Three distinct patterns of mental health response following accidents in mountain sports – a follow-up study of individuals treated at a tertiary trauma center

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# Abstract

**Background:** Generally the restorative effect of physical activity in alpine environments on mental and physical health is recognized. However, a risk of accidents is inherent to every sport, which is also true for physical activity performed in an alpine environment. We aimed to characterize mental health in individuals following mountain sport accidents requiring professional medical management.

**Methods:** Adult victims of mountain sport accidents treated at the hospital of the Medical University of Innsbruck (Austria) between 2018 and 2020 completed a cross-sectional survey at a median of 44 months following the admission (n = 307). Symptoms of post-traumatic stress disorder (PTSD, PCL-5), anxiety, depression, and somatization (PHQ), resilience (RS-13), sense of coherence (SOC-9L), post-traumatic growth (PTGI) and quality of life (EUROHIS-QOL), as well as sociodemographic and clinical information were obtained from an online survey and extracted from electronic health records. Mental health outcome patterns were investigated by semi-supervised medoid clustering and modeled by machine learning.

**Results:** Symptoms of PTSD were observed in 19% of participants. Three comparably sized subsets of participants were identified: a (1) neutral, (2) post-traumatic growth and (3) post-traumatic stress cluster. The post-traumatic stress cluster was characterized by high prevalence of symptoms of mental disorder, low resilience, low sense of coherence and low quality of life and associated with younger age, high frequency of pre-existing mental disorders and persisting physical health consequences of the accident. Individuals in this cluster self-reported a need for psychological or psychiatric support following the accident and more cautious behavior during mountain sports since the accident. Prediction of cluster assignment using machine learning based on 40 variables available during acute medical treatment of accident victims was only possible to a limited extend.

**Conclusion:** A subset of individuals develop symptoms of mental health disorders including symptoms of PTSD when assessed at least 6 months after the mountain sport accident. Since early identification of these vulnerable patients remains challenging, psychoeducational measures for all patients and low-threshold access to psychological or psychiatric support is key for a successful interdisciplinary management of victims of mountain sport accidents.

# Introduction

Physical activity in mountains, such as hiking, climbing, alpine skiing or ski touring, are gaining increasing popularity, a trend that has been accelerated by the COVID-19 pandemic (1). Outdoor physical activity in general and mountain sports in particular exert beneficial effects on physical and mental health (2–4). However, mountain sports are also associated with the risk of accidents and injury (5,6), which could consequently lead to development of post-traumatic stress disorder (PTSD) or other stress-related mental disorders.

Diagnostic symptoms of PTSD include re-experiencing (such as involuntary, intrusive memories, dreams or flashbacks of the traumatic event; criterion B), avoidance (such as avoiding memories, thoughts or external reminders related to the traumatic event; criterion C), negative alterations in cognition and mood (criterion D) and hyper-arousal (such as irritability, problems with sleep or concentration; criterion E) (7), which can occur following self-experiencing or witnessing of a traumatic event such as actual or threatened death, injury, physical or sexual violence. PTSD may co-occur with symptoms of depression and anxiety and substance use disorders (8,9). Although most individuals have experienced at least one traumatic event during their lifetime (10), the prevalence of PTSD in the general European population has been estimated between 0.38 to 6.67% (11). Yet, up to 26% of acute physical trauma patients have been reported to develop PTSD (12).

So far it has proven difficult to reliably identify predictors of PTSD development in accident victims. While some recent studies on specific populations have identified risk factors, such as pre-existing psychological distress or mental disorders (13), others have focused on the arousal and stress reaction during acute trauma and peri-traumatic period as candidate predictors (15). Overall, combination of factors seem to yield more promising results compared to single variables (14). These results require validation and further generalization before they can be implemented into routine care (16,17).

Resilience, i.e. the ability to recover from adversities can protect an individual from developing PTSD and can help to explain why most people do not develop PTSD despite the high prevalence of traumatic events in the general population (18,19). Generally individuals who perform mountain sports display high resilience values (2). Post-traumatic growth can occur in individuals struggling to overcome the consequences of a traumatic event and describes a psychological adjustment reaction with strengthened positive perceptions of self, others, one’s life, and meaning of the event (20). Although both resilience and post-traumatic growth are salutogenic concepts their exact link is still under investigation (21).

With the increasing popularity of mountain sports, it is essential to characterize the mental health consequences of accidents during mountain sports. Prevention of PTSD or other mental health problems following an accident is key to preserve the positive mental health effects of physical activity in mountain environments despite the inherent accident risk. Available data on the consequences of accidents during mountain sports on mental health are scarce. Prior studies were performed on mountain guides, rescuers or avalanche accident survivors and mostly lack clinical data (22–27). Here we (1) aimed to characterize the current sample of 307 individuals treated at the Hospital of the Medical University of Innsbruck (Austria) after a mountain sport accident by analyzing mental health readouts such as symptoms of PTSD, anxiety, depression, and somatization, resilience, sense of coherence, post-traumatic growth and quality of life in by means of semi-supervised clustering. (2) We searched for early sociodemographic and clinical predictors of mental health impairment following a mountain sport accident with a machine learning approach.

