The Combination of Supervised and Unsupervised Learning based Risk Stratification and Phenotyping in Pulmonary Arterial Hypertension - a Long-term Retrospective Multicenter Trial

Supplementary Material

Innsbruck PAH registry

2023-03-24

# Supplementary Methods

## Software

Data transformation, analysis and result visualization was accomplished by R version 4.2.3.

Import of the study data was accomplished with the *readxl* package1. Tabular data were handled by the *tidyverse* package bundle2 and the packages *rlang*3 and [*trafo*](https://github.com/PiotrTymoszuk/trafo). Text data were handled with the *stringi* package4.

Exploratory data analysis and statistical hypothesis for numeric and categorical variables testing was done with *rstatix*5 and [*ExDA*](https://github.com/PiotrTymoszuk/ExDA). Survival analysis was accomplished with the packages *survival*6, *survminer*7, *pec*8, *rms*9, *glmnet*10, *survmisc*11 and the development package [*coxExtensions*](https://github.com/PiotrTymoszuk/coxExtensions). Cross-validation folds were generated with *caret*12. Clustering analysis was done with the packages *cluster*13, *philentropy*14, *factoextra*15, *umap*16 and [*clustTools*](https://github.com/PiotrTymoszuk/clustTools).

Results were visualized with *ggplot2*2 (plots of model estimates, violin, bubble, bar and Forest plots), [*ExDA*](https://github.com/PiotrTymoszuk/ExDA) (stack plots, plots of effect sizes/significance of differences between clusters), *survminer* and [*coxExtensions*](https://github.com/PiotrTymoszuk/coxExtensions) (Kaplan-Meier plots) and [*clustTools*](https://github.com/PiotrTymoszuk/clustTools) (UMAP layout plots, cluster diagnostic plots). Tables were generated with *flextable*17, figures were created with *cowplot*18.

The Supplementary Material file was built with *rmarkdown* environment19 with the packages *knitr*20 and *bookdown*21.

## Data import and transformation, visualization, descriptive statistic

The study dataset was imported into from an Excel file (function read\_excel(), package *readxl*) and formatted with an in-house developed script.

For univariable survival modeling and construction of candidate risk signatures, a set of categorical 19 demographic, biochemical, right-heart catheter, laboratory, ultrasound and lung function parameters recorded at PH diagnosis was used. To improve normality of some independent variables (NT-pro-BNP, RDW, TF-Sat, ferritin) prior to survival modeling, log transformation was applied. For the list of modeling variables and their transformation and stratification scheme, see: **Supplementary Table S1**.

## Descriptive statistic, hypothesis testing, multiple comparisons

Numeric variables are presented as medians with interquartile ranges (IQR) and ranges. Categorical variables are shown as percentages and counts within the complete observation set for each variable category (function explore(), package [*ExDA*](https://github.com/PiotrTymoszuk/ExDA)).

As the some of the analyzed numeric variables were non-normally distributed as checked by Shapiro-Wilk test (function explore(what = 'normality'), package [*ExDA*](https://github.com/PiotrTymoszuk/ExDA)), differences in median values of numeric variables between the study cohorts or participant clusters were investigated by Mann-Whitney test and r effect size statistic. Differences in frequency distribution of categorical variables between the study cohorts or participant clusters were assessed by test and Cramer’s V effect size statistic. Statistical significance for the comparisons was determined with the function compare\_variables() from the [*ExDA*](https://github.com/PiotrTymoszuk/ExDA) package.

Risk strata returned by established PAH risk assessment tools (mRASP: modified risk assessment score of PAH; COMPERA: comparative, prospective registry of newly initiated therapies for pulmonary hypertension score; SPAHR: Swedish pulmonary arterial hypertension registry score; FPHR 3p: french pulmonary arterial hypertension registry score; FPHR 4p: french pulmonary arterial hypertension registry score, 4 parameters) was converted to numeric values (mRASP/COMPERA/SPAHR: low: 1, intermediate: 2, high risk: 3, FPHR scales: number of risk factors). Their correlation with the newly developed Elastic Net score was assessed by Spearman’s test (function correlate\_variables(type = 'spearman'), package [*ExDA*](https://github.com/PiotrTymoszuk/ExDA)).

Differences in survival between the participant clusters or participants stratified by risk score tertiles or by the PAH cluster assignment were compared by Kaplan-Meier (KM) analysis with log-rank test6,7,22 (functions surv\_fit() and surv\_pvalue(), package *survminer*). Fractions of surviving participants were presented in Kaplan-Meier plots (function ggsurvplot(), package *survminer*).

For each analysis task and cohort, p values were corrected for multiple comparisons with Benjamini-Hochberg method23.

## Univariable Cox survival modeling

Association of independent categorical and numeric variables (**Supplementary Table S1**) with overall survival time was assessed by series of univariable Cox proportional hazard models constructed for the Innsbruck and Linz/Vienna cohort using the *survival*6 and [*coxExtensions*](https://github.com/PiotrTymoszuk/coxExtensions) packages (functions coxph() and as\_coxex()). Numeric variables were normalized median-centered (function scale(x, center = median(x))). To account for non-linear associations of numeric independent variables, both 1st and 2nd order terms were included in the Cox models. Significance of the hazard ratio estimates was determined by Wald Z test. P values were corrected for multiple comparisons with Benjamini-Hochberg method23. Proportional hazard assumption was checked with the summary(type = 'assumptions') method (package [*coxExtensions*](https://github.com/PiotrTymoszuk/coxExtensions), a wrapper around cox.zph() from survival)24. Concordance indexes (C-index)25, integrated Brier scores (IBS)26 and 11 served as measures of predictive performance of the Cox models and were computed with the summary(type = 'fit') method (package [*coxExtensions*](https://github.com/PiotrTymoszuk/coxExtensions), employing the concordance(), pec() and rsq() functions from the packages *survival*, *pec* and *survMisc*).

For the full modeling results, see: **Supplementary Table S3**.

## Multivariable Cox survival modeling with the Elastic Net technique

Multi-parameter Cox modeling with the set of independent categorical and numeric variables (**Supplementary Table S1**) was accomplished by Elastic Net machine learning technique10,27. Data pre-processing included normalization and median centering of numeric independent variables (function scale(x, center = median(x))) and conversion of categorical features to dummy numeric variables (function model.matrix()). To account for non-linear associations of numeric independent variables, both 1st and 2nd order terms were included in the elastic net model development. The Elastic Net Cox proportional hazard model was trained in the Innsbruck cohort (function glmnet(alpha = 0.5, family = 'cox'), package *glmnet*). The optimal lambda parameter ( = 0.166) for the training cohort model construction was found by 200-repetition 10-fold cross-validation (function cv.glmnet(), package *glmnet*) and corresponded to the minimum of cross-validation error (lambda.min parameter). The values of non-zero Elastic Net model coefficients are presented in **Figure 2A**. Subsequently, the Elastic Net model linear predictor (LP) scores were calculated for the training IBK and test Linz/Vienna cohort and their association with overall survival was assessed by univariable Cox modeling28. C-index, IBS and served as performance measures as described above for univariable Cox modeling11,25,26 (**Supplementary Table S4**). Differences in survival between study participants stratified by the LP score tertiles were assessed by log-rank test as described above7,22.

## LASSO ensemble model

Multi-parameter Cox model including the newly developed Elastic Net score and established PAH risk assessment tools available for both cohorts (mRASP, SPAHR, COMPERA: low, intermediate and high risk strata, FPHR 3p and FPHR 4p: number of risk factors coded as categorical variable) were constructed with the LASSO algorithm10,27,29. This LASSO ensemble model was trained in the Innsbruck cohort (function glmnet(alpha = 0.5, family = 'cox'), package *glmnet*). The optimal lambda parameter ( = 0.166) for the training cohort model construction was found by 200-repetition 10-fold cross-validation (function cv.glmnet(), package *glmnet*) and corresponded to the minimum of cross-validation error (lambda.min parameter). The values of non-zero LASSO model coefficients are presented in **Supplementary Figure 3A**. Subsequently, the LASSO model linear predictor (LP) scores were calculated for the training IBK and test Linz/Vienna cohort and their association with overall survival was assessed by univariable Cox modeling28 analogically to the Elastic Net scores described above. Performance metrics of the LASSO ensemble are presented in **Supplementary Table S4**.

## Comparison of risk assessment tools

Univariable Cox models were constructed for the Elastic Net score and the LASSO ensemble described above and for the established PAH risk assessment tools available for both cohorts (mRASP, SPAHR, COMPERA: low, intermediate and high risk strata, FPHR 3p and FPHR 4p: number of risk factors coded as categorical variable) as described for univariable Cox modeling. Performance metrics: C-indexes, IBS and are presented in **Supplementary Table S4**.

## Clustering of the study participants

Clustering of the study participants in the Innsbruck training cohort in respect to the variables found associated with overall survival by the Elastic Net Cox modeling (**Figure 2A**; Age, SMWD, log RDW, CI, PVR, log NT-pro-BNP, RAA) was done with the PAM algorithm (partitioning around medoids)13 with the cosine distance between the study participants14 (clustering objects generated with kcluster(distance\_method = 'cosine', clust\_fun = 'pam', k = 2), package [*clustTools*](https://github.com/PiotrTymoszuk/clustTools)). Prior to clustering, the numeric variables were normalized and median centered (function scale(x, center = median(x))). The PAM/cosine distance clustering procedure demonstrated the superior fraction of ‘explained’ clustering variance (ratio of total between cluster sum of squared to total sum of squares) and the optimal performance in 10-fold cross-validation measured by the fraction of correct cluster assignments30 as compared with hierarchical clustering, k-means and self-organizing map algorithms (**Supplementary Figure S4A**). Clustering variances and CV accuracy were computed with the var() and cv() methods provided by the [*clustTools*](https://github.com/PiotrTymoszuk/clustTools) called for the clustering objects.

