.05098	1.05230	0.13211	0.386	0.6996
pc3_	-0.05414	0.94730	0.13964	-0.388	0.6982
pc4_	-0.16769	0.84562	0.14081	-1.191	0.2337
age_	0.34124	1.40669	0.17805	1.917	0.0553
sex	-0.32162	0.72497	0.32029	-1.004	0.3153
bmi_0_	-0.11446	0.89185	0.17755	-0.645	0.5191
hyp_0	0.00000	1.00000	0.00000	NaN	NaN
cvd_0	0.09546	1.10017	0.41575	0.230	0.8184
B_dep_0	-0.18880	0.82796	0.39266	-0.481	0.6306
trig_0	-0.11839	0.88835	0.11493	-1.030	0.3030
baseline_hdl	-0.37413	0.68789	0.60139	-0.622	0.5339
stroke_0	-0.33491	0.71540	1.10075	-0.304	0.7609
B_smokstatus_0	0.14017	1.15046	0.38681	0.362	0.7171
exercise_light	1.00185	2.72333	0.42555	2.354	0.0186
exercise_vig	-1.15760	0.31424	0.46233	-2.504	0.0123
EduLevel_low	0.44896	1.56669	0.56666	0.792	0.4282
EduLevel_med	0.70294	2.01968	0.52983	1.327	0.1846
wealth_med	-0.40747	0.66533	0.41063	-0.992	0.3211
wealth_low	0.23269	1.26199	0.38410	0.606	0.5446
t2dm_	0.40250	1.49555	0.13742	2.929	0.0034

Likelihood ratio test=45.4 on 22 df, p=0.002357

n= 216, number of events= 65

[[9]]

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
sz20_	4.702e-02	1.048e+00	2.498e-01	0.188
pc1_	-8.432e-02	9.191e-01	2.529e-01	-0.333
pc2_	-7.434e-02	9.284e-01	2.579e-01	-0.288
pc3_	-2.424e-01	7.847e-01	2.395e-01	-1.012
pc4_	2.398e-01	1.271e+00	2.322e-01	1.033
age_	5.530e-01	1.738e+00	3.340e-01	1.656
sex	-1.799e-01	8.353e-01	5.680e-01	-0.317
bmi_0_	-4.108e-01	6.631e-01	1.532e+00	-0.268
hyp_0	0.000e+00	1.000e+00	0.000e+00	NaN
cvd_0	-1.538e-01	8.575e-01	8.036e-01	-0.191
B_dep_0	8.204e-01	2.271e+00	6.990e-01	1.174
trig_0	1.084e+00	2.956e+00	1.133e+00	0.956
baseline_hdl	-1.383e-01	8.708e-01	7.383e-01	-0.187
stroke_0	-1.641e+01	7.469e-08	1.555e+04	-0.001
B_smokstatus_0	6.148e-01	1.849e+00	8.530e-01	0.721

exercise_light	-1.693e+01	4.428e-08	6.612e+03	-0.003
exercise_vig	2.466e-02	1.025e+00	5.245e-01	0.047
EduLevel_low	-3.607e-01	6.972e-01	9.823e-01	-0.367
EduLevel_med	4.232e-01	1.527e+00	7.852e-01	0.539
wealth_med	8.626e-01	2.369e+00	5.808e-01	1.485
wealth_low	-4.465e-02	9.563e-01	8.886e-01	-0.050
t2dm_	2.279e-02	1.023e+00	2.798e-01	0.081

p

sz20_	0.8507
pc1_	0.7389
pc2_	0.7732
pc3_	0.3113
pc4_	0.3017
age_	0.0977
sex	0.7514
bmi_0_	0.7886
hyp_0	NaN
cvd_0	0.8482
B_dep_0	0.2406
trig_0	0.3388
baseline_hdl	0.8514
stroke_0	0.9992
B_smokstatus_0	0.4710
exercise_light	0.9980
exercise_vig	0.9625
EduLevel_low	0.7134
EduLevel_med	0.5899
wealth_med	0.1375
wealth_low	0.9599
t2dm_	0.9351

Likelihood ratio test=15.27 on 22 df, p=0.8506

n= 587, number of events= 17

[[10]]

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
sz20_	-3.319e-01	7.175e-01	2.196e-01	-1.511
pc1_	-5.071e-01	6.023e-01	2.508e-01	-2.022
pc2_	6.083e-02	1.063e+00	2.051e-01	0.297
pc3_	-4.119e-02	9.596e-01	2.532e-01	-0.163
pc4_	-1.943e-02	9.808e-01	2.539e-01	-0.077
age_	-3.807e-01	6.833e-01	2.626e-01	-1.450
sex	-2.540e-02	9.749e-01	5.340e-01	-0.048
bmi_0_	-1.023e-01	9.028e-01	3.173e-01	-0.322
hyp_0	0.000e+00	1.000e+00	0.000e+00	NaN
cvd_0	-5.064e-01	6.027e-01	6.055e-01	-0.836
B_dep_0	2.128e-01	1.237e+00	5.312e-01	0.401
trig_0	9.078e-01	2.479e+00	9.471e-01	0.959
baseline_hdl	-3.108e-01	7.328e-01	7.379e-01	-0.421
stroke_0	6.948e-01	2.003e+00	8.714e-01	0.797
B_smokstatus_0	3.928e-01	1.481e+00	6.939e-01	0.566

exercise_light	-1.682e+01	4.978e-08	4.374e+03	-0.004
exercise_vig	-3.483e-01	7.059e-01	5.227e-01	-0.666
EduLevel_low	1.049e-01	1.111e+00	8.085e-01	0.130
EduLevel_med	-3.643e-01	6.947e-01	7.220e-01	-0.505
wealth_med	2.073e-01	1.230e+00	6.006e-01	0.345
wealth_low	-3.621e-01	6.962e-01	6.727e-01	-0.538
t2dm_	-2.696e-02	9.734e-01	2.087e-01	-0.129

p

sz20_	0.1307
pc1_	0.0432
pc2_	0.7668
pc3_	0.8708
pc4_	0.9390
age_	0.1471
sex	0.9621
bmi_0_	0.7472
hyp_0	NaN
cvd_0	0.4030
B_dep_0	0.6887
trig_0	0.3378
baseline_hdl	0.6736
stroke_0	0.4253
B_smokstatus_0	0.5713
exercise_light	0.9969
exercise_vig	0.5053
EduLevel_low	0.8968
EduLevel_med	0.6138
wealth_med	0.7300
wealth_low	0.5903
t2dm_	0.8973

Likelihood ratio test=22.5 on 22 df, p=0.4305

n= 160, number of events= 29

[[11]]

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
sz20_	-2.966e-01	7.434e-01	4.059e-01	-0.731
pc1_	-1.145e-01	8.918e-01	5.838e-01	-0.196
pc2_	-4.399e-01	6.441e-01	5.193e-01	-0.847
pc3_	2.028e-01	1.225e+00	4.951e-01	0.410
pc4_	-3.105e-01	7.331e-01	5.542e-01	-0.560
age_	3.448e-01	1.412e+00	5.622e-01	0.613
sex	8.622e-01	2.368e+00	1.043e+00	0.827
bmi_0_	-1.162e+00	3.127e-01	1.221e+00	-0.952
hyp_0	0.000e+00	1.000e+00	0.000e+00	NaN
cvd_0	-1.894e+01	5.921e-09	1.253e+04	-0.002
B_dep_0	1.400e+00	4.057e+00	1.108e+00	1.264
trig_0	2.921e+00	1.856e+01	4.057e+00	0.720
baseline_hdl	-2.512e+00	8.111e-02	1.526e+00	-1.646
stroke_0	-1.768e+01	2.105e-08	3.618e+04	0.000
B_smokstatus_0	-1.830e+01	1.126e-08	1.329e+04	-0.001

exercise_light	-1.757e+01	2.335e-08	2.266e+04	-0.001
exercise_vig	-1.597e+00	2.025e-01	1.528e+00	-1.045
EduLevel_low	1.577e+00	4.842e+00	1.688e+00	0.934
EduLevel_med	9.949e-01	2.704e+00	1.575e+00	0.632
wealth_med	-1.835e+00	1.596e-01	1.204e+00	-1.525
wealth_low	-1.012e+00	3.637e-01	1.439e+00	-0.703
t2dm_	6.480e-01	1.912e+00	4.224e-01	1.534

p

sz20_	0.4650
pc1_	0.8445
pc2_	0.3969
pc3_	0.6820
pc4_	0.5753
age_	0.5396
sex	0.4084
bmi_0_	0.3410
hyp_0	NaN
cvd_0	0.9988
B_dep_0	0.2061
trig_0	0.4715
baseline_hdl	0.0997
stroke_0	0.9996
B_smokstatus_0	0.9989
exercise_light	0.9994
exercise_vig	0.2959
EduLevel_low	0.3501
EduLevel_med	0.5277
wealth_med	0.1273
wealth_low	0.4822
t2dm_	0.1250

Likelihood ratio test=16.83 on 22 df, p=0.7727

n= 172, number of events= 7

[[12]]