# Methods

## Ethics

The study was conducted in accordance with the Declaration of Helsinki and European data policies. All participants gave electronically signed written informed consent to participate. Participants’ data were processed in anonymized form. The study protocol was approved by the ethics committee of the Medical University of Innsbruck (approval number: 1472/2020).

## Participants

Individuals treated following a mountain sport accident at the Department for Orthopedics and Traumatology at the Hospital of the Medical University of Innsbruck between 1st January 2018 and 31st December 2020 were screened for participation. Individuals fulfilling the inclusion criteria: hospital admission at least 6 months prior to the start of the study, age 18 years, residence in a German-speaking region, and proficiency in German (n = 4559) were invited to participate in the online study survey via conventional mail. Out of the invited subjects, 387 completed the survey. Surveys of 307 individuals with the complete psychometry data were analyzed (overall response rate: 6.7%, **Figure 1**, **Supplementary Tables S1** - **S3**).

## Procedures

Details on study procedures and variables are provided in **Supplementary Methods** and **Supplementary Table S1**.

Sociodemographic (e.g. age, sex, education, profession, income, prior mountain sport accidents and prior traumatic events), accident- and recovery-related (e.g. mode of rescue, psychological support, persistent physical health consequences) and psychometric variables were recorded with a cross-sectional online survey.

The psychometric battery consisted of German versions of assessment tools for symptoms of PTSD (PCL-5: PTSD checklist for DSM-5) (28), anxiety (GAD-7: 7-item general anxiety disorder scale) (29), depression (PHQ-9: 9-item patient health questionnaire for depressive symptoms) (30), panic (PHQ-panic module) (30), somatization (PHQ-15: 15-item patient health questionnaire of common somatic symptoms as a substitute for somatization) (31), resilience (RS13: 13-item resilience scale) (32), sense of coherence (SOC-9L: Leipzig 9-item sense of coherence questionnaire) (33), quality of life (EUROHIS-QOL 8: 8-item EUROHIS project quality of life scale) (34), and post-traumatic growth (PTGI: post-traumatic growth inventory) (20). The tools displayed good-to-excellent consistency (35) (McDonald’s > 0.8, **Supplementary Table S2** and **S3**).

Clinically significant symptoms of anxiety (GAD-7 10), depression (PHQ-9 11) (36), somatization (PHQ-15 11) (31) and resilience classes (low: RS-13 0 - 65, moderate: 66 - 72, high: 73) (32) were defined with published cutoffs.

Separate scores were calculated for domains B, C, D and E of the PCL-5 tool along with the total score being the sum of all items. Each PCL-5 item was scored as 0: not at all, 1: a little bit, 2: moderate, 3: quite a bit and 4: extremely. Participants positive for the B domain or C domain PTSD symptoms were identified by at least one item per domain scored with ‘moderate’ or higher. Participants positive for the D or E domain PTSD symptoms were identified by at least two items per domain scored with ‘moderate’ or higher. Significant PTSD symptoms were assumed in participants screened positive for at least one of the B, C, D or E PCL-5 domains (28).

Traumatic events prior to the accident were assessed with the DIA-X tool (37). Flashback frequency was surveyed as none, > 1/year and > 1/month. Smoking was surveyed as a single yes/no question. Problematic alcohol use was investigated with the CAGE tool (38). Data on the type of the mountain sport activity during the accident, injury diagnosis, severity (AIS: abbreviated injury scale) (39) and body location, hospitalization, surgery and number of surgical ICD-10 diagnoses were extracted from electronic patient records.

The training and test participant subsets (3:1 size ratio) were obtained by random splitting which minimized differences in in sociodemographic, medical history, clinical and accident- and injury-related variables assessed by Gower distance (**Figure 1**). The training subset size (n = 230) was sufficient for clustering analysis as assessed by Hopkins metrics of 0.74 indicative of substantial spontaneous clustering tendency (40).

## Analysis endpoints

The primary analysis endpoint was assessment of mental health following an accident in mountain sports by clustering using standardized psychometric tools for symptoms of PTSD, anxiety, depression, panic, somatization, resilience, sense of coherence, post-traumatic growth, and quality of life. The secondary analysis endpoints were (1) to compare demographic, socioeconomic and clinical characteristics between the different mental health clusters of mountain accident victims and (2) to identify patients at risk of mental disorders based on a set of explanatory variables available during acute medical treatment after mountain sport accidents.