The choice of cluster number k was based on the bend of the within-cluster sum-of-squares curve, the peak value of silhouette statistic15,31 and optimal cluster assignment accuracy in 10-fold CV assessed for the PAM/cosine distance clustering structures generated for multiple k values (methods plot() and cv(), package [*clustTools*](https://github.com/PiotrTymoszuk/clustTools)) (**Supplementary Figure S4B**).

The importance of specific clustering features was determined by comparing the ‘explained’ clustering variances of the original clustering structure with the clustering objects with randomly re-shuffled clustering features32 with the impact() function from the [*clustTools*](https://github.com/PiotrTymoszuk/clustTools) package.

Assignment of the test Linz/Vienna cohort participants to the PAH clusters defined in the Innsbruck cohort was done with an inverse distance-weighted 7-nearest neighbor label propagation procedure32–34. Assignment of participants to the PAH clusters was visualized with two-dimensional cosine distance UMAP (Uniform Manifold Approximation and Projection, function plot(type = 'components', red\_fun = 'umap', kdim = '2', with = 'data'), package [*clustTools*](https://github.com/PiotrTymoszuk/clustTools))16,35.

Differences in study variables between the participant clusters were determined by Mann-Whitney or test as described above (**Supplementary Tables S5 - S5**). Differences in overall survival between the clusters were compared with KM method and log-rank test as described above.

# Data and code availability

The study data set is available at serious request to the corresponding author. The analysis R code was deposited on GitHub (<https://github.com/PiotrTymoszuk/PAH-biomarker>).

# Supplementary Tables

Supplementary Table S1: Study variables.

| **Name in R** | **Description** | **Name in the report** | **Unit** | **Stratification** | **Used in risk modeling** |
| --- | --- | --- | --- | --- | --- |
| age\_fc | age at the diagnosis | Age | years |  | yes |
| SMWD | Six Minute Walk Distance | SMWD | m |  | yes |
| mPAP | Mean pulmonary arterial pressure | mPAP | mmHg |  | yes |
| event3 | 3-year mortality | 3-year mortality |  | no; yes | no |
| event5 | 5-year mortality | 5-year mortality |  | no; yes | no |
| death\_study | overall mortality during the study period | Overall mortality |  |  | no |
| death\_study\_fct | overall mortality during the study period | Overall mortality |  | no; yes | no |
| surv\_months | survival time from the diagnosis | OS | months |  | no |
| sex | sex | Sex |  | female; male | yes |
| PVR | Pulmonary vascular resistance | PVR | Wood |  | yes |
| PCWP | Pulmonary capillary wedge pressure | PCWP | mmHg |  | yes |
| anemia | anemia | Anemia |  | no; yes | yes |
| RDW\_log | RDW | log RDW | % |  | yes |
| renal\_ins | Renal insufficiency, GFR < 60% | Renal insufficiency |  | no; yes | yes |
| FT\_log | Ferritin | log FT | ng/ml |  | yes |
| TSAT\_log | Transferrin saturation | log TF-Sat | % |  | yes |
| MCV | MCV | MCV | fl |  | yes |
| NTproBNP\_log | NT-pro-BNP | log NT-pro-BNP | pg/ml |  | yes |
| percardial\_effusion | Percardial effusion | Percardial effusion |  | no; yes | yes |
| RA\_area | Right Atrial Area | RAA | cm2 |  | yes |
| cardiac\_index | Cardiac index | CI |  |  | yes |
| mRAP | Mean right atrial pressure | mRAP | mmHg |  | yes |
| WHOFc\_class | WHO Functional Classification | WHO class |  | I/II; III/IV | yes |
| SO2\_RL\_class | O2 saturation | SO2 | % | ≥95; <95 | yes |
| mRASP | modified Risk Assessment Score of PAH | mRASP |  | low; int; high | no |
| Compera | Comparative, Prospective Registry of Newly Initiated Therapies for Pulmonary Hypertension score | COMPERA |  | low; int; high | no |
| SPAHR | Swedish Pulmonary Arterial Hypertension Registry score | SPAHR |  | low; int; high | no |
| FRENCH3p | French Pulmonary Arterial Hypertension Registry score, 3 parameters | FPHR 3p |  | 0; 1; 2; 3 | no |
| FRENCH4p | French Pulmonary Arterial Hypertension Registry score, 4 parameters | FPHR 4p |  | 0; 1; 2; 3; 4 | no |
| Reveal\_lite2\_3\_cat | Reveal Lite, Risk Classes | Reveal Lite |  | low; int; high | no |
| Reveal2\_risk\_3\_cat | Reveal 2.0 Risk classes | Reveal 2.0 |  | low; int; high | no |

Supplementary Table S2: Supplementary characteristic of the study cohorts. Numeric variables are presented as medians with interquartile ranges (IQR) and ranges. Categorical variables are presented as percentages and counts within the complete observation set.

| **Variablea** | **IBK** | **LZ/W** | **Significanceb** | **Effect sizeb** |
| --- | --- | --- | --- | --- |
| Participants, n | 100 | 83 |  |  |
| mRAP, mmHg | 10 [IQR: 6 - 13] range: 2 - 26 | 6 [IQR: 3 - 9] range: 0 - 20 | p < 0.001 | r = 0.43 |
| SO2, % | ≥95: 47% (47) <95: 53% (53) | ≥95: 48% (40) <95: 52% (43) | ns (p = 0.99) | V = 0.012 |
| log NT-pro-BNP, pg/ml | 6.6 [IQR: 5.1 - 7.7] range: 3.4 - 11 | 6.6 [IQR: 5.3 - 7.4] range: 3.2 - 10 | ns (p = 0.95) | r = 0.0079 |
| CI | 2.4 [IQR: 1.9 - 2.8] range: 1.6 - 4.2 | 2.6 [IQR: 2.2 - 3] range: 1.4 - 3.8 | p = 0.033 | r = 0.19 |
| RAA, cm² | 22 [IQR: 17 - 24] range: 13 - 34 | 18 [IQR: 17 - 23] range: 13 - 30 | p = 0.014 | r = 0.22 |
| MCV, fl | 88 [IQR: 85 - 91] range: 58 - 100 | 89 [IQR: 86 - 93] range: 76 - 110 | ns (p = 0.33) | r = 0.1 |
| log RDW, % | 2.7 [IQR: 2.6 - 2.8] range: 2.5 - 3.1 | 2.7 [IQR: 2.7 - 2.8] range: 2.5 - 3.1 | p = 0.014 | r = 0.22 |
| log FT, ng/ml | 4.1 [IQR: 3.5 - 4.7] range: 1.1 - 6.5 | 4.4 [IQR: 3.4 - 4.9] range: 1.9 - 7.1 | ns (p = 0.6) | r = 0.056 |
| log TF-Sat, % | 3 [IQR: 2.6 - 3.3] range: 0.69 - 4.3 | 3 [IQR: 2.5 - 3.4] range: 0.69 - 4.5 | ns (p = 0.52) | r = 0.066 |
| mRASP, risk strata | low: 32% (32) int: 52% (52) high: 16% (16) | low: 39% (32) int: 53% (44) high: 8.4% (7) | ns (p = 0.4) | V = 0.12 |
| COMPERA, risk strata | low: 22% (22) int: 64% (64) high: 14% (14) | low: 39% (32) int: 59% (49) high: 2.4% (2) | p = 0.014 | V = 0.25 |
| SPAHR, risk strata | low: 24% (24) int: 65% (65) high: 11% (11) | low: 34% (28) int: 65% (54) high: 1.2% (1) | p = 0.043 | V = 0.21 |
| FPHR 3p, number of risk factors | 0: 8% (8) 1: 18% (18) 2: 30% (30) 3: 44% (44) | 0: 13% (11) 1: 20% (17) 2: 30% (25) 3: 36% (30) | ns (p = 0.72) | V = 0.11 |
| FPHR 4p, number of risk factors | 0: 3% (3) 1: 10% (10) 2: 26% (26) 3: 31% (31) 4: 30% (30) | 0: 13% (11) 1: 25% (21) 2: 25% (21) 3: 27% (22) 4: 9.6% (8) | p = 0.002 | V = 0.35 |
| 3-year mortality | 13% (13) | 11% (9) | ns (p = 0.94) | V = 0.033 |
| Overall mortality | 33% (33) | 24% (20) | ns (p = 0.38) | V = 0.098 |
| amRAP: mean right atrial pressure; SO2: oxygen saturation; NT-pro-BNP: N-terminal pro-brain natriuretic peptide; CI: cardiac index; RAA: right atrial area; MCV: mean corpuscular volume; RDW: red blood cell distribution width; FT: ferritin; mRASP: modified risk assessment score of PAH; COMPERA: comparative, prospective registry of newly initiated therapies for pulmonary hypertension score; SPAHR: Swedish pulmonary arterial hypertension registry score; FPHR 3p: french pulmonary arterial hypertension registry score, 3 parameters; FPHR 4p: french pulmonary arterial hypertension registry score, 4 parameters. | | | | |
| bNumeric variables: Mann-Whitney U test with r effect size statistic; categorical variables: χ² test with Cramer V effect size statistic; survival: log-rank test. | | | | |