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
sz20_	1.521e-01	1.164e+00	1.741e-01	0.874
pc1_	-1.252e-01	8.824e-01	1.802e-01	-0.694
pc2_	1.210e-02	1.012e+00	1.813e-01	0.067
pc3_	4.433e-02	1.045e+00	1.760e-01	0.252
pc4_	4.455e-02	1.046e+00	1.687e-01	0.264
age_	-3.912e-01	6.763e-01	2.528e-01	-1.547
sex	-8.856e-01	4.125e-01	3.638e-01	-2.434
bmi_0_	2.874e-01	1.333e+00	2.213e-01	1.298
hyp_0	0.000e+00	1.000e+00	0.000e+00	NaN
cvd_0	-1.319e-01	8.764e-01	6.537e-01	-0.202
B_dep_0	-4.993e-01	6.070e-01	4.385e-01	-1.139
trig_0	-1.074e-01	8.981e-01	1.489e-01	-0.722
baseline_hdl	-1.278e-01	8.801e-01	6.195e-01	-0.206
stroke_0	-1.309e+01	2.068e-06	3.056e+03	-0.004
B_smokstatus_0	7.334e-01	2.082e+00	3.724e-01	1.970

exercise_light	8.549e-02	1.089e+00	6.779e-01	0.126
exercise_vig	-1.020e-01	9.031e-01	3.868e-01	-0.264
EduLevel_low	3.854e-01	1.470e+00	6.762e-01	0.570
EduLevel_med	1.802e-01	1.197e+00	6.702e-01	0.269
wealth_med	6.886e-01	1.991e+00	5.203e-01	1.324
wealth_low	9.602e-01	2.612e+00	5.127e-01	1.873
t2dm_	2.250e-01	1.252e+00	1.551e-01	1.451

p

sz20_	0.3822
pc1_	0.4874
pc2_	0.9468
pc3_	0.8012
pc4_	0.7917
age_	0.1218
sex	0.0149
bmi_0_	0.1941
hyp_0	NaN
cvd_0	0.8401
B_dep_0	0.2549
trig_0	0.4705
baseline_hdl	0.8366
stroke_0	0.9966
B_smokstatus_0	0.0489
exercise_light	0.8996
exercise_vig	0.7921
EduLevel_low	0.5688
EduLevel_med	0.7880
wealth_med	0.1856
wealth_low	0.0611
t2dm_	0.1469

Likelihood ratio test=23 on 22 df, p=0.4017

n= 219, number of events= 46

[[13]]

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	p
sz20_	-0.01943	0.98076	0.15994	-0.121	0.9033
pc1_	-0.05875	0.94294	0.17609	-0.334	0.7387
pc2_	-0.27982	0.75592	0.17717	-1.579	0.1143
pc3_	-0.21255	0.80852	0.17419	-1.220	0.2224
pc4_	-0.23437	0.79107	0.18160	-1.291	0.1969
age_	0.23693	1.26735	0.22030	1.075	0.2822
sex	-1.17824	0.30782	0.40706	-2.895	0.0038
bmi_0_	0.49890	1.64692	0.19724	2.529	0.0114
hyp_0	0.00000	1.00000	0.00000	NaN	NaN
cvd_0	0.52644	1.69290	0.48003	1.097	0.2728
B_dep_0	0.47664	1.61066	0.46335	1.029	0.3036
trig_0	1.28808	3.62581	1.04208	1.236	0.2164
baseline_hdl	0.69436	2.00243	0.64934	1.069	0.2849
stroke_0	1.70850	5.52067	1.10355	1.548	0.1216
B_smokstatus_0	0.83424	2.30306	0.44714	1.866	0.0621

exercise_light	0.32573	1.38504	0.61617	0.529	0.5971
exercise_vig	0.26684	1.30584	0.40304	0.662	0.5079
EduLevel_low	0.16464	1.17897	0.62291	0.264	0.7915
EduLevel_med	-0.32681	0.72122	0.59882	-0.546	0.5852
wealth_med	0.55843	1.74793	0.44308	1.260	0.2075
wealth_low	-0.09509	0.90929	0.48871	-0.195	0.8457
t2dm_	0.45515	1.57641	0.19461	2.339	0.0193

Likelihood ratio test=35.22 on 22 df, p=0.03677

n= 399, number of events= 38

[[14]]

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
sz20_	1.998e-01	1.221e+00	2.501e-01	0.799
pc1_	-2.734e-01	7.608e-01	2.428e-01	-1.126
pc2_	-5.477e-01	5.783e-01	2.303e-01	-2.378
pc3_	-9.252e-02	9.116e-01	2.839e-01	-0.326
pc4_	-3.480e-01	7.061e-01	2.821e-01	-1.234
age_	3.004e-01	1.350e+00	2.942e-01	1.021
sex	2.577e-01	1.294e+00	4.843e-01	0.532
bmi_0_	1.773e+00	5.891e+00	9.619e-01	1.844
hyp_0	1.634e-01	1.178e+00	4.413e-01	0.370
cvd_0	3.132e-01	1.368e+00	6.440e-01	0.486
B_dep_0	1.227e+00	3.411e+00	5.785e-01	2.121
trig_0	-2.054e-01	8.143e-01	1.991e-01	-1.032
baseline_hdl	3.308e+00	2.732e+01	1.218e+00	2.717
stroke_0	-1.670e+01	5.566e-08	6.048e+03	-0.003
B_smokstatus_0	2.988e-01	1.348e+00	4.869e-01	0.614
exercise_light	-2.510e+00	8.124e-02	1.376e+00	-1.824
exercise_vig	-1.146e+00	3.181e-01	5.828e-01	-1.966
EduLevel_low	-7.482e-01	4.732e-01	7.573e-01	-0.988
EduLevel_med	-6.955e-01	4.988e-01	7.624e-01	-0.912
wealth_med	5.283e-02	1.054e+00	6.188e-01	0.085
wealth_low	3.721e-01	1.451e+00	5.056e-01	0.736
t2dm_	-8.291e-03	9.917e-01	1.858e-01	-0.045

	p
sz20_	0.42438
pc1_	0.26017
pc2_	0.01740
pc3_	0.74447
pc4_	0.21726
age_	0.30719
sex	0.59464
bmi_0_	0.06523
hyp_0	0.71119
cvd_0	0.62677
B_dep_0	0.03391
trig_0	0.30221
baseline_hdl	0.00659
stroke_0	0.99780
B_smokstatus_0	0.53947


```
exercise_light 0.06816
exercise_vig   0.04934
EduLevel_low   0.32311
EduLevel_med   0.36160
wealth_med     0.93196
wealth_low     0.46180
t2dm_          0.96442
```

Likelihood ratio test=33.61 on 22 df, p=0.05385

n= 130, number of events= 32

```

$test
  T      AUCROC      BS  BS_scaled  C_score
1 10 0.7057479 0.09405853 0.01861455 0.6762464
2 10 0.7000949 0.08683577 -0.05992456 0.6731115
3 10 0.7222338 0.07753214 0.01904457 0.6995087
  Calib_slope Calib_alpha test cv_n
1 0.4085616 0.07763810 1 1
2 0.3608876 0.02919129 1 2
3 0.3453140 0.02963581 1 3

$train
  T      AUCROC      BS  BS_scaled  C_score
1 10 0.8468496 0.06364370 0.2127516 0.8176759
2 10 0.8539843 0.06850565 0.2183306 0.8305529
3 10 0.8301162 0.07197724 0.1863173 0.7963035
  Calib_slope Calib_alpha test cv_n
1 1.009545 0.03732963 0 1
2 1.029468 0.03998930 0 2
3 1.021937 0.04347909 0 3

$testaverage
  T      AUCROC      BS  BS_scaled
10.00000000 0.70935886 0.08614214 -0.00742181
  C_score Calib_slope Calib_alpha test
0.68295553 0.37158773 0.04548840 1.00000000

$trainaverage
  T      AUCROC      BS  BS_scaled
10.00000000 0.8436500 0.0680422 0.2057998
  C_score Calib_slope Calib_alpha test
0.8148441 1.0203165 0.0402660 0.0000000

$model_list
$model_list[[1]]
$model_list[[1]]$vimp10
  bmi_0_      trig_0      hyp_0
0.0462592804 0.0087652941 0.0081749051
  pc2_      t2dm_      wealth_low
0.0072019955 0.0061080080 0.0046154275
  sex B_smokstatus_0 baseline_hdl
0.0031276687 0.0017127709 0.0015332970
  pc4_      stroke_0  EduLevel_med
0.0013104278 0.0011652414 0.0004816331
  exercise_vig      age_      wealth_med
0.0004398451 0.0004344877 -0.0002367985

$model_list[[1]]$treemodel
n= 3978

node), split, n, deviance, yval
  * denotes terminal node

1) root 3978 2090.3410 1.0000000
  2) bmi_0_ < 0.310594 2513 856.6503 0.5153783
    4) trig_0 < 1.45 1413 300.4899 0.3021157

```

```

8) hyp_0< 0.5 1037 165.9046 0.2243909 *
9) hyp_0>=0.5 376 128.1333 0.5542855 *
5) trig_0>=1.45 1100 532.6327 0.7988587
10) hyp_0< 0.5 765 287.7574 0.5994389 *
11) hyp_0>=0.5 335 234.8774 1.2866080 *
3) bmi_0_>=0.310594 1465 1109.4060 1.8792320
6) hyp_0< 0.5 805 512.3770 1.3744330
12) trig_0< 2.75 671 360.8545 1.1276720 *
13) trig_0>=2.75 134 138.9062 2.6050880 *
7) hyp_0>=0.5 660 577.8229 2.5254730
14) bmi_0_< 1.281423 467 360.0770 1.9894790 *
15) bmi_0_>=1.281423 193 205.7574 3.6969770 *