## Statistical analysis

Details on statistical analysis are provided in **Supplementary Methods.**

Statistical analysis was done with R version 4.2.3. Numeric variables are displayed as medians with interquartile ranges. Categorical variables are presented as percentages and counts. Differences in numeric variables were analyzed by Mann-Whitney or Kruskal-Wallis test with r or effect size statistics. Differences in categorical variable distribution were assessed by test with Cramer V effect size statistic (41).

The training subset was clustered in respect to normalized median-centered psychometric scores by partition around medoids with cosine distance (42,43). This algorithm had a good explanatory performance (ratio of between-cluster sum of squares to total sum of squares), separation between clusters (average silhouette width) (44) and the superior accuracy in 10-fold cross-validation (45) as compared with the hierarchical clustering and KMEANS algorithms. The k = 3 cluster number choice was based on the bend of within-cluster sum of squares curve and maximal mean silhouette statistic (40,44). The training subset observations were assigned to the clusters with an inverse distance-weighted 27-nearest neighbor classifier.

Multi-parameter machine learning classifiers of the mental health cluster assignment were trained with the canonical random forest algorithm (46,47), regularized neural network with a single hidden layer (48), support vector machine algorithm with radial kernel (49,50), recursive partitioning (51,52), multinomial elastic net regression (53,54), and conditional random forest algorithm (55–57). The cluster assignment was predicted for the test subset and the classifiers’ performance at predicting the cluster assignment was assessed by the accuracy and statistics, Brier score as well as sensitivity and specificity of predicted assignment to the post-traumatic stress cluster (58–60).

# Results

## Characteristic of the study cohort

Surveys with complete psychometric data from 307 individuals were analyzed (**Figure 1**). The median time between the trauma center admission after the accident and the survey completion was 1343 days (interquartile range: 804 - 1441 days). As compared with the analyzed participants, patients who did not respond to the study invitation were characterized by a higher frequency of biking accidents, lower rate of alpine ski/snowboard accidents, less severe injury, lower hospitalization and surgery rates, and fewer surgical diagnoses (**Supplementary Table S4**). The survey participants excluded from the analysis due to missingness of psychometric data had significantly lower income, less severe injuries, more frequent injuries of the upper limbs, less frequently required hospitalization or surgery, and suffered less frequently from persistent physical health consequences of the accident than the analyzed participants (**Supplementary Table S5**). Effect sizes of the differences between the included and excluded participants were weak (V 0.22 or r 0.15).

The analyzed participants were predominantly middle-aged (median: 51, interquartile range: 33 to 60 years) and 45% of them were females. The vast majority had a secondary or tertiary education (83%) and were professionally active (68%). Less than 8% of participants worked in a search and rescue or mountain sport profession. Annual household incomes of 45000 Euro were reported by 42% of participants. Less than 10% of participants were smokers or at risk of problematic alcohol use defined as CAGE 2. Pre-existing physical disorders were reported by 15% of participants with cardiovascular, neurological and metabolic illness being the most frequent. Mental disorders diagnosed by a health care professional prior to the accident affected 5.2% of the cohort with affective (2.3%) and somatoform disorders (1.6%) as the leading pre-existing mental disorders. Four of ten participants had experienced or witnessed a traumatic event prior to the accident, 10.4% participants had been exposed to two or more traumatic events (**Table 1**). Mountain sport accidents in the past were reported by 38% of participants. Most of the investigated accidents occurred during alpine skiing, snowboarding or cross-country skiing (64%) followed by biking (16%) and classical mountain sports such as climbing, hiking, mountaineering or back-country skiing (14%).

One-third of participants were alone during the accident (32%) and, in most cases, were the only person responsible for the accident. Professional rescue service was involved in 29% of the accidents. In 35% of participants the injury severity was moderate (AIS 2) and in 28% severe-to-critical (AIS 3). Limb injuries were the most common followed by injuries of the head and face (**Supplementary Figure S1A**). Hospitalization and surgery rates were 26% and 14%, respectively. Psychological or psychiatric support after the accident was provided to 9.1% of individuals. A subset of participants (7.5%), who had not received psychological or psychiatric support, declared a need for psychological or psychiatric support following the accident. Persisting physical consequences of the accident were reported by 37% of participants and flashbacks during mountain sport at the time of the questionnaire completion were observed in 40% of the cohort. Although most individuals returned to the same mountain sport following the accident (85%), 65% of all participants described their behavior during mountain sport as more cautious (**Table 2**).