Supplementary Table S3: Results of univariable Cox modeling in the Innsbruck (IBK) and Linz/Vienna (LZ/W) cohort

| **Cohort** | **Variablea** | **Level** | **Order** | **HR, 95% CIb** | **Significance** | **C index, 95% CIc** | **R²** | **IBSd** |
| --- | --- | --- | --- | --- | --- | --- | --- | --- |
| IBK | Age |  | 1 | 2.6 [1.3 - 4.9] | p = 0.016 | 0.68 [0.59 - 0.78] | 0.33 | 0.11 |
|  | 2 | 0.72 [0.36 - 1.5] | ns (p = 0.45) | 0.68 [0.59 - 0.78] | 0.33 | 0.11 |
| SMWD |  | 1 | 0.41 [0.24 - 0.7] | p = 0.0068 | 0.7 [0.59 - 0.8] | 0.3 | 0.17 |
|  | 2 | 0.82 [0.56 - 1.2] | ns (p = 0.4) | 0.7 [0.59 - 0.8] | 0.3 | 0.17 |
| mPAP |  | 1 | 4.6 [2.2 - 9.4] | p < 0.001 | 0.74 [0.65 - 0.83] | 0.38 | 0.14 |
|  | 2 | 0.63 [0.45 - 0.88] | p = 0.02 | 0.74 [0.65 - 0.83] | 0.38 | 0.14 |
| Sex | male |  | 2 [1 - 4] | ns (p = 0.12) | 0.61 [0.52 - 0.7] | 0.069 | 0.18 |
| PVR |  | 1 | 3.5 [1.6 - 7.8] | p = 0.0095 | 0.68 [0.59 - 0.78] | 0.25 | 0.18 |
|  | 2 | 0.73 [0.52 - 1] | ns (p = 0.15) | 0.68 [0.59 - 0.78] | 0.25 | 0.18 |
| PCWP |  | 1 | 1.3 [0.76 - 2.2] | ns (p = 0.45) | 0.51 [0.41 - 0.61] | 0.017 | 0.18 |
|  | 2 | 0.92 [0.71 - 1.2] | ns (p = 0.59) | 0.51 [0.41 - 0.61] | 0.017 | 0.18 |
| Anemia | yes |  | 1.4 [0.59 - 3.2] | ns (p = 0.55) | 0.54 [0.46 - 0.62] | 0.0093 | 0.2 |
| log RDW |  | 1 | 1.8 [0.96 - 3.3] | ns (p = 0.14) | 0.66 [0.55 - 0.77] | 0.2 | 0.18 |
|  | 2 | 0.99 [0.75 - 1.3] | ns (p = 0.92) | 0.66 [0.55 - 0.77] | 0.2 | 0.18 |
| Renal insufficiency | yes |  | 2.5 [1.3 - 5] | p = 0.021 | 0.6 [0.51 - 0.69] | 0.13 | 0.17 |
| log FT |  | 1 | 1.3 [0.88 - 1.9] | ns (p = 0.3) | 0.6 [0.49 - 0.71] | 0.039 | 0.18 |
|  | 2 | 0.93 [0.72 - 1.2] | ns (p = 0.59) | 0.6 [0.49 - 0.71] | 0.039 | 0.18 |
| log TF-Sat |  | 1 | 1 [0.7 - 1.5] | ns (p = 0.9) | 0.58 [0.48 - 0.68] | 0.056 | 0.2 |
|  | 2 | 1.2 [0.97 - 1.4] | ns (p = 0.16) | 0.58 [0.48 - 0.68] | 0.056 | 0.2 |
| MCV |  | 1 | 1.4 [1 - 2.1] | ns (p = 0.12) | 0.54 [0.41 - 0.66] | 0.13 | 0.18 |
|  | 2 | 1.2 [1.1 - 1.3] | p = 0.0087 | 0.54 [0.41 - 0.66] | 0.13 | 0.18 |
| log NT-pro-BNP |  | 1 | 3.9 [2 - 7.7] | p < 0.001 | 0.76 [0.67 - 0.85] | 0.49 | 0.11 |
|  | 2 | 0.58 [0.4 - 0.84] | p = 0.013 | 0.76 [0.67 - 0.85] | 0.49 | 0.11 |
| Percardial effusion | yes |  | 2.1 [0.91 - 4.9] | ns (p = 0.15) | 0.56 [0.48 - 0.63] | 0.048 | 0.19 |
| RAA |  | 1 | 3.6 [1.8 - 7.2] | p = 0.0034 | 0.73 [0.65 - 0.81] | 0.43 | 0.12 |
|  | 2 | 0.46 [0.29 - 0.74] | p = 0.0068 | 0.73 [0.65 - 0.81] | 0.43 | 0.12 |
| CI |  | 1 | 0.31 [0.19 - 0.52] | p < 0.001 | 0.77 [0.68 - 0.85] | 0.4 | 0.15 |
|  | 2 | 1.3 [0.84 - 2] | ns (p = 0.35) | 0.77 [0.68 - 0.85] | 0.4 | 0.15 |
| mRAP |  | 1 | 1.5 [0.93 - 2.5] | ns (p = 0.16) | 0.56 [0.46 - 0.66] | 0.074 | 0.17 |
|  | 2 | 0.92 [0.72 - 1.2] | ns (p = 0.59) | 0.56 [0.46 - 0.66] | 0.074 | 0.17 |
| WHO class | III/IV |  | 1.9 [0.86 - 4] | ns (p = 0.18) | 0.55 [0.46 - 0.64] | 0.05 | 0.19 |
| SO2 | <95 |  | 1.5 [0.76 - 3.1] | ns (p = 0.34) | 0.58 [0.49 - 0.67] | 0.027 | 0.19 |
| LZ/W | Age |  | 1 | 3.8 [1.2 - 12] | ns (p = 0.076) | 0.69 [0.56 - 0.82] | 0.26 | 0.13 |
|  | 2 | 1.4 [0.78 - 2.5] | ns (p = 0.43) | 0.69 [0.56 - 0.82] | 0.26 | 0.13 |
| SMWD |  | 1 | 0.44 [0.24 - 0.8] | p = 0.048 | 0.68 [0.56 - 0.8] | 0.3 | 0.14 |
|  | 2 | 0.93 [0.62 - 1.4] | ns (p = 0.79) | 0.68 [0.56 - 0.8] | 0.3 | 0.14 |
| mPAP |  | 1 | 1.9 [1.1 - 3.4] | ns (p = 0.086) | 0.63 [0.5 - 0.76] | 0.19 | 0.15 |
|  | 2 | 0.81 [0.53 - 1.2] | ns (p = 0.49) | 0.63 [0.5 - 0.76] | 0.19 | 0.15 |
| Sex | male |  | 5.7 [2.2 - 15] | p = 0.011 | 0.73 [0.64 - 0.83] | 0.33 | 0.12 |
| PVR |  | 1 | 5.8 [2 - 16] | p = 0.011 | 0.74 [0.61 - 0.87] | 0.38 | 0.15 |
|  | 2 | 0.54 [0.33 - 0.87] | ns (p = 0.064) | 0.74 [0.61 - 0.87] | 0.38 | 0.15 |
| PCWP |  | 1 | 0.73 [0.31 - 1.7] | ns (p = 0.55) | 0.61 [0.46 - 0.76] | 0.17 | 0.15 |
|  | 2 | 0.48 [0.2 - 1.1] | ns (p = 0.25) | 0.61 [0.46 - 0.76] | 0.17 | 0.15 |
| Anemia | yes |  | 1.9 [0.69 - 5.4] | ns (p = 0.4) | 0.52 [0.43 - 0.6] | 0.042 | 0.15 |
| log RDW |  | 1 | 1.7 [0.91 - 3] | ns (p = 0.25) | 0.61 [0.47 - 0.74] | 0.1 | 0.15 |
|  | 2 | 0.93 [0.69 - 1.2] | ns (p = 0.68) | 0.61 [0.47 - 0.74] | 0.1 | 0.15 |
| Renal insufficiency | yes |  | 1.5 [0.49 - 4.5] | ns (p = 0.55) | 0.49 [0.41 - 0.57] | 0.014 | 0.16 |
| log FT |  | 1 | 1.4 [0.93 - 2.1] | ns (p = 0.26) | 0.58 [0.43 - 0.72] | 0.075 | 0.15 |
|  | 2 | 1.1 [0.88 - 1.4] | ns (p = 0.54) | 0.58 [0.43 - 0.72] | 0.075 | 0.15 |
| log TF-Sat |  | 1 | 0.61 [0.34 - 1.1] | ns (p = 0.25) | 0.61 [0.47 - 0.75] | 0.097 | 0.14 |
|  | 2 | 0.86 [0.59 - 1.3] | ns (p = 0.55) | 0.61 [0.47 - 0.75] | 0.097 | 0.14 |
| MCV |  | 1 | 1.4 [0.82 - 2.5] | ns (p = 0.4) | 0.62 [0.46 - 0.78] | 0.059 | 0.15 |
|  | 2 | 0.84 [0.54 - 1.3] | ns (p = 0.55) | 0.62 [0.46 - 0.78] | 0.059 | 0.15 |
| log NT-pro-BNP |  | 1 | 2.2 [1.3 - 3.6] | p = 0.02 | 0.67 [0.52 - 0.82] | 0.31 | 0.12 |
|  | 2 | 0.96 [0.69 - 1.3] | ns (p = 0.83) | 0.67 [0.52 - 0.82] | 0.31 | 0.12 |
| Percardial effusion | yes |  | 4 [0.89 - 18] | ns (p = 0.22) | 0.53 [0.47 - 0.59] | 0.071 | 0.15 |
| RAA |  | 1 | 1.5 [0.66 - 3.5] | ns (p = 0.49) | 0.65 [0.53 - 0.78] | 0.12 | 0.15 |
|  | 2 | 1 [0.62 - 1.7] | ns (p = 0.95) | 0.65 [0.53 - 0.78] | 0.12 | 0.15 |
| CI |  | 1 | 0.63 [0.35 - 1.1] | ns (p = 0.27) | 0.68 [0.55 - 0.82] | 0.087 | 0.16 |
|  | 2 | 0.84 [0.56 - 1.3] | ns (p = 0.54) | 0.68 [0.55 - 0.82] | 0.087 | 0.16 |
| mRAP |  | 1 | 2.4 [1.2 - 4.8] | ns (p = 0.072) | 0.72 [0.59 - 0.84] | 0.25 | 0.15 |
|  | 2 | 0.79 [0.52 - 1.2] | ns (p = 0.43) | 0.72 [0.59 - 0.84] | 0.25 | 0.15 |
| WHO class | III/IV |  | 5.5 [2 - 15] | p = 0.011 | 0.69 [0.58 - 0.79] | 0.36 | 0.13 |
| SO2 | <95 |  | 1.7 [0.71 - 4.3] | ns (p = 0.4) | 0.56 [0.43 - 0.68] | 0.045 | 0.15 |
| aSMWD: six-minute walking distance; mPAP: mean pulmonary arterial pressure; PVR: pulmonary vascular resistance; PCWP: pulmonary capillary wedge pressure; RDW: red blood cell distribution width; FT: ferritin; TF-Sat: transferrin saturation; MCV: mean corpuscular volume; NT-pro-BNP: N-terminal pro-brain natriuretic peptide; RAA: right atrial area; CI: cardiac index; WHO class: WHO functional class; SO2: oxygen saturation. | | | | | | | | |
| bHR, 95% CI: hazard ratio with 95% confidence intervals. | | | | | | | | |
| cC index, 95% CI: concordance index with 95% confidence intervals. | | | | | | | | |
| dIBS: integrated Brier score. | | | | | | | | |