```

```
$model_list[[1]]$coxmodels
```

```
$model_list[[1]]$coxmodels[[1]]
```

```
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
sz20_	0.01694	1.01709	0.12861	0.132
pc1_	0.05151	1.05286	0.12331	0.418
pc2_	-0.05923	0.94249	0.13566	-0.437
pc3_	0.09622	1.10100	0.13271	0.725
pc4_	0.01002	1.01007	0.12821	0.078
age_	-0.03781	0.96290	0.15157	-0.249
sex	-0.56318	0.56939	0.29296	-1.922
bmi_0_	-0.14681	0.86345	0.48370	-0.304
hyp_0	0.00000	1.00000	0.00000	NaN
cvd_0	-0.10429	0.90097	0.34296	-0.304
B_dep_0	0.62920	1.87611	0.32134	1.958
trig_0	0.14639	1.15765	0.08089	1.810
baseline_hdl	0.56199	1.75416	0.45862	1.225
stroke_0	0.44076	1.55389	0.48485	0.909
B_smokstatus_0	-0.44291	0.64216	0.42678	-1.038
exercise_light	-0.20743	0.81267	0.61445	-0.338
exercise_vig	-0.33957	0.71208	0.31817	-1.067
EduLevel_low	0.10573	1.11152	0.45805	0.231
EduLevel_med	0.07523	1.07814	0.43621	0.172
wealth_med	0.06387	1.06595	0.38080	0.168
wealth_low	0.61167	1.84350	0.35471	1.724
t2dm_	0.42399	1.52804	0.13335	3.180
p				
sz20_	0.89519			
pc1_	0.67614			
pc2_	0.66240			
pc3_	0.46844			
pc4_	0.93771			
age_	0.80300			
sex	0.05455			
bmi_0_	0.76149			
hyp_0	NaN			
cvd_0	0.76106			
B_dep_0	0.05023			
trig_0	0.07033			

```
baseline_hdl 0.22042
stroke_0     0.36332
B_smokstatus_0 0.29936
exercise_light 0.73568
exercise_vig 0.28585
EduLevel_low 0.81745
EduLevel_med 0.86307
wealth_med   0.86680
wealth_low   0.08464
t2dm_        0.00148
```

Likelihood ratio test=34.17 on 22 df, p=0.04718
n= 467, number of events= 67

```
$model_list[[1]]$coxmodels[[2]]
```

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	p
sz20_	0.20026	1.22172	0.21063	0.951	0.3417
pc1_	0.40662	1.50173	0.16289	2.496	0.0125
pc2_	-0.23100	0.79374	0.17825	-1.296	0.1950
pc3_	-0.02248	0.97777	0.16605	-0.135	0.8923
pc4_	0.22598	1.25355	0.14984	1.508	0.1315
age_	0.29917	1.34873	0.20058	1.492	0.1358
sex	0.81086	2.24983	0.39759	2.039	0.0414
bmi_0_	0.46241	1.58790	0.40995	1.128	0.2593
hyp_0	0.00000	1.00000	0.00000	NaN	NaN
cvd_0	-0.92612	0.39609	0.65556	-1.413	0.1577
B_dep_0	0.32772	1.38780	0.45005	0.728	0.4665
trig_0	0.16789	1.18280	0.18275	0.919	0.3583
baseline_hdl	-1.01935	0.36083	0.62785	-1.624	0.1045
stroke_0	1.23950	3.45388	0.81157	1.527	0.1267
B_smokstatus_0	-0.32286	0.72408	0.46533	-0.694	0.4878
exercise_light	1.52228	4.58265	0.59273	2.568	0.0102
exercise_vig	-0.03926	0.96150	0.40190	-0.098	0.9222
EduLevel_low	1.91969	6.81883	1.07861	1.780	0.0751
EduLevel_med	1.66621	5.29209	1.04728	1.591	0.1116
wealth_med	-0.41288	0.66174	0.45692	-0.904	0.3662
wealth_low	-0.14671	0.86354	0.46099	-0.318	0.7503
t2dm_	0.34775	1.41588	0.19050	1.825	0.0679

Likelihood ratio test=46.12 on 22 df, p=0.001907
n= 765, number of events= 36

```
$model_list[[1]]$coxmodels[[3]]
```

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
sz20_	-1.767e-01	8.381e-01	2.840e-01	-0.622
pc1_	1.362e-01	1.146e+00	2.841e-01	0.480

pc2_	-3.989e-01	6.710e-01	2.608e-01	-1.530
pc3_	-2.958e-01	7.439e-01	2.923e-01	-1.012
pc4_	-8.918e-02	9.147e-01	3.032e-01	-0.294
age_	5.980e-01	1.818e+00	3.693e-01	1.619
sex	-2.088e+00	1.239e-01	7.892e-01	-2.646
bmi_0_	-5.236e-01	5.924e-01	6.293e-01	-0.832
hyp_0	0.000e+00	1.000e+00	0.000e+00	NaN
cvd_0	7.070e-01	2.028e+00	7.237e-01	0.977
B_dep_0	1.212e+00	3.360e+00	9.119e-01	1.329
trig_0	8.545e-01	2.350e+00	1.188e+00	0.719
baseline_hdl	1.271e-01	1.136e+00	9.505e-01	0.134
stroke_0	5.487e-01	1.731e+00	1.174e+00	0.467
B_smokstatus_0	-1.427e-01	8.670e-01	8.504e-01	-0.168
exercise_light	-1.836e+01	1.064e-08	7.928e+03	-0.002
exercise_vig	-3.684e-01	6.918e-01	6.450e-01	-0.571
EduLevel_low	3.682e-01	1.445e+00	1.238e+00	0.297
EduLevel_med	-1.614e-01	8.510e-01	1.194e+00	-0.135
wealth_med	1.628e+00	5.095e+00	8.053e-01	2.022
wealth_low	-1.284e-01	8.795e-01	9.303e-01	-0.138
t2dm_	3.885e-01	1.475e+00	2.975e-01	1.306

p

sz20_	0.53385
pc1_	0.63157
pc2_	0.12610
pc3_	0.31149
pc4_	0.76863
age_	0.10534
sex	0.00814
bmi_0_	0.40535
hyp_0	NaN
cvd_0	0.32863
B_dep_0	0.18386
trig_0	0.47195
baseline_hdl	0.89362
stroke_0	0.64033
B_smokstatus_0	0.86675
exercise_light	0.99815
exercise_vig	0.56789
EduLevel_low	0.76616
EduLevel_med	0.89251
wealth_med	0.04317
wealth_low	0.89022
t2dm_	0.19154

Likelihood ratio test=26.31 on 22 df, p=0.2385

n= 376, number of events= 15

\$model_list[[1]]\$coxmodels[[4]]

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
sz20_	-7.145e-02	9.310e-01	1.857e-01	-0.385
pc1_	-1.758e-01	8.388e-01	2.038e-01	-0.862

pc2_	2.972e-04	1.000e+00	2.001e-01	0.001
pc3_	-2.326e-01	7.925e-01	2.000e-01	-1.163
pc4_	-1.239e-01	8.835e-01	2.239e-01	-0.553
age_	4.026e-01	1.496e+00	2.340e-01	1.721
sex	-3.832e-01	6.817e-01	4.261e-01	-0.899
bmi_0_	5.297e-01	1.698e+00	5.510e-01	0.961
hyp_0	0.000e+00	1.000e+00	0.000e+00	NaN
cvd_0	8.003e-01	2.226e+00	4.396e-01	1.820
B_dep_0	-1.312e-01	8.770e-01	5.439e-01	-0.241
trig_0	2.168e-02	1.022e+00	1.389e-01	0.156
baseline_hdl	-1.599e+00	2.020e-01	7.399e-01	-2.161
stroke_0	-1.951e-01	8.227e-01	6.814e-01	-0.286
B_smokstatus_0	1.079e+00	2.941e+00	4.201e-01	2.568
exercise_light	-1.675e+01	5.314e-08	3.501e+03	-0.005
exercise_vig	-1.705e-01	8.432e-01	4.697e-01	-0.363
EduLevel_low	8.075e-01	2.242e+00	1.100e+00	0.734
EduLevel_med	6.890e-01	1.992e+00	1.092e+00	0.631
wealth_med	1.477e-01	1.159e+00	6.517e-01	0.227
wealth_low	8.294e-01	2.292e+00	5.648e-01	1.469
t2dm_	1.913e-01	1.211e+00	1.763e-01	1.085

p

sz20_	0.7004
pc1_	0.3885
pc2_	0.9988
pc3_	0.2449
pc4_	0.5800
age_	0.0853
sex	0.3685
bmi_0_	0.3364
hyp_0	NaN
cvd_0	0.0687
B_dep_0	0.8093
trig_0	0.8759
baseline_hdl	0.0307
stroke_0	0.7746
B_smokstatus_0	0.0102
exercise_light	0.9962
exercise_vig	0.7166
EduLevel_low	0.4628
EduLevel_med	0.5281
wealth_med	0.8207
wealth_low	0.1420
t2dm_	0.2780

Likelihood ratio test=45.8 on 22 df, p=0.002095

n= 335, number of events= 32

\$model_list[[1]]\$coxmodels[[5]]

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	p
sz20_	-0.11388	0.89237	0.13017	-0.875	0.3817
pc1_	-0.14244	0.86724	0.17318	-0.822	0.4108

pc2_	0.06210	1.06407	0.14022	0.443	0.6579
pc3_	-0.04159	0.95927	0.15869	-0.262	0.7933
pc4_	-0.28688	0.75060	0.18255	-1.572	0.1161
age_	0.37663	1.45737	0.20228	1.862	0.0626
sex	-0.42539	0.65352	0.36550	-1.164	0.2445
bmi_0_	-0.24551	0.78231	0.24883	-0.987	0.3238
hyp_0	0.00000	1.00000	0.00000	NaN	NaN
cvd_0	-0.88819	0.41140	0.52752	-1.684	0.0922
B_dep_0	-0.23209	0.79287	0.41348	-0.561	0.5746
trig_0	-0.16571	0.84729	0.14097	-1.176	0.2398
baseline_hdl	-0.67294	0.51021	0.58267	-1.155	0.2481
stroke_0	0.65948	1.93379	1.14804	0.574	0.5657
B_smokstatus_0	0.89645	2.45088	0.40616	2.207	0.0273
exercise_light	0.99777	2.71223	0.51668	1.931	0.0535
exercise_vig	-0.79215	0.45287	0.46907	-1.689	0.0913
EduLevel_low	0.06125	1.06317	0.58687	0.104	0.9169
EduLevel_med	0.25773	1.29399	0.56990	0.452	0.6511
wealth_med	-0.37272	0.68886	0.43414	-0.859	0.3906
wealth_low	-0.36278	0.69574	0.43729	-0.830	0.4068
t2dm_	0.25631	1.29216	0.15205	1.686	0.0918

Likelihood ratio test=30.97 on 22 df, p=0.09679

n= 193, number of events= 52

\$model_list[[1]]\$coxmodels[[6]]

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
sz20_	-1.588e-01	8.531e-01	2.275e-01	-0.698
pc1_	-1.382e-01	8.709e-01	2.488e-01	-0.556
pc2_	2.139e-01	1.239e+00	2.378e-01	0.900
pc3_	-2.794e-01	7.562e-01	2.589e-01	-1.079
pc4_	1.176e-01	1.125e+00	2.342e-01	0.502
age_	4.371e-01	1.548e+00	3.110e-01	1.405
sex	-1.264e+00	2.825e-01	5.918e-01	-2.136
bmi_0_	5.923e-01	1.808e+00	5.891e-01	1.006
hyp_0	0.000e+00	1.000e+00	0.000e+00	NaN
cvd_0	-8.450e-02	9.190e-01	8.020e-01	-0.105
B_dep_0	4.617e-01	1.587e+00	8.297e-01	0.556
trig_0	7.952e-01	2.215e+00	1.047e+00	0.759
baseline_hdl	7.463e-01	2.109e+00	6.535e-01	1.142
stroke_0	-1.547e+01	1.912e-07	1.878e+04	-0.001
B_smokstatus_0	1.367e+00	3.925e+00	6.523e-01	2.096
exercise_light	-1.692e+01	4.490e-08	6.167e+03	-0.003
exercise_vig	-4.906e-02	9.521e-01	4.904e-01	-0.100
EduLevel_low	-1.372e+00	2.536e-01	1.172e+00	-1.171
EduLevel_med	4.104e-01	1.507e+00	6.039e-01	0.680
wealth_med	-1.136e-01	8.926e-01	5.476e-01	-0.208
wealth_low	-5.347e-01	5.858e-01	8.087e-01	-0.661
t2dm_	2.614e-01	1.299e+00	2.622e-01	0.997
p				
sz20_	0.4850			
pc1_	0.5784			