At least one diagnostic criterion of PTSD defined as positive scoring of the B, C, D or E domains of the PCL-5 tool was met by 19% of participants. The positive domain B scoring was the most common (11%) and the domain D was least frequently scored positive.2%). Solely four patients (1.3%) were screened positive for all four diagnostic criteria of PTSD (**Table 3**, **Supplementary Figure S1BC**). Clinically relevant symptoms of anxiety (2.3%), depression (5.5%) and somatization (4.9%) were rare in the investigated cohort consisting of 68% highly resilient individuals (**Table 3**). We could not observe any significant association of values of psychometric scores or frequency of mental health problems with the time between trauma center admission and survey completion (not shown).

For clustering and modeling, the cohort was split into the training (n = 230) and test subset (n = 77). The sole significant differences between these subsets concerned resilience class distribution and scores of quality of life, their effect size was weak (**Supplementary Table S6**).

## Three clusters of mental health response after mountain sport accidents

To explore symptoms of PTSD (assessed by positive scoring in at least one PCL-5 domain) and further facets of mental health following accidents in mountain sports , we subjected the participants to medoid clustering in respect to a broad range of psychometric scores (**Supplementary Table S2**). Three mental health clusters were identified in the training subset of the study cohort: ‘neutral’, ‘PTG’ (post-traumatic growth) and ‘PTS’ (post-traumatic stress) and named after their key mental characteristic (**Supplementary Figure S3**). Subsequently, the mental health cluster assignment could be validated in the test subset as evident from comparable fractions of explained clustering variance (training: 0.55, test: 0.52), comparable average silhouette statistics (training: 0.3, test: 0.26), comparable cluster sizes, good visual cluster separation and high similarity between the corresponding clusters in the training and test cohort subsets (**Supplementary Figures S4 - S6**).

The mental health clusters were of approximately equal size but significantly differed in psychometric characteristics. The neutral cluster was characterized by few positive scorings in individuals PCL-5 domains, and low levels of anxiety, depression, , somatization, and post-traumatic growth along with high ratings of sense of coherence, resilience and quality of life. The PTG cluster demonstrated similarly low scores of major mental health disorders, high resilience and sense of coherence. Its key characteristic were the highest levels of post-traumatic growth. The remaining PTS cluster displayed the highest scores of PTSD symptoms (operationalized as positive scoring on at least one PTSD domain), anxiety, depression, , somatization as well as poor sense of coherence, and low resilience and low quality of life. Post-traumatic growth scores in the PTS cluster were higher than in neutral but lower than in PTG cluster participants (**Figure 2**, **Supplementary Table S7**). Consequently, clinically relevant symptoms of anxiety (PTS: 6.4%), depression (PTS: 14%) and somatization (PTS: 10%) were present virtually only found in the PTS cluster. Furthermore, frequencies of low (PTS: 42%) and moderate resilience classes (PTS: 28%) peaked in the PTS cluster. Finally, frequency of flashbacks, PTSD symptoms specified by the B, C, D and E PCL-5 domains and frequency of individuals scoring positive for at least one diagnostic domain of PCL-5 (PTS: 35%) were the highest in PTS cluster individuals (**Figure 3**, **Supplementary Table S8**).

## Demographic, socioeconomic and clinical characteristic of the mental health clusters

Among 48 investigated demographic, socioeconomic, clinical, accident- and recovery-related variables, only 6 features (age, pre-existing physical illness, pre-existing mental disorder, need for psychological or psychiatric support after the accident, persistent physical health consequences of the accident, and caution during mountain sport following the accident) were found to differ significantly between the mental health clusters. The effect size of these differences was weak (**Supplementary Table S9**).

In more detail, individuals in the PTS cluster were the youngest of all participants and had the highest frequency of pre-existing mental disorders. Participants suffering from pre-existing physical illnesses were significantly enriched in the PTG and PTS clusters Differences in gender, income and education, in frequency of search and rescue professionals and prior traumatic events were not significantbetween the clusters (**Figure 4**, **Supplementary Figure S7**). We could not observe any statistically significant differences in being responsible for the accident, rescue mode, accident season, mountain sport type, and numbers of injured persons between the clusters (**Figure 5A**, **Supplementary Figure S7**, **Supplementary Table S9**). There were no significant differences in severity and injury and its location between the clusters. (**Figure 5**, **Supplementary Table S9**).

There were no significant differences in rates of participants who received psychological or psychiatric support after the accident between the clusters. Among PTG and PTS cluster individuals, 8.5% and 14%, respectively, reported need for psychological or psychiatric support following the accident which is significantly higher as compared with none in the neutral cluster. Frequency of participants suffering from persistent physical health consequences of the accident was the highest in the PTS cluster (PTS: 52%, PTG: 32%, neutral: 27%). Analogically, PTS cluster patients reported the highest rates of cautious behavior during mountain sport following the accident (PTS: 78%, PTG: 67%, neutral: 49%). By contrast, percentages of participants who returned to the same mountain sport after the accident were comparable between