Supplementary Table S4: Performance of the Elastic Net score and established risk assessment tools at predicting overall PAH survival.

| **Cohort** | **Variablea** | **C index, 95% CIb** | **R²** | **IBSc** |
| --- | --- | --- | --- | --- |
| IBK | ElasticNet | 0.82 [0.75 - 0.89] | 0.66 | 0.098 |
| LASSO ensemble | 0.84 [0.78 - 0.9] | 0.76 | 0.081 |
| mRASP | 0.72 [0.64 - 0.8] | 0.44 | 0.12 |
| COMPERA | 0.7 [0.63 - 0.78] | 0.41 | 0.13 |
| SPAHR | 0.73 [0.66 - 0.79] | 0.51 | 0.11 |
| FPHR 3p | 0.66 [0.58 - 0.74] | 0.37 | 0.14 |
| FPHR 4p | 0.69 [0.61 - 0.77] | 0.36 | 0.13 |
| Reveal Lite | 0.69 [0.61 - 0.77] | 0.4 | 0.11 |
| Reveal 2.0 | 0.68 [0.61 - 0.76] | 0.36 | 0.12 |
| LZ/W | ElasticNet | 0.77 [0.66 - 0.88] | 0.52 | 0.11 |
| LASSO ensemble | 0.74 [0.64 - 0.85] | 0.49 | 0.11 |
| mRASP | 0.67 [0.59 - 0.76] | 0.37 | 0.11 |
| COMPERA | 0.64 [0.54 - 0.74] | 0.23 | 0.14 |
| SPAHR | 0.66 [0.57 - 0.75] | 0.29 | 0.13 |
| FPHR 3p | 0.69 [0.58 - 0.8] | 0.38 | 0.13 |
| FPHR 4p | 0.75 [0.65 - 0.86] | 0.34 | 0.14 |
| aLASSO ensemble: LASSO regression model with the Elastic Net score and established risk assessment scales; mRASP: modified risk assessment score of PAH; COMPERA: comparative, prospective registry of newly initiated therapies for pulmonary hypertension score; SPAHR: Swedish pulmonary arterial hypertension registry score; FPHR 3p: french pulmonary arterial hypertension registry score, 3 parameters; FPHR 4p: french pulmonary arterial hypertension registry score, 4 parameters. | | | | |
| bC index, 95% CI: concordance index with 95% confidence intervals. | | | | |
| cIBS: integrated Brier score. | | | | |

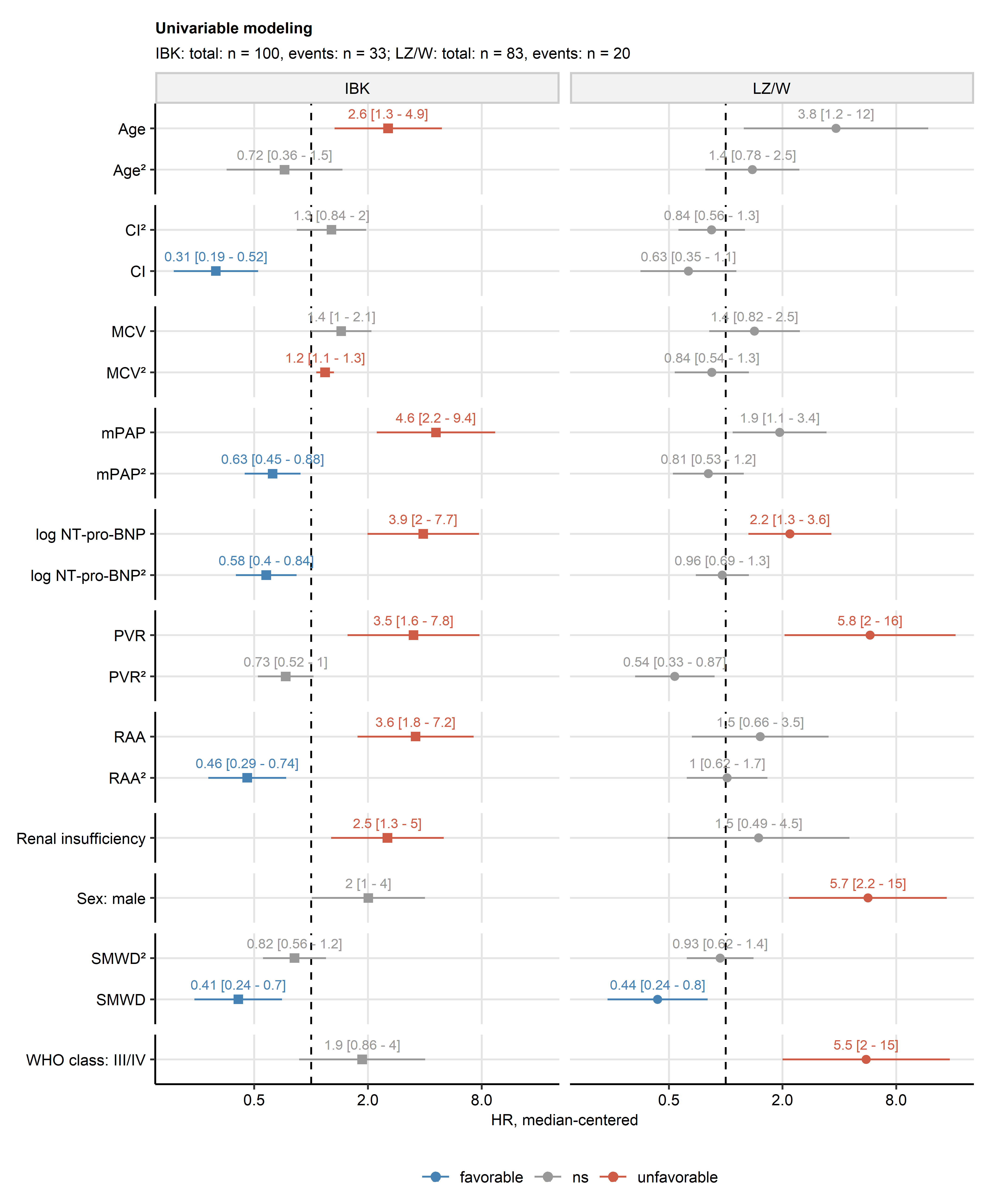
Supplementary Table S5: Characteristic of the participant clusters in the Innsbruck cohort. Numeric variables are presented as medians with interquartile ranges (IQR) and ranges. Categorical variables are presented as percentages and counts within the complete observation set.