```
pc2_      0.3684
pc3_      0.2804
pc4_      0.6156
age_      0.1599
sex       0.0327
bmi_0_    0.3147
hyp_0     NaN
cvd_0     0.9161
B_dep_0   0.5779
trig_0    0.4476
baseline_hdl 0.2535
stroke_0  0.9993
B_smokstatus_0 0.0361
exercise_light 0.9978
exercise_vig 0.9203
EduLevel_low 0.2416
EduLevel_med 0.4968
wealth_med 0.8356
wealth_low 0.5085
t2dm_     0.3189
```

```
Likelihood ratio test=23.42 on 22 df, p=0.3785
n= 1037, number of events= 18
```

```
$model_list[[1]]$coxmodels[[7]]
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
sz20_	-1.424e-01	8.673e-01	1.235e-01	-1.153
pc1_	-1.329e-01	8.755e-01	1.461e-01	-0.910
pc2_	-2.276e-01	7.964e-01	1.416e-01	-1.608
pc3_	-9.059e-02	9.134e-01	1.349e-01	-0.671
pc4_	1.547e-02	1.016e+00	1.439e-01	0.107
age_	-2.019e-01	8.172e-01	1.900e-01	-1.062
sex	-5.984e-01	5.497e-01	3.096e-01	-1.933
bmi_0_	4.207e-01	1.523e+00	1.747e-01	2.408
hyp_0	0.000e+00	1.000e+00	0.000e+00	NaN
cvd_0	2.818e-01	1.326e+00	4.211e-01	0.669
B_dep_0	4.539e-01	1.574e+00	3.338e-01	1.360
trig_0	2.375e-01	1.268e+00	2.677e-01	0.887
baseline_hdl	3.741e-01	1.454e+00	4.537e-01	0.825
stroke_0	-1.609e+01	1.029e-07	3.531e+03	-0.005
B_smokstatus_0	5.482e-01	1.730e+00	3.412e-01	1.607
exercise_light	-3.412e-01	7.109e-01	5.774e-01	-0.591
exercise_vig	-4.333e-01	6.483e-01	3.443e-01	-1.258
EduLevel_low	-1.393e-01	8.699e-01	4.440e-01	-0.314
EduLevel_med	-7.398e-01	4.772e-01	4.373e-01	-1.692
wealth_med	3.502e-01	1.419e+00	3.744e-01	0.935
wealth_low	4.520e-01	1.571e+00	3.928e-01	1.151
t2dm_	5.685e-02	1.059e+00	1.368e-01	0.416
p				
sz20_	0.2488			
pc1_	0.3630			


```

pc2_      0.1079
pc3_      0.5020
pc4_      0.9144
age_      0.2880
sex       0.0532
bmi_0_    0.0160
hyp_0     NaN
cvd_0     0.5034
B_dep_0   0.1739
trig_0    0.3750
baseline_hdl 0.4096
stroke_0  0.9964
B_smokstatus_0 0.1081
exercise_light 0.5546
exercise_vig 0.2082
EduLevel_low 0.7536
EduLevel_med 0.0907
wealth_med 0.3496
wealth_low 0.2499
t2dm_     0.6777

```

Likelihood ratio test=34.7 on 22 df, p=0.04168
n= 671, number of events= 59

```
$model_list[[1]]$coxmodels[[8]]
```

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
sz20_	0.179222	1.196286	0.273881	0.654
pc1_	-0.034531	0.966058	0.257754	-0.134
pc2_	-0.348431	0.705794	0.227214	-1.533
pc3_	0.249628	1.283548	0.287760	0.867
pc4_	-0.851719	0.426681	0.330806	-2.575
age_	-0.686242	0.503465	0.374861	-1.831
sex	-0.406378	0.666058	0.561094	-0.724
bmi_0_	-0.102032	0.903001	0.332109	-0.307
hyp_0	0.000000	1.000000	0.000000	NaN
cvd_0	0.542073	1.719569	1.165347	0.465
B_dep_0	-0.371556	0.689660	0.639685	-0.581
trig_0	-0.044057	0.956900	0.278789	-0.158
baseline_hdl	0.134906	1.144430	1.007948	0.134
stroke_0	0.000000	1.000000	0.000000	NaN
B_smokstatus_0	-0.424924	0.653820	0.672920	-0.631
exercise_light	1.308874	3.702002	0.923701	1.417
exercise_vig	0.000657	1.000657	0.527044	0.001
EduLevel_low	1.709021	5.523552	0.926717	1.844
EduLevel_med	0.642069	1.900409	0.867916	0.740
wealth_med	-0.121170	0.885884	0.607929	-0.199
wealth_low	-0.079401	0.923669	0.640941	-0.124
t2dm_	0.554183	1.740519	0.245206	2.260
p				
sz20_	0.5129			
pc1_	0.8934			

```

pc2_          0.1252
pc3_          0.3857
pc4_          0.0100
age_          0.0672
sex           0.4689
bmi_0_        0.7587
hyp_0         NaN
cvd_0         0.6418
B_dep_0       0.5613
trig_0        0.8744
baseline_hdl  0.8935
stroke_0      NaN
B_smokstatus_0 0.5277
exercise_light 0.1565
exercise_vig  0.9990
EduLevel_low  0.0652
EduLevel_med  0.4594
wealth_med     0.8420
wealth_low     0.9014
t2dm_         0.0238

```

Likelihood ratio test=25.88 on 22 df, p=0.257
n= 134, number of events= 26

```
$model_list[[1]]$clusters
```

```

[1] 1.989479 0.599439 0.554286 1.286608 3.696977 0.224391
[7] 1.127672 2.605088

```

```
$model_list[[2]]
```

```
$model_list[[2]]$vimp10
```

bmi_0_	trig_0	hyp_0
0.0447364618	0.0151361566	0.0117574361
age_	baseline_hdl	t2dm_
0.0100098220	0.0053352617	0.0050058942
EduLevel_low	pc4_	stroke_0
0.0034224723	0.0022185604	0.0021664515
sex	exercise_vig	pc2_
0.0015662161	0.0006213739	0.0005771305
exercise_light	sz20_	EduLevel_med
0.0004001569	0.0003480480	0.0002163010

```
$model_list[[2]]$treemodel
```

```
n= 3979
```

```

node), split, n, deviance, yval
* denotes terminal node

```

- 1) root 3979 2236.1680 1.0000000
- 2) bmi_0_ < 0.1228089 2229 743.9276 0.4601545
 - 4) hyp_0 < 0.5 1598 398.4751 0.3080419
 - 8) age_ < 0.5083334 1210 207.3218 0.1956276 *
 - 9) age_ >= 0.5083334 388 172.8389 0.7329705 *
 - 5) hyp_0 >= 0.5 631 320.9262 0.8918953
 - 10) trig_0 < 1.85 451 162.3999 0.6234802 *

```

11) trig_0>=1.85 180 148.9302 1.5300410 *
3) bmi_0_>=0.1228089 1750 1358.8730 1.7221500
6) trig_0< 1.65 815 460.3069 1.0175820
12) hyp_0< 0.5 483 222.2847 0.6623664 *
13) hyp_0>=0.5 332 223.9358 1.5931330 *
7) trig_0>=1.65 935 858.5827 2.3610590
14) bmi_0_< 1.167619 647 529.2193 1.8556470 *
15) bmi_0_>=1.167619 288 311.3518 3.4898110 *

```

```
$model_list[[2]]$coxmodels
```

```
$model_list[[2]]$coxmodels[[1]]
```

```
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
sz20_	0.15059	1.16252	0.12400	1.215
pc1_	0.15848	1.17173	0.10005	1.584
pc2_	-0.02463	0.97567	0.10415	-0.237
pc3_	0.10016	1.10535	0.11756	0.852
pc4_	-0.09411	0.91018	0.09636	-0.977
age_	0.10273	1.10819	0.12758	0.805
sex	-0.19871	0.81979	0.23140	-0.859
bmi_0_	0.09784	1.10279	0.34634	0.282
hyp_0	0.55125	1.73543	0.20688	2.665
cvd_0	0.40296	1.49624	0.28918	1.393
B_dep_0	0.35869	1.43145	0.28212	1.271
trig_0	0.15369	1.16613	0.08270	1.858
baseline_hdl	0.16842	1.18343	0.37462	0.450
stroke_0	-0.22277	0.80030	0.61914	-0.360
B_smokstatus_0	0.31922	1.37606	0.27261	1.171
exercise_light	-0.46655	0.62716	0.48594	-0.960
exercise_vig	-0.06475	0.93731	0.24379	-0.266
EduLevel_low	0.43551	1.54575	0.36264	1.201
EduLevel_med	-0.13792	0.87117	0.35512	-0.388
wealth_med	-0.05209	0.94925	0.27267	-0.191
wealth_low	0.09561	1.10033	0.27136	0.352
t2dm_	0.17344	1.18939	0.11043	1.571

p

sz20_	0.22455
pc1_	0.11319
pc2_	0.81304
pc3_	0.39421
pc4_	0.32872
age_	0.42067
sex	0.39047
bmi_0_	0.77756
hyp_0	0.00771
cvd_0	0.16348
B_dep_0	0.20358
trig_0	0.06313
baseline_hdl	0.65302
stroke_0	0.71899
B_smokstatus_0	0.24159
exercise_light	0.33700

```

exercise_vig    0.79056
EduLevel_low    0.22978
EduLevel_med    0.69774
wealth_med      0.84850
wealth_low      0.72460
t2dm_           0.11627

```

Likelihood ratio test=35.95 on 22 df, p=0.03078

n= 647, number of events= 99

```
$model_list[[2]]$coxmodels[[2]]
```

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
sz20_	-1.116e-01	8.944e-01	2.387e-01	-0.468
pc1_	1.615e-01	1.175e+00	2.256e-01	0.716
pc2_	1.590e-02	1.016e+00	2.206e-01	0.072
pc3_	-4.436e-01	6.417e-01	2.079e-01	-2.134
pc4_	7.008e-02	1.073e+00	2.066e-01	0.339
age_	2.901e-01	1.337e+00	4.584e-01	0.633
sex	-1.151e-01	8.913e-01	5.060e-01	-0.227
bmi_0_	2.166e+00	8.724e+00	8.149e-01	2.658
hyp_0	0.000e+00	1.000e+00	0.000e+00	NaN
cvd_0	2.407e-01	1.272e+00	7.684e-01	0.313
B_dep_0	5.207e-01	1.683e+00	6.243e-01	0.834
trig_0	4.296e-01	1.537e+00	1.479e-01	2.906
baseline_hdl	-1.099e+00	3.333e-01	7.564e-01	-1.453
stroke_0	-1.463e+01	4.413e-07	4.840e+03	-0.003
B_smokstatus_0	8.309e-01	2.295e+00	4.946e-01	1.680
exercise_light	9.870e-01	2.683e+00	8.379e-01	1.178
exercise_vig	-7.675e-01	4.642e-01	5.759e-01	-1.333
EduLevel_low	1.502e-02	1.015e+00	9.051e-01	0.017
EduLevel_med	4.672e-01	1.595e+00	7.893e-01	0.592
wealth_med	7.379e-01	2.092e+00	6.405e-01	1.152
wealth_low	1.151e+00	3.162e+00	6.734e-01	1.709
t2dm_	2.574e-01	1.294e+00	2.448e-01	1.052

p

sz20_	0.64013
pc1_	0.47414
pc2_	0.94253
pc3_	0.03283
pc4_	0.73445
age_	0.52688
sex	0.82009
bmi_0_	0.00786
hyp_0	NaN
cvd_0	0.75404
B_dep_0	0.40432
trig_0	0.00367
baseline_hdl	0.14634
stroke_0	0.99759
B_smokstatus_0	0.09295
exercise_light	0.23885

```

exercise_vig    0.18264
EduLevel_low    0.98676
EduLevel_med    0.55395
wealth_med      0.24927
wealth_low      0.08737
t2dm_           0.29290

```

Likelihood ratio test=49.22 on 22 df, p=0.0007465

n= 1210, number of events= 21

```
$model_list[[2]]$coxmodels[[3]]
```

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	p
sz20_	-0.09579	0.90866	0.16380	-0.585	0.5587
pc1_	-0.09406	0.91023	0.17004	-0.553	0.5801
pc2_	0.01446	1.01457	0.16408	0.088	0.9298
pc3_	0.26211	1.29966	0.17155	1.528	0.1265
pc4_	0.12679	1.13518	0.16625	0.763	0.4457
age_	-0.11587	0.89059	0.19466	-0.595	0.5517
sex	-0.19336	0.82418	0.39890	-0.485	0.6279
bmi_0_	0.21568	1.24071	0.23461	0.919	0.3579
hyp_0	0.00000	1.00000	0.00000	NaN	NaN
cvd_0	0.03008	1.03054	0.38657	0.078	0.9380
B_dep_0	0.39772	1.48843	0.42474	0.936	0.3491
trig_0	-0.04110	0.95974	0.59914	-0.069	0.9453
baseline_hdl	0.14173	1.15227	0.59035	0.240	0.8103
stroke_0	0.95290	2.59322	0.52826	1.804	0.0713
B_smokstatus_0	0.09305	1.09752	0.52842	0.176	0.8602
exercise_light	0.05938	1.06118	0.70453	0.084	0.9328
exercise_vig	-0.47726	0.62048	0.42587	-1.121	0.2624
EduLevel_low	1.36325	3.90888	1.06058	1.285	0.1987
EduLevel_med	1.57956	4.85283	1.03535	1.526	0.1271
wealth_med	-0.24672	0.78136	0.44140	-0.559	0.5762
wealth_low	-0.20142	0.81757	0.46127	-0.437	0.6624
t2dm_	0.22344	1.25037	0.16317	1.369	0.1709

Likelihood ratio test=23.83 on 22 df, p=0.3562

n= 332, number of events= 42