| **Variablea** | **Cluster #1** | **Cluster #2** | **Significanceb** | **Effect sizeb** |
| --- | --- | --- | --- | --- |
| Participants, n | 46 | 54 |  |  |
| 3-year mortality | 6.5% (3) | 19% (10) | ns (p = 0.16) | V = 0.18 |
| 5-year mortality | 6.5% (3) | 33% (18) | p = 0.0039 | V = 0.33 |
| Overall mortality | 11% (5) | 52% (28) | p < 0.001 | V = 0.43 |
| Age, years | 58 [IQR: 48 - 68] range: 19 - 80 | 69 [IQR: 65 - 74] range: 26 - 84 | p = 0.0013 | r = 0.34 |
| SMWD, m | 370 [IQR: 310 - 450] range: 120 - 580 | 240 [IQR: 160 - 330] range: 50 - 610 | p < 0.001 | r = 0.47 |
| mPAP, mmHg | 30 [IQR: 27 - 37] range: 26 - 87 | 47 [IQR: 40 - 55] range: 26 - 120 | p < 0.001 | r = 0.56 |
| Sex | female: 78% (36) male: 22% (10) | female: 52% (28) male: 48% (26) | p = 0.017 | V = 0.27 |
| PVR, Wood | 6.8 [IQR: 5.9 - 9.6] range: 3.3 - 38 | 14 [IQR: 11 - 22] range: 4.3 - 43 | p < 0.001 | r = 0.6 |
| PCWP, mmHg | 12 [IQR: 9 - 14] range: 6 - 23 | 14 [IQR: 11 - 19] range: 4 - 32 | p = 0.041 | r = 0.22 |
| Anemia | 15% (7) | 22% (12) | ns (p = 0.53) | V = 0.089 |
| log RDW, % | 2.6 [IQR: 2.6 - 2.7] range: 2.5 - 2.9 | 2.7 [IQR: 2.6 - 2.8] range: 2.6 - 3.1 | p < 0.001 | r = 0.38 |
| Renal insufficiency | 17% (8) | 50% (27) | p = 0.0024 | V = 0.34 |
| log FT, ng/ml | 4 [IQR: 3.3 - 4.7] range: 1.1 - 6.5 | 4.2 [IQR: 3.8 - 4.9] range: 1.8 - 6.5 | ns (p = 0.23) | r = 0.12 |
| log TF-Sat, % | 3 [IQR: 2.7 - 3.4] range: 1.8 - 4.3 | 2.8 [IQR: 2.4 - 3.2] range: 0.69 - 4 | ns (p = 0.16) | r = 0.15 |
| MCV, fl | 88 [IQR: 85 - 90] range: 76 - 96 | 88 [IQR: 86 - 92] range: 58 - 100 | ns (p = 0.33) | r = 0.1 |
| log NT-pro-BNP, pg/ml | 5.1 [IQR: 4.4 - 5.8] range: 3.4 - 7.5 | 7.6 [IQR: 6.9 - 8.1] range: 4.9 - 11 | p < 0.001 | r = 0.78 |
| Percardial effusion | 6.5% (3) | 24% (13) | p = 0.046 | V = 0.24 |
| RAA, cm² | 17 [IQR: 16 - 21] range: 13 - 27 | 24 [IQR: 23 - 27] range: 15 - 34 | p < 0.001 | r = 0.67 |
| CI, NA | 2.7 [IQR: 2.4 - 3] range: 1.8 - 4.2 | 2 [IQR: 1.9 - 2.3] range: 1.6 - 3.5 | p < 0.001 | r = 0.6 |
| mRAP, mmHg | 8 [IQR: 6 - 12] range: 2 - 18 | 11 [IQR: 8 - 14] range: 2 - 26 | p = 0.023 | r = 0.24 |
| WHO class | I/II: 48% (22) III/IV: 52% (24) | I/II: 31% (17) III/IV: 69% (37) | ns (p = 0.16) | V = 0.17 |
| SO2 | ≥95: 57% (26) <95: 43% (20) | ≥95: 39% (21) <95: 61% (33) | ns (p = 0.15) | V = 0.18 |
| mRASP, risk strata | low: 65% (30) int: 35% (16) high: 0% (0) | low: 3.7% (2) int: 67% (36) high: 30% (16) | p < 0.001 | V = 0.69 |
| COMPERA, risk strata | low: 46% (21) int: 54% (25) high: 0% (0) | low: 1.9% (1) int: 72% (39) high: 26% (14) | p < 0.001 | V = 0.59 |
| SPAHR, risk strata | low: 50% (23) int: 50% (23) high: 0% (0) | low: 1.9% (1) int: 78% (42) high: 20% (11) | p < 0.001 | V = 0.6 |
| FPHR 3p, number of risk factors | 0: 17% (8) 1: 30% (14) 2: 35% (16) 3: 17% (8) | 0: 0% (0) 1: 7.4% (4) 2: 26% (14) 3: 67% (36) | p < 0.001 | V = 0.56 |
| FPHR 4p, number of risk factors | 0: 6.5% (3) 1: 17% (8) 2: 37% (17) 3: 30% (14) 4: 8.7% (4) | 0: 0% (0) 1: 3.7% (2) 2: 17% (9) 3: 31% (17) 4: 48% (26) | p < 0.001 | V = 0.5 |
| Reveal Lite, risk strata | low: 70% (32) int: 17% (8) high: 13% (6) | low: 5.6% (3) int: 11% (6) high: 83% (45) | p < 0.001 | V = 0.73 |
| Reveal 2.0, risk strata | low: 65% (30) int: 17% (8) high: 17% (8) | low: 3.7% (2) int: 11% (6) high: 85% (46) | p < 0.001 | V = 0.72 |
| aSMWD: six-minute walking distance; mPAP: mean pulmonary arterial pressure; PVR: pulmonary vascular resistance; PCWP: pulmonary capillary wedge pressure; RDW: red blood cell distribution width; FT: ferritin; TF-Sat: transferrin saturation; MCV: mean corpuscular volume; NT-pro-BNP: N-terminal pro-brain natriuretic peptide; RAA: right atrial area; CI: cardiac index; WHO class: WHO functional class; SO2: oxygen saturation, mRASP: modified risk assessment score of PAH; COMPERA: comparative, prospective registry of newly initiated therapies for pulmonary hypertension score; SPAHR: Swedish pulmonary arterial hypertension registry score; FPHR 3p: french pulmonary arterial hypertension registry score, 3 parameters; FPHR 4p: french pulmonary arterial hypertension registry score, 4 parameters. | | | | |
| bNumeric variables: Mann-Whitney U test with r effect size statistic; categorical variables: χ² test with Cramer V effect size statistic. | | | | |

Supplementary Table S6: Characteristic of the participant clusters in the Linz/Vienna cohort. Numeric variables are presented as medians with interquartile ranges (IQR) and ranges. Categorical variables are presented as percentages and counts within the complete observation set.