```
$model_list[[2]]$coxmodels[[4]]
```

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
sz20_	2.292e-01	1.258e+00	1.226e-01	1.869
pc1_	-1.993e-01	8.193e-01	1.299e-01	-1.535
pc2_	6.457e-02	1.067e+00	1.277e-01	0.506
pc3_	-3.797e-05	1.000e+00	1.388e-01	0.000
pc4_	-3.164e-02	9.689e-01	1.221e-01	-0.259
age_	1.791e-01	1.196e+00	1.612e-01	1.111

sex	-4.967e-01	6.085e-01	2.643e-01	-1.879
bmi_0_	3.311e-02	1.034e+00	1.518e-01	0.218
hyp_0	4.453e-01	1.561e+00	2.430e-01	1.832
cvd_0	1.031e-01	1.109e+00	3.779e-01	0.273
B_dep_0	-4.393e-01	6.445e-01	3.464e-01	-1.268
trig_0	-1.347e-01	8.740e-01	1.002e-01	-1.344
baseline_hdl	-2.069e-01	8.131e-01	4.650e-01	-0.445
stroke_0	1.816e-01	1.199e+00	1.080e+00	0.168
B_smokstatus_0	4.695e-01	1.599e+00	3.156e-01	1.488
exercise_light	7.137e-01	2.041e+00	4.413e-01	1.617
exercise_vig	-6.131e-01	5.417e-01	3.389e-01	-1.809
EduLevel_low	8.271e-01	2.287e+00	5.987e-01	1.382
EduLevel_med	7.311e-01	2.077e+00	5.732e-01	1.275
wealth_med	6.232e-02	1.064e+00	3.650e-01	0.171
wealth_low	3.292e-01	1.390e+00	3.603e-01	0.914
t2dm_	4.578e-01	1.581e+00	1.248e-01	3.669

p

sz20_	0.061615
pc1_	0.124865
pc2_	0.613080
pc3_	0.999782
pc4_	0.795494
age_	0.266585
sex	0.060228
bmi_0_	0.827411
hyp_0	0.066891
cvd_0	0.785102
B_dep_0	0.204752
trig_0	0.179093
baseline_hdl	0.656348
stroke_0	0.866473
B_smokstatus_0	0.136807
exercise_light	0.105820
exercise_vig	0.070430
EduLevel_low	0.167120
EduLevel_med	0.202179
wealth_med	0.864404
wealth_low	0.360871
t2dm_	0.000243

Likelihood ratio test=48.7 on 22 df, p=0.000877

n= 288, number of events= 79

\$model_list[[2]]\$coxmodels[[5]]

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	p
sz20_	0.09828	1.10327	0.23663	0.415	0.6779
pc1_	0.06503	1.06719	0.20222	0.322	0.7478
pc2_	0.14379	1.15464	0.25468	0.565	0.5724
pc3_	0.38248	1.46591	0.25822	1.481	0.1386
pc4_	0.24017	1.27146	0.20551	1.169	0.2425
age_	-0.02841	0.97199	0.42720	-0.067	0.9470

sex	0.44457	1.55983	0.52531	0.846	0.3974
bmi_0_	-0.51606	0.59687	0.50672	-1.018	0.3085
hyp_0	0.00000	1.00000	0.00000	NaN	NaN
cvd_0	-0.04389	0.95706	0.58407	-0.075	0.9401
B_dep_0	-0.68904	0.50206	0.77704	-0.887	0.3752
trig_0	-0.27094	0.76267	0.28359	-0.955	0.3394
baseline_hdl	-1.57191	0.20765	0.76556	-2.053	0.0400
stroke_0	1.42758	4.16862	1.13645	1.256	0.2091
B_smokstatus_0	-1.04822	0.35056	1.13612	-0.923	0.3562
exercise_light	-0.56413	0.56886	1.13336	-0.498	0.6187
exercise_vig	-0.57051	0.56524	0.55867	-1.021	0.3072
EduLevel_low	-0.69629	0.49843	0.76479	-0.910	0.3626
EduLevel_med	-0.75571	0.46968	0.73065	-1.034	0.3010
wealth_med	-0.80665	0.44635	0.71074	-1.135	0.2564
wealth_low	0.94185	2.56472	0.53478	1.761	0.0782
t2dm_	0.57513	1.77735	0.25163	2.286	0.0223

Likelihood ratio test=24.62 on 22 df, p=0.3158

n= 388, number of events= 22

```
$model_list[[2]]$coxmodels[[6]]
```

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
sz20_	9.289e-02	1.097e+00	2.063e-01	0.450
pc1_	-1.463e-01	8.639e-01	2.686e-01	-0.545
pc2_	-2.042e-01	8.153e-01	2.201e-01	-0.928
pc3_	-2.399e-02	9.763e-01	3.011e-01	-0.080
pc4_	3.974e-01	1.488e+00	2.779e-01	1.430
age_	2.642e-01	1.302e+00	2.838e-01	0.931
sex	1.013e+00	2.754e+00	6.323e-01	1.602
bmi_0_	3.648e-01	1.440e+00	7.854e-01	0.465
hyp_0	0.000e+00	1.000e+00	0.000e+00	NaN
cvd_0	-1.416e+00	2.426e-01	8.281e-01	-1.710
B_dep_0	2.864e-01	1.332e+00	6.598e-01	0.434
trig_0	-1.763e-01	8.384e-01	2.603e-01	-0.677
baseline_hdl	-3.099e+00	4.507e-02	1.096e+00	-2.828
stroke_0	-7.503e-01	4.722e-01	1.221e+00	-0.615
B_smokstatus_0	1.070e+00	2.915e+00	5.098e-01	2.099
exercise_light	-1.882e+01	6.724e-09	9.086e+03	-0.002
exercise_vig	1.291e-02	1.013e+00	5.725e-01	0.023
EduLevel_low	1.696e+01	2.319e+07	6.352e+03	0.003
EduLevel_med	1.703e+01	2.479e+07	6.352e+03	0.003
wealth_med	2.841e-01	1.329e+00	6.253e-01	0.454
wealth_low	2.710e-01	1.311e+00	5.996e-01	0.452
t2dm_	3.291e-01	1.390e+00	2.375e-01	1.385

	p
sz20_	0.65257
pc1_	0.58587
pc2_	0.35363
pc3_	0.93650
pc4_	0.15270
age_	0.35185

```
sex          0.10912
bmi_0_       0.64228
hyp_0        NaN
cvd_0        0.08721
B_dep_0      0.66419
trig_0       0.49841
baseline_hdl 0.00468
stroke_0     0.53878
B_smokstatus_0 0.03584
exercise_light 0.99835
exercise_vig 0.98200
EduLevel_low 0.99787
EduLevel_med 0.99786
wealth_med   0.64958
wealth_low   0.65132
t2dm_        0.16593
```

Likelihood ratio test=35.22 on 22 df, p=0.03671
n= 180, number of events= 23

```
$model_list[[2]]$coxmodels[[7]]
```

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
sz20_	-0.22661	0.79723	0.21460	-1.056
pc1_	0.28313	1.32727	0.21177	1.337
pc2_	0.01265	1.01273	0.21288	0.059
pc3_	-0.10797	0.89766	0.23514	-0.459
pc4_	-0.02565	0.97468	0.21155	-0.121
age_	0.68067	1.97520	0.28114	2.421
sex	-1.25094	0.28623	0.55805	-2.242
bmi_0_	0.35824	1.43081	0.64009	0.560
hyp_0	0.00000	1.00000	0.00000	NaN
cvd_0	-0.29740	0.74275	0.70693	-0.421
B_dep_0	0.66770	1.94975	0.64409	1.037
trig_0	0.19843	1.21948	0.71998	0.276
baseline_hdl	0.10308	1.10858	0.72778	0.142
stroke_0	1.94037	6.96136	0.58183	3.335
B_smokstatus_0	0.32365	1.38217	0.85399	0.379
exercise_light	0.32433	1.38310	1.24524	0.260
exercise_vig	0.78886	2.20089	0.48997	1.610
EduLevel_low	-0.69534	0.49891	0.83943	-0.828
EduLevel_med	0.40481	1.49902	0.65799	0.615
wealth_med	-1.40233	0.24602	0.71875	-1.951
wealth_low	-0.36620	0.69336	0.63642	-0.575
t2dm_	0.41556	1.51523	0.26328	1.578
p				
sz20_	0.290984			
pc1_	0.181235			
pc2_	0.952622			
pc3_	0.646117			
pc4_	0.903514			
age_	0.015475			


```
sex            0.024986
bmi_0_         0.575700
hyp_0          NaN
cvd_0          0.673983
B_dep_0        0.299896
trig_0         0.782856
baseline_hdl   0.887370
stroke_0       0.000853
B_smokstatus_0 0.704699
exercise_light 0.794512
exercise_vig   0.107394
EduLevel_low   0.407476
EduLevel_med   0.538404
wealth_med     0.051048
wealth_low     0.565010
t2dm_          0.114467
```

Likelihood ratio test=35.55 on 22 df, p=0.0339
n= 451, number of events= 22

```
$model_list[[2]]$coxmodels[[8]]
```

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
sz20_	0.092746	1.097183	0.184441	0.503
pc1_	0.163986	1.178198	0.205694	0.797
pc2_	-0.127102	0.880643	0.207556	-0.612
pc3_	-0.174549	0.839836	0.180257	-0.968
pc4_	0.006118	1.006137	0.210580	0.029
age_	0.612737	1.845475	0.249759	2.453
sex	-1.361952	0.256160	0.496555	-2.743
bmi_0_	0.595637	1.814185	0.258399	2.305
hyp_0	0.000000	1.000000	0.000000	NaN
cvd_0	-0.389457	0.677424	0.663163	-0.587
B_dep_0	1.146017	3.145638	0.513666	2.231
trig_0	1.844822	6.326974	0.866991	2.128
baseline_hdl	-0.753566	0.470685	0.794126	-0.949
stroke_0	1.629750	5.102601	1.149722	1.418
B_smokstatus_0	0.019684	1.019879	0.707747	0.028
exercise_light	0.315577	1.371051	0.841838	0.375
exercise_vig	0.321785	1.379588	0.451976	0.712
EduLevel_low	-0.262077	0.769452	0.772722	-0.339
EduLevel_med	-0.268631	0.764426	0.694049	-0.387
wealth_med	1.079594	2.943483	0.525738	2.053
wealth_low	-0.903410	0.405186	0.743985	-1.214
t2dm_	0.515064	1.673745	0.203742	2.528
p				
sz20_	0.61507			
pc1_	0.42532			
pc2_	0.54029			
pc3_	0.33288			
pc4_	0.97682			
age_	0.01415			

```
sex          0.00609
bmi_0_       0.02116
hyp_0        NaN
cvd_0        0.55702
B_dep_0      0.02568
trig_0       0.03335
baseline_hdl 0.34266
stroke_0     0.15633
B_smokstatus_0 0.97781
exercise_light 0.70776
exercise_vig 0.47650
EduLevel_low 0.73449
EduLevel_med 0.69872
wealth_med   0.04003
wealth_low   0.22464
t2dm_        0.01147
```

Likelihood ratio test=45.2 on 22 df, p=0.002507
n= 483, number of events= 28

```
$model_list[[2]]$clusters
[1] 1.855647 0.195628 1.593133 3.489811 0.732971 1.530041
[7] 0.623480 0.662366
```

```
$model_list[[3]]
$model_list[[3]]$vimp10
      bmi_0_      trig_0      t2dm_
0.0501911470 0.0105764204 0.0081634803
      hyp_0      pc2_      pc4_
0.0049587200 0.0041929935 0.0034377043
      wealth_low      age_      sex
0.0030292219 0.0029134615 0.0022003964
      B_dep_0      stroke_0 exercise_light
0.0019859550 0.0013853294 0.0009512279
      cvd_0 B_smokstatus_0      pc3_
0.0007434285 0.0005901883 0.0003876076
```

```
$model_list[[3]]$treemodel
n= 3979
```

```
node), split, n, deviance, yval
* denotes terminal node
```

- 1) root 3979 2275.804000 1.0000000
- 2) bmi_0_ < 0.422484 2698 1074.604000 0.5797049
 - 4) trig_0 < 2.05 2114 686.485500 0.4570699
 - 8) bmi_0_ < -0.041124 1561 422.973300 0.3651620 *
 - 9) bmi_0_ >= -0.041124 553 254.165400 0.7274050 *
 - 5) trig_0 >= 2.05 584 368.182300 1.0261840
 - 10) bmi_0_ >= 0.3333768 32 1.465301 0.2673493 *
 - 11) bmi_0_ < 0.3333768 552 362.400400 1.0825610 *
- 3) bmi_0_ >= 0.422484 1281 1076.318000 1.9398860
 - 6) t2dm_ < 0.4308968 850 622.564700 1.5261740
 - 12) trig_0 < 1.25 184 68.308610 0.5596226 *

```

13) trig_0>=1.25 666 536.408300 1.8224620 *
7) t2dm_>=0.4308968 431 434.870600 2.7844210
14) trig_0< 1.65 203 174.285200 2.0172130 *
15) trig_0>=1.65 228 254.537800 3.3716600 *

```

```
$model_list[[3]]$coxmodels
```

```
$model_list[[3]]$coxmodels[[1]]
```

```
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
sz20_	0.13296	1.14221	0.11013	1.207
pc1_	0.02041	1.02062	0.10079	0.203
pc2_	-0.34225	0.71017	0.11089	-3.086
pc3_	-0.01998	0.98022	0.09683	-0.206
pc4_	-0.25231	0.77700	0.10830	-2.330
age_	-0.11427	0.89201	0.13621	-0.839
sex	-0.21471	0.80677	0.24152	-0.889
bmi_0_	0.28614	1.33129	0.10597	2.700
hyp_0	0.09465	1.09928	0.21180	0.447
cvd_0	0.43969	1.55223	0.29442	1.493
B_dep_0	-0.12345	0.88387	0.29497	-0.419
trig_0	-0.01500	0.98511	0.07945	-0.189
baseline_hdl	-1.02951	0.35718	0.43012	-2.394
stroke_0	-1.03080	0.35672	1.02758	-1.003
B_smokstatus_0	0.48948	1.63146	0.26104	1.875
exercise_light	0.18226	1.19992	0.42079	0.433
exercise_vig	-0.14373	0.86612	0.24829	-0.579
EduLevel_low	0.32861	1.38904	0.38088	0.863
EduLevel_med	0.16861	1.18366	0.35745	0.472
wealth_med	0.13011	1.13896	0.27841	0.467
wealth_low	0.30394	1.35519	0.27831	1.092
t2dm_	0.31233	1.36661	0.17180	1.818

p

sz20_	0.22729
pc1_	0.83951
pc2_	0.00203
pc3_	0.83652
pc4_	0.01982
age_	0.40149
sex	0.37399
bmi_0_	0.00693
hyp_0	0.65495
cvd_0	0.13533
B_dep_0	0.67557
trig_0	0.85025
baseline_hdl	0.01669
stroke_0	0.31580
B_smokstatus_0	0.06078
exercise_light	0.66492
exercise_vig	0.56267
EduLevel_low	0.38827
EduLevel_med	0.63713
wealth_med	0.64026

```
wealth_low      0.27480
t2dm_           0.06907
```

```
Likelihood ratio test=50.01 on 22 df, p=0.0005854
n= 666, number of events= 102
```

```
$model_list[[3]]$coxmodels[[2]]
```

```
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
sz20_	0.06677	1.06905	0.18739	0.356
pc1_	-0.15110	0.85976	0.23694	-0.638
pc2_	-0.26560	0.76675	0.20108	-1.321
pc3_	-0.09696	0.90759	0.18418	-0.526
pc4_	-0.07812	0.92485	0.23526	-0.332
age_	-0.11801	0.88868	0.24827	-0.475
sex	-1.29729	0.27327	0.49460	-2.623
bmi_0_	0.54440	1.72357	0.25462	2.138
hyp_0	0.37082	1.44892	0.41074	0.903
cvd_0	0.13821	1.14822	0.53367	0.259
B_dep_0	1.07285	2.92369	0.48199	2.226
trig_0	-0.27628	0.75860	0.78702	-0.351
baseline_hdl	0.53325	1.70446	0.64675	0.825
stroke_0	1.16581	3.20853	0.82619	1.411
B_smokstatus_0	0.34824	1.41657	0.47880	0.727
exercise_light	-0.28443	0.75244	0.84095	-0.338
exercise_vig	-0.27874	0.75673	0.54382	-0.513
EduLevel_low	1.13301	3.10499	0.90371	1.254
EduLevel_med	0.78704	2.19688	0.85591	0.920
wealth_med	-0.48854	0.61352	0.54469	-0.897
wealth_low	-0.10390	0.90132	0.49910	-0.208
t2dm_	0.16237	1.17629	0.34153	0.475

	p
sz20_	0.72161
pc1_	0.52366
pc2_	0.18654
pc3_	0.59857
pc4_	0.73984
age_	0.63453
sex	0.00872
bmi_0_	0.03251
hyp_0	0.36663
cvd_0	0.79564
B_dep_0	0.02602
trig_0	0.72556
baseline_hdl	0.40965
stroke_0	0.15822
B_smokstatus_0	0.46704
exercise_light	0.73519
exercise_vig	0.60825
EduLevel_low	0.20994
EduLevel_med	0.35781
wealth_med	0.36977

```
wealth_low      0.83510
t2dm_           0.63450
```

Likelihood ratio test=27.59 on 22 df, p=0.1897
n= 203, number of events= 33

```
$model_list[[3]]$coxmodels[[3]]
```

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
sz20_	6.613e-01	1.937e+00	2.257e-01	2.930
pc1_	1.110e-01	1.117e+00	1.658e-01	0.669
pc2_	-9.100e-02	9.130e-01	1.733e-01	-0.525
pc3_	-4.309e-02	9.578e-01	1.832e-01	-0.235
pc4_	-7.992e-02	9.232e-01	1.631e-01	-0.490
age_	4.390e-01	1.551e+00	2.352e-01	1.867
sex	-7.120e-03	9.929e-01	4.059e-01	-0.018
bmi_0_	-6.880e-02	9.335e-01	1.314e+00	-0.052
hyp_0	9.740e-01	2.648e+00	3.576e-01	2.724
cvd_0	-7.281e-01	4.828e-01	6.548e-01	-1.112
B_dep_0	6.250e-01	1.868e+00	4.622e-01	1.352
trig_0	2.980e-01	1.347e+00	4.817e-01	0.619
baseline_hdl	-4.137e-01	6.612e-01	5.999e-01	-0.690
stroke_0	5.425e-01	1.720e+00	1.133e+00	0.479
B_smokstatus_0	4.355e-01	1.546e+00	5.402e-01	0.806
exercise_light	-1.610e+01	1.018e-07	3.115e+03	-0.005
exercise_vig	7.765e-01	2.174e+00	3.653e-01	2.125
EduLevel_low	-8.240e-01	4.387e-01	6.687e-01	-1.232
EduLevel_med	9.452e-02	1.099e+00	5.328e-01	0.177
wealth_med	6.146e-01	1.849e+00	4.175e-01	1.472
wealth_low	-2.029e-01	8.163e-01	5.260e-01	-0.386
t2dm_	5.191e-01	1.681e+00	2.051e-01	2.531

	p
sz20_	0.00339
pc1_	0.50343
pc2_	0.59953
pc3_	0.81404
pc4_	0.62409
age_	0.06191
sex	0.98601
bmi_0_	0.95824
hyp_0	0.00646
cvd_0	0.26615
B_dep_0	0.17633
trig_0	0.53612
baseline_hdl	0.49049
stroke_0	0.63217
B_smokstatus_0	0.42013
exercise_light	0.99588
exercise_vig	0.03355
EduLevel_low	0.21788
EduLevel_med	0.85919
wealth_med	0.14102

```
wealth_low      0.69963
t2dm_           0.01139
```

Likelihood ratio test=38.57 on 22 df, p=0.01578
n= 553, number of events= 35