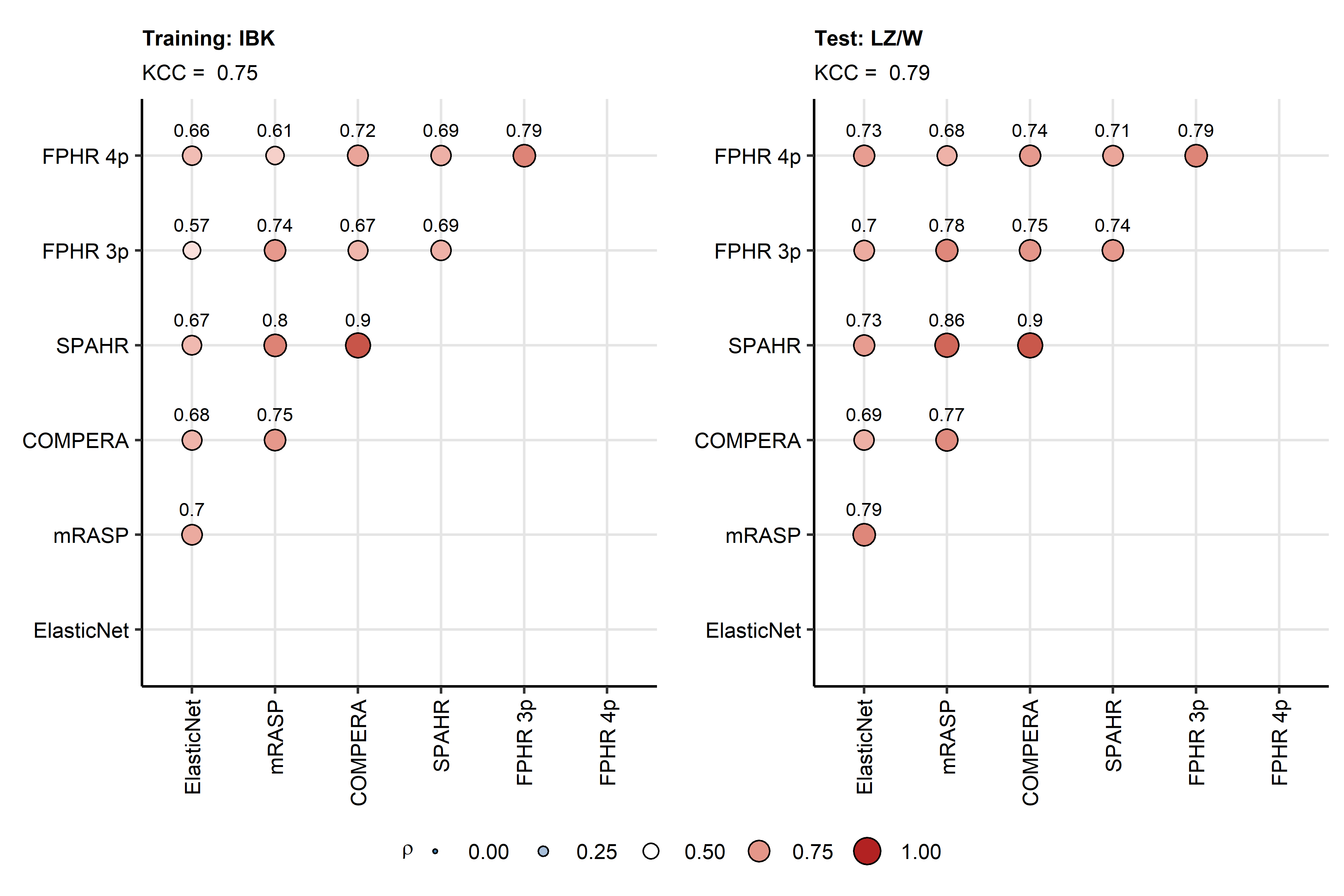
| **Variablea** | **Cluster #1** | **Cluster #2** | **Significanceb** | **Effect sizeb** |
| --- | --- | --- | --- | --- |
| Participants, n | 35 | 48 |  |  |
| 3-year mortality | 5.7% (2) | 15% (7) | ns (p = 0.38) | V = 0.14 |
| 5-year mortality | 5.7% (2) | 19% (9) | ns (p = 0.21) | V = 0.19 |
| Overall mortality | 14% (5) | 31% (15) | ns (p = 0.17) | V = 0.2 |
| Age, years | 63 [IQR: 46 - 70] range: 23 - 81 | 71 [IQR: 64 - 75] range: 26 - 82 | p = 0.0043 | r = 0.34 |
| SMWD, m | 440 [IQR: 350 - 510] range: 190 - 620 | 320 [IQR: 180 - 370] range: 50 - 580 | p < 0.001 | r = 0.49 |
| mPAP, mmHg | 34 [IQR: 28 - 39] range: 18 - 57 | 44 [IQR: 35 - 50] range: 28 - 67 | p < 0.001 | r = 0.48 |
| Sex | female: 80% (28) male: 20% (7) | female: 56% (27) male: 44% (21) | ns (p = 0.069) | V = 0.25 |
| PVR, Wood | 3.7 [IQR: 3 - 5] range: 1.4 - 10 | 6.6 [IQR: 4.9 - 9] range: 2.3 - 20 | p < 0.001 | r = 0.47 |
| PCWP, mmHg | 10 [IQR: 8 - 12] range: 1 - 27 | 12 [IQR: 8.5 - 14] range: 1 - 25 | ns (p = 0.22) | r = 0.15 |
| Anemia | 5.7% (2) | 25% (12) | ns (p = 0.069) | V = 0.25 |
| log RDW, % | 2.7 [IQR: 2.6 - 2.8] range: 2.5 - 2.8 | 2.8 [IQR: 2.7 - 2.8] range: 2.5 - 3.1 | p = 0.0077 | r = 0.32 |
| Renal insufficiency | 11% (4) | 23% (11) | ns (p = 0.34) | V = 0.15 |
| log FT, ng/ml | 4.2 [IQR: 3 - 4.8] range: 1.9 - 5.7 | 4.4 [IQR: 3.7 - 5] range: 2.5 - 7.1 | ns (p = 0.1) | r = 0.2 |
| log TF-Sat, % | 3.1 [IQR: 2.7 - 3.3] range: 1.6 - 4.1 | 2.9 [IQR: 2.4 - 3.4] range: 0.69 - 4.5 | ns (p = 0.38) | r = 0.1 |
| MCV, fl | 89 [IQR: 84 - 93] range: 78 - 110 | 90 [IQR: 87 - 93] range: 76 - 100 | ns (p = 0.49) | r = 0.077 |
| log NT-pro-BNP, pg/ml | 5.3 [IQR: 4.9 - 6.1] range: 3.2 - 7 | 7.3 [IQR: 6.7 - 7.9] range: 4.8 - 10 | p < 0.001 | r = 0.66 |
| Percardial effusion | 0% (0) | 6.2% (3) | ns (p = 0.38) | V = 0.17 |
| RAA, cm² | 17 [IQR: 15 - 17] range: 13 - 20 | 22 [IQR: 19 - 25] range: 15 - 30 | p < 0.001 | r = 0.69 |
| CI, NA | 2.8 [IQR: 2.5 - 3.1] range: 1.8 - 3.8 | 2.4 [IQR: 2.1 - 2.8] range: 1.4 - 3.6 | p = 0.0086 | r = 0.31 |
| mRAP, mmHg | 3 [IQR: 1 - 5] range: 0 - 16 | 8 [IQR: 5.8 - 9.2] range: 1 - 20 | p < 0.001 | r = 0.55 |
| WHO class | I/II: 83% (29) III/IV: 17% (6) | I/II: 31% (15) III/IV: 69% (33) | p < 0.001 | V = 0.51 |
| SO2 | ≥95: 60% (21) <95: 40% (14) | ≥95: 40% (19) <95: 60% (29) | ns (p = 0.15) | V = 0.2 |
| mRASP, risk strata | low: 86% (30) int: 14% (5) high: 0% (0) | low: 4.2% (2) int: 81% (39) high: 15% (7) | p < 0.001 | V = 0.83 |
| COMPERA, risk strata | low: 83% (29) int: 17% (6) high: 0% (0) | low: 6.2% (3) int: 90% (43) high: 4.2% (2) | p < 0.001 | V = 0.78 |
| SPAHR, risk strata | low: 80% (28) int: 20% (7) high: 0% (0) | low: 0% (0) int: 98% (47) high: 2.1% (1) | p < 0.001 | V = 0.84 |
| FPHR 3p, number of risk factors | 0: 31% (11) 1: 37% (13) 2: 23% (8) 3: 8.6% (3) | 0: 0% (0) 1: 8.3% (4) 2: 35% (17) 3: 56% (27) | p < 0.001 | V = 0.67 |
| FPHR 4p, number of risk factors | 0: 31% (11) 1: 40% (14) 2: 23% (8) 3: 5.7% (2) 4: 0% (0) | 0: 0% (0) 1: 15% (7) 2: 27% (13) 3: 42% (20) 4: 17% (8) | p < 0.001 | V = 0.66 |
| aSMWD: six-minute walking distance; mPAP: mean pulmonary arterial pressure; PVR: pulmonary vascular resistance; PCWP: pulmonary capillary wedge pressure; RDW: red blood cell distribution width; FT: ferritin; TF-Sat: transferrin saturation; MCV: mean corpuscular volume; NT-pro-BNP: N-terminal pro-brain natriuretic peptide; RAA: right atrial area; CI: cardiac index; WHO class: WHO functional class; SO2: oxygen saturation, mRASP: modified risk assessment score of PAH; COMPERA: comparative, prospective registry of newly initiated therapies for pulmonary hypertension score; SPAHR: Swedish pulmonary arterial hypertension registry score; FPHR 3p: french pulmonary arterial hypertension registry score, 3 parameters; FPHR 4p: french pulmonary arterial hypertension registry score, 4 parameters. | | | | |
| bNumeric variables: Mann-Whitney U test with r effect size statistic; categorical variables: χ² test with Cramer V effect size statistic. | | | | |

# Supplementary Figures



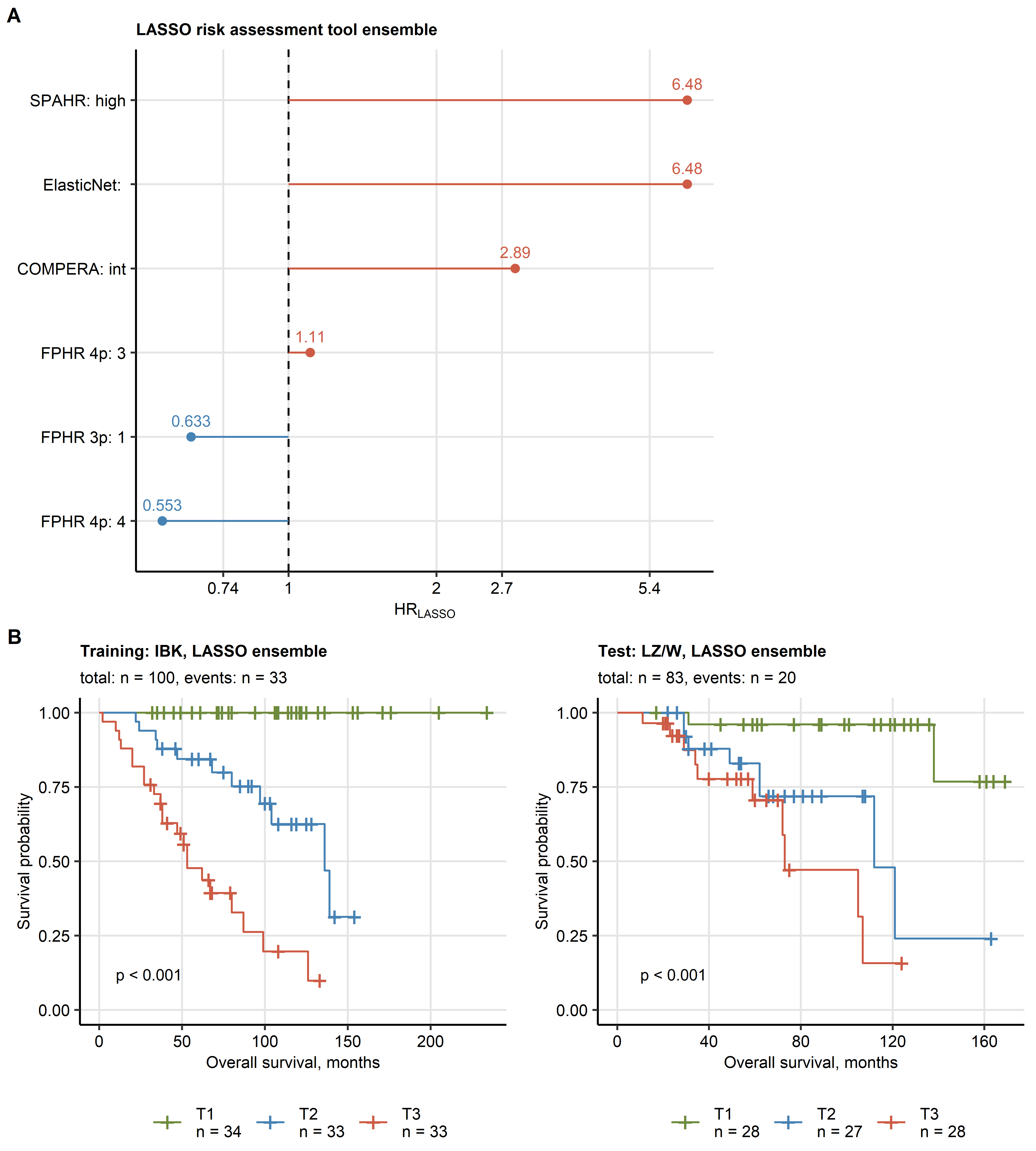
**Supplementary Figure S1. Univariable Cox proportional hazard modeling.**

*Association of candidate risk factors (Supplementary Table S1) with overall survival was investigated with a series of univariable Cox proportional hazard models. Numeric independent variables were normalized and median-centered and their first and second order terms included in the models. Hazard ratio (HR) estimate significance was determined by Wald Z test and adjusted for multiple testing with Benjamini-Hochberg method. HR values with 95 confidence intervals for variables significantly associated with the survival in at least one Innsbruck (IBK) or Linz/Vienna cohort (LZ/W) were presented in a Forest plot. Numbers of complete observations and mortality are indicated under the plot.* *CI: cardiac index; MCV: mean corpuscular volume; mPAP: mean pulmonary arterial pressure; NT-pro-BNP: N terminal pro brain natriuretic peptide; PVR: pulmonary vascular resistance; RAA: right atrial area; SMWD: six minute walking distance.*



**Supplementary Figure S2. Correlation of the Elastic Net score and established risk assessment scores.**

*Established PAH risk scales (mRASP: modified risk assessment score of PAH; COMPERA: comparative, prospective registry of newly initiated therapies for pulmonary hypertension score; SPAHR: Swedish pulmonary arterial hypertension registry score; FPHR 3p: french pulmonary arterial hypertension registry score; FPHR 4p: french pulmonary arterial hypertension registry score, 4 parameters) were displayed in ordinal scales (mRASP, COMPERA, SPAHR: low: 1, intermediate [int]: 2, high risk: 3, FPHR scales: number of risk factors). Their correlation with the newly developed ELASTIC Net score was assessed by Spearman’s test corrected for multiple testing with Benjamini-Hochberg method. Consistency of the entire risk assessment battery was investigated with Kendall’s coefficient of concordance (KCC). Correlation coefficients are presented as points. Point size and color codes for the value. Points are labeled with their values. KCC values are displayed in the plot captions.*

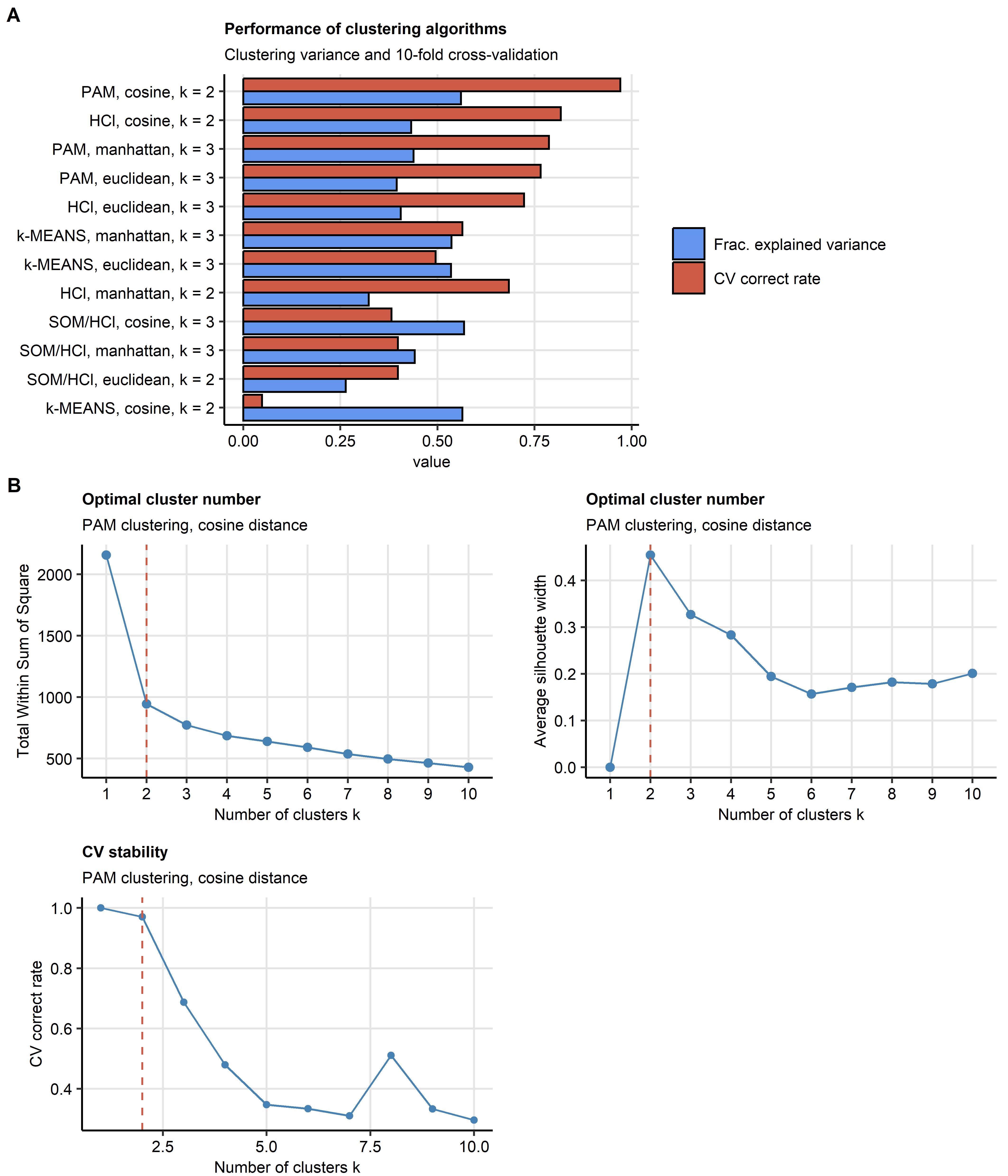


**Supplementary Figure S3. Development of a LASSO ensemble model including the Elastic Net score and established PAH risk assessment tools.**

*The LASSO Cox regression ensemble model was trained in the Innsbruck cohort (IBK) with the Elastic Net score and established PAH risk scales (mRASP: modified risk assessment score of PAH; COMPERA: comparative, prospective registry of newly initiated therapies for pulmonary hypertension score; SPAHR: Swedish pulmonary arterial hypertension registry score; FPHR 3p: french pulmonary arterial hypertension registry score; FPHR 4p: french pulmonary arterial hypertension registry score, 4 parameters) as explanatory variables. The Elastic Net score was expressed as a numeric variables, the other risk tools were included in the model as categorical variables (mRASP, COMPERA, SPAHR: low, intermediate [int], high risk: 3, FPHR scales: number of risk factors). Numbers of complete observations and mortality is indicated in B.*

*(A) Non-zero model coefficients represented as hazard ratios (HR). Plot points are labeled with their HR values.*

*(B) Association of overall survival with the LASSO ensemble linear prediction score in the training IBK and test Linz/Vienna (LZ/W) cohort was assessed by Kaplan-Meier analysis. Significance of the survival differences in the study participants stratified by the linear predictor score tertiles (T1: 0 - 33, T2: 34 - 66, T3: 66 - 100 percentile) was determined by log-rank test adjusted for multiple testing with Benjamini-Hochberg method. P values are shown in the plots, numbers of complete observations and mortality are indicated in the plot captions.*