```
$model_list[[3]]$coxmodels[[4]]
```

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
sz20_	0.02555	1.02588	0.14731	0.173
pc1_	0.01177	1.01184	0.14653	0.080
pc2_	-0.10114	0.90380	0.14192	-0.713
pc3_	-0.04114	0.95969	0.16094	-0.256
pc4_	0.21048	1.23427	0.12896	1.632
age_	0.23956	1.27069	0.16752	1.430
sex	0.38384	1.46790	0.31636	1.213
bmi_0_	1.05455	2.87068	0.40535	2.602
hyp_0	0.21465	1.23943	0.29176	0.736
cvd_0	-0.17502	0.83944	0.46003	-0.380
B_dep_0	0.30753	1.36006	0.37728	0.815
trig_0	-0.04432	0.95665	0.13036	-0.340
baseline_hdl	-0.02009	0.98011	0.51700	-0.039
stroke_0	-0.61931	0.53832	1.05292	-0.588
B_smokstatus_0	0.54726	1.72852	0.33360	1.640
exercise_light	0.35751	1.42976	0.64097	0.558
exercise_vig	-0.26190	0.76959	0.34030	-0.770
EduLevel_low	0.94115	2.56294	0.65120	1.445
EduLevel_med	0.27985	1.32293	0.64726	0.432
wealth_med	0.13125	1.14025	0.38040	0.345
wealth_low	-0.05333	0.94807	0.38944	-0.137
t2dm_	0.22650	1.25421	0.13957	1.623

	p
sz20_	0.86230
pc1_	0.93598
pc2_	0.47606
pc3_	0.79823
pc4_	0.10265
age_	0.15272
sex	0.22502
bmi_0_	0.00928
hyp_0	0.46190
cvd_0	0.70361
B_dep_0	0.41500
trig_0	0.73389
baseline_hdl	0.96900
stroke_0	0.55641
B_smokstatus_0	0.10091
exercise_light	0.57700
exercise_vig	0.44152
EduLevel_low	0.14839
EduLevel_med	0.66548
wealth_med	0.73007

```
wealth_low      0.89109
t2dm_           0.10462
```

Likelihood ratio test=34.63 on 22 df, p=0.04229
n= 552, number of events= 53

```
$model_list[[3]]$coxmodels[[5]]
```

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
sz20_	-0.03302	0.96752	0.14510	-0.228
pc1_	0.23815	1.26890	0.13701	1.738
pc2_	0.05222	1.05361	0.14252	0.366
pc3_	-0.01932	0.98086	0.13766	-0.140
pc4_	0.05257	1.05397	0.13062	0.402
age_	0.47762	1.61223	0.17114	2.791
sex	-0.70627	0.49348	0.33078	-2.135
bmi_0_	-0.12872	0.87922	0.35548	-0.362
hyp_0	0.76150	2.14148	0.31042	2.453
cvd_0	-0.19167	0.82558	0.44807	-0.428
B_dep_0	-0.43949	0.64436	0.54255	-0.810
trig_0	-0.22853	0.79571	0.40036	-0.571
baseline_hdl	-0.41038	0.66340	0.43835	-0.936
stroke_0	1.33555	3.80207	0.48427	2.758
B_smokstatus_0	0.24323	1.27537	0.42016	0.579
exercise_light	-1.04345	0.35224	1.04982	-0.994
exercise_vig	-0.01494	0.98517	0.32089	-0.047
EduLevel_low	0.68321	1.98022	0.60166	1.136
EduLevel_med	0.84444	2.32668	0.54772	1.542
wealth_med	-0.07732	0.92559	0.35337	-0.219
wealth_low	-0.17185	0.84211	0.38993	-0.441
t2dm_	0.14340	1.15419	0.15710	0.913

	p
sz20_	0.82001
pc1_	0.08218
pc2_	0.71405
pc3_	0.88836
pc4_	0.68735
age_	0.00526
sex	0.03275
bmi_0_	0.71728
hyp_0	0.01416
cvd_0	0.66882
B_dep_0	0.41791
trig_0	0.56814
baseline_hdl	0.34918
stroke_0	0.00582
B_smokstatus_0	0.56265
exercise_light	0.32026
exercise_vig	0.96286
EduLevel_low	0.25615
EduLevel_med	0.12313
wealth_med	0.82680

```
wealth_low      0.65942
t2dm_           0.36136
```

Likelihood ratio test=48.69 on 22 df, p=0.0008796
n= 1561, number of events= 50

```
$model_list[[3]]$coxmodels[[6]]
```

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
sz20_	-6.102e-01	5.432e-01	3.891e-01	-1.568
pc1_	-6.174e-01	5.394e-01	4.803e-01	-1.285
pc2_	-7.432e-02	9.284e-01	4.773e-01	-0.156
pc3_	4.674e-01	1.596e+00	5.280e-01	0.885
pc4_	3.082e-01	1.361e+00	4.316e-01	0.714
age_	9.790e-01	2.662e+00	4.865e-01	2.012
sex	-1.190e+00	3.042e-01	1.019e+00	-1.168
bmi_0_	-2.357e-01	7.900e-01	8.526e-01	-0.276
hyp_0	1.047e+00	2.851e+00	9.626e-01	1.088
cvd_0	-1.842e+00	1.586e-01	1.352e+00	-1.362
B_dep_0	-1.953e+01	3.312e-09	1.520e+04	-0.001
trig_0	6.362e+00	5.796e+02	3.707e+00	1.716
baseline_hdl	1.518e+00	4.561e+00	1.236e+00	1.228
stroke_0	-2.039e+01	1.390e-09	4.206e+04	0.000
B_smokstatus_0	-1.832e+01	1.106e-08	1.456e+04	-0.001
exercise_light	-1.885e+01	6.501e-09	3.195e+04	-0.001
exercise_vig	-7.061e-01	4.936e-01	9.686e-01	-0.729
EduLevel_low	-6.542e-01	5.198e-01	1.392e+00	-0.470
EduLevel_med	-8.930e-01	4.094e-01	1.350e+00	-0.662
wealth_med	4.313e-01	1.539e+00	1.187e+00	0.363
wealth_low	2.983e-01	1.348e+00	1.514e+00	0.197
t2dm_	-3.578e-01	6.992e-01	6.561e-01	-0.545

	p
sz20_	0.1168
pc1_	0.1987
pc2_	0.8763
pc3_	0.3761
pc4_	0.4751
age_	0.0442
sex	0.2427
bmi_0_	0.7822
hyp_0	0.2765
cvd_0	0.1731
B_dep_0	0.9990
trig_0	0.0861
baseline_hdl	0.2194
stroke_0	0.9996
B_smokstatus_0	0.9990
exercise_light	0.9995
exercise_vig	0.4660
EduLevel_low	0.6384
EduLevel_med	0.5083
wealth_med	0.7163


```
wealth_low      0.8438
t2dm_           0.5856
```

Likelihood ratio test=24.55 on 22 df, p=0.319
n= 184, number of events= 9

```
$model_list[[3]]$coxmodels[[7]]
```

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	p
sz20_	0.03439	1.03499	0.13551	0.254	0.7997
pc1_	-0.13944	0.86985	0.14376	-0.970	0.3321
pc2_	0.16135	1.17509	0.13241	1.219	0.2230
pc3_	0.17090	1.18638	0.16827	1.016	0.3098
pc4_	-0.17783	0.83709	0.14630	-1.215	0.2242
age_	-0.15115	0.85972	0.19957	-0.757	0.4488
sex	-0.31534	0.72954	0.32603	-0.967	0.3334
bmi_0_	0.08296	1.08650	0.18498	0.448	0.6538
hyp_0	0.54736	1.72868	0.26530	2.063	0.0391
cvd_0	0.22823	1.25637	0.39334	0.580	0.5618
B_dep_0	0.46220	1.58756	0.34814	1.328	0.1843
trig_0	-0.01318	0.98691	0.11884	-0.111	0.9117
baseline_hdl	-0.09273	0.91144	0.58637	-0.158	0.8744
stroke_0	0.60500	1.83125	1.26482	0.478	0.6324
B_smokstatus_0	0.08984	1.09400	0.37902	0.237	0.8126
exercise_light	-0.11120	0.89476	0.53316	-0.209	0.8348
exercise_vig	-0.75435	0.47032	0.35661	-2.115	0.0344
EduLevel_low	0.64763	1.91101	0.54215	1.195	0.2323
EduLevel_med	0.58626	1.79726	0.50649	1.158	0.2471
wealth_med	-0.21363	0.80765	0.38953	-0.548	0.5834
wealth_low	0.28505	1.32983	0.37021	0.770	0.4413
t2dm_	0.03500	1.03562	0.27428	0.128	0.8985

Likelihood ratio test=26.46 on 22 df, p=0.2324
n= 228, number of events= 63

```
$model_list[[3]]$coxmodels[[8]]
```

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	p
sz20_	0	1	0	NaN	NaN
pc1_	0	1	0	NaN	NaN
pc2_	0	1	0	NaN	NaN
pc3_	0	1	0	NaN	NaN
pc4_	0	1	0	NaN	NaN
age_	0	1	0	NaN	NaN
sex	0	1	0	NaN	NaN
bmi_0_	0	1	0	NaN	NaN
hyp_0	0	1	0	NaN	NaN
cvd_0	0	1	0	NaN	NaN

B_dep_0	0	1	0 NaN NaN
trig_0	0	1	0 NaN NaN
baseline_hdl	0	1	0 NaN NaN
stroke_0	0	1	0 NaN NaN
B_smokstatus_0	0	1	0 NaN NaN
exercise_light	0	1	0 NaN NaN
exercise_vig	0	1	0 NaN NaN
EduLevel_low	0	1	0 NaN NaN
EduLevel_med	0	1	0 NaN NaN
wealth_med	0	1	0 NaN NaN
wealth_low	0	1	0 NaN NaN
t2dm_	0	1	0 NaN NaN

Likelihood ratio test=0 on 22 df, p=1

n= 32, number of events= 0

\$model_list[[3]]\$clusters

[1] 1.822462 2.017213 0.727405 1.082561 0.365162 0.559623

[7] 3.371660 0.267349

\$time

Time difference of 2.178813 mins