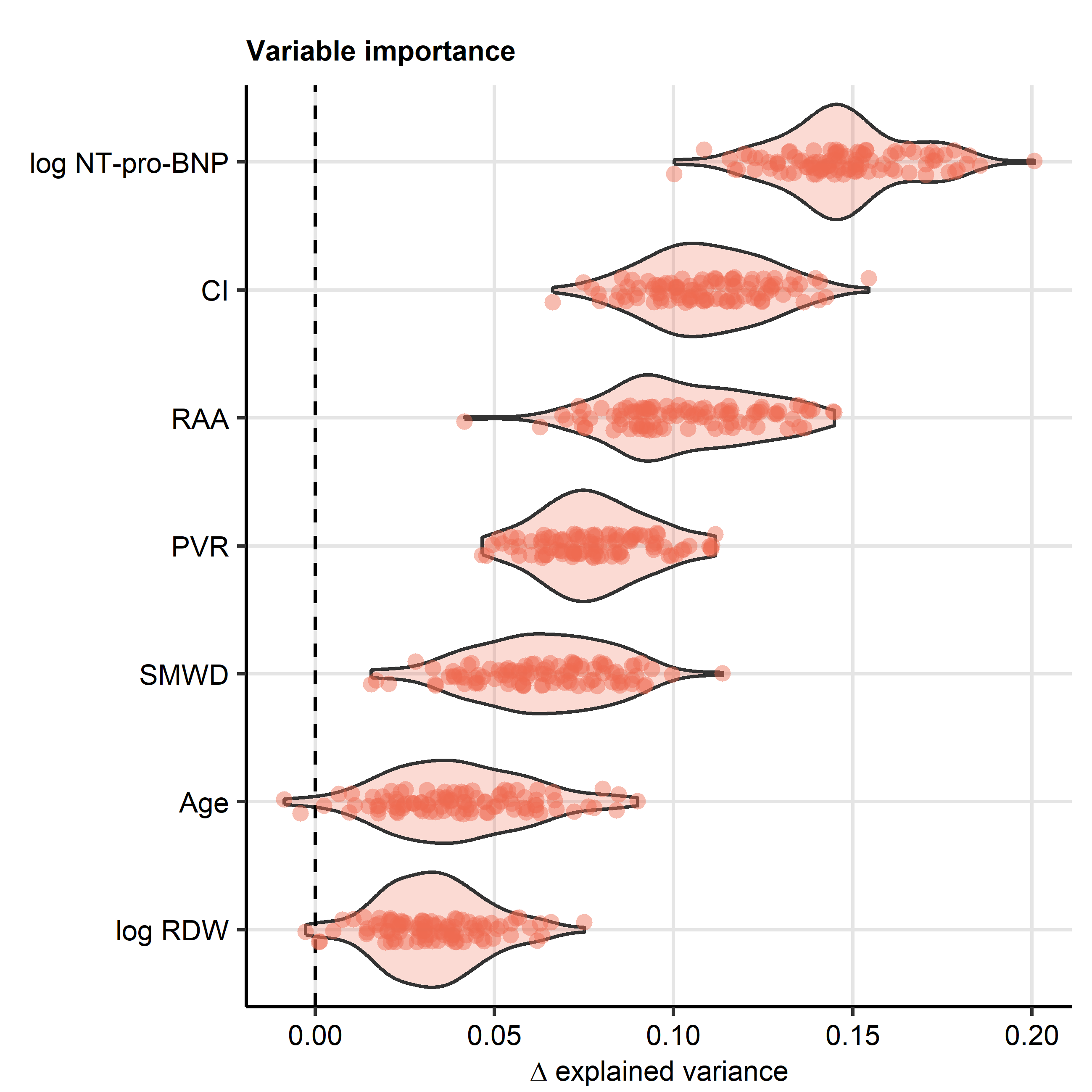


**Supplementary Figure S4. Development of the PAH clusters.**

*Clustering of the training Innsbruck (IBK) cohort participants in respect to the survival-associated factors identified by elastic-net modeling (Figure 2A) was investigated by PAM (partitioning around medoids) algorithm and cosine distance measure.*

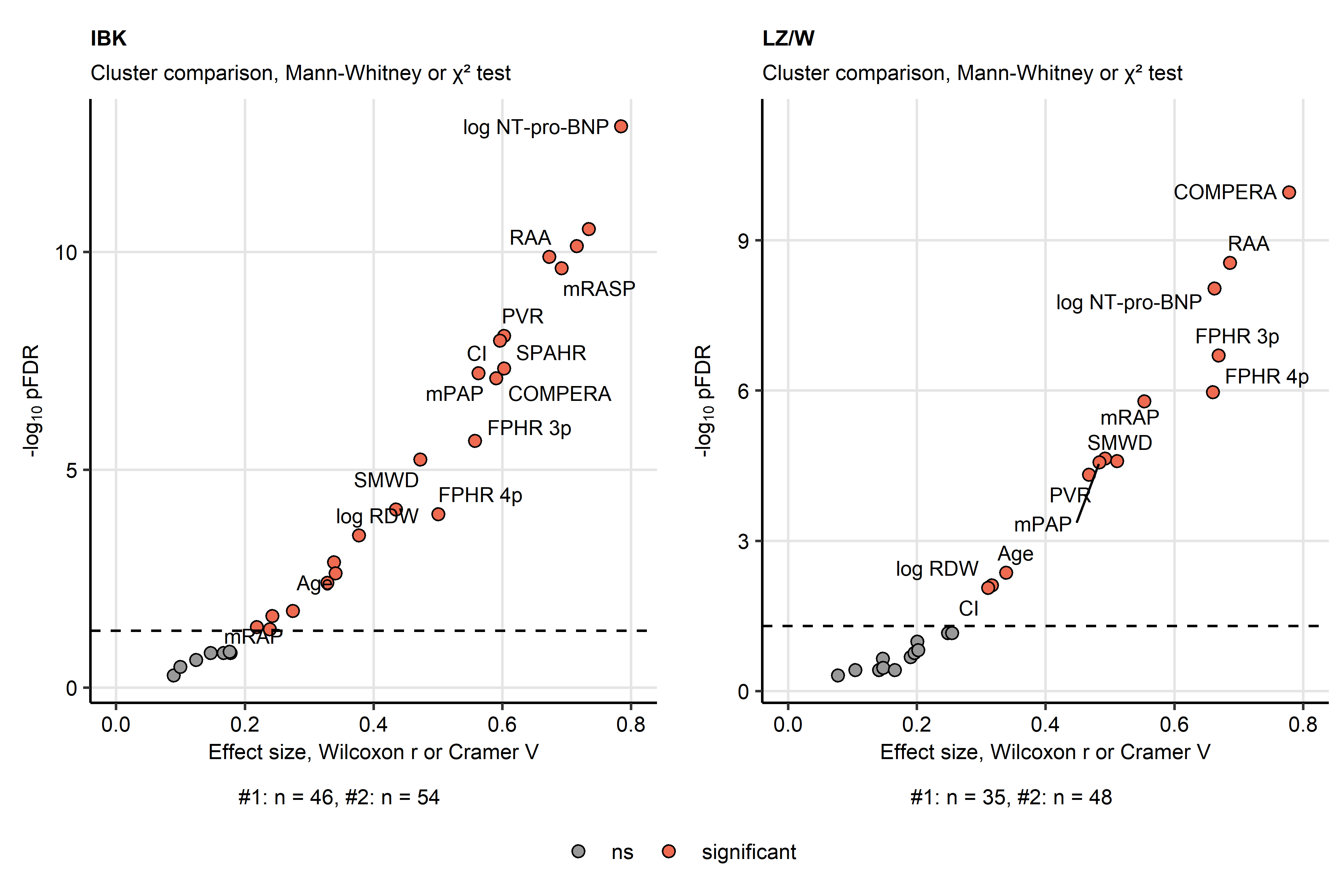
*(A) Comparison of the ‘explained’ clustering variance (ratio of between-cluster sum of squares to total sum of squares) and correct cluster assignment rate in 10-fold cross-validation (CV) for clustering of the IBK cohort with various algorithms (PAM, HCl: hierarchical clustering, k-means and SOM/HCl: combined self-organizing map/hierarchical clustering) and distance statistics (Euclidean, Manhattan and cosine distance) and cluster numbers k. Number of clusters k for each clustering procedure was determined by the bend of the curve of within-cluster sum of squares and maximal mean silhouette statistic. Algorithm, distance measure and the cluster number k are indicated in the Y axis. Note the superior ‘explained’ variance fraction and CV performance of the PAM algorithm/cosine distance procedure.*

*(B) Determination of the optimal cluster number for the PAM/cosine distance procedure by the bend of the total within-cluster sum of squares curve, peak of the mean silhouette statistic and correct cluster assignment rate in 10-fold CV. The dashed vertical line indicates the chosen number of PAM clusters.*



**Supplementary Figure S5. Permutation importance of the variables used for development of the PAH clusters.**

*Importance of particular clustering factors was determined by comparing the fractions of ‘explained’ clustering variance (ratio of between-cluster sum of squares to total sum of squares) between the original clustering structure and clustering objects with randomly re-shuffled clustering variables in the Innsbruck (IBK) cohort. The procedure was repeated 100 times. Differences in explained clustering variances () are presented in violin plots. Values for single procedure repetitions are depicted as points.*



**Supplementary Figure S6. Differences in study variables between the PAH clusters.** *Training Innsbruck (IBK) cohort participants were clustered as presented in Figure 4 and Supplementary Figure S4. Cluster assignment of the test Linz/Vienna (LZ/W) cohort participants was accomplished by k-nearest neighbor label propagation procedure. Differences in the study variables (Supplementary Table S1) between the clusters were determined by Mann-Whitney test with r effect size statistic or by test with Cramer V effect size statistic for numeric and categorical features, respectively. P values were adjusted for multiple testing with Benjamini-Hochberg method (pFDR).* *Significance (pFDR) and effect size are presented in scatter plots. Each point represents one study parameter, parameters significantly different between the clusters are highlighted in red. Parameters found significant in both cohorts are labeled with their names. The significance cutoff is depicted as a dashed line. Numbers of participants assigned to the clusters are presented under the plots.* *CI: cardiac index; mPAP: mean pulmonary arterial pressure; NT-pro-BNP: N terminal pro brain natriuretic peptide; PVR: pulmonary vascular resistance; RAA: right atrial area; SMWD: six minute walking distance; mPAP: mean pulmonary arterial pressure; RDW: red blood cell distribution width; mRAP: mean right atrial pressure; FPHR: French pulmonary hypertension register; SPAHR: Swedish pulmonary arterial hypertension register; COMPERA: comparative, prospective registry of newly initiated therapies for pulmonary hypertension; mRASP: modified risk assessment score of PAH.*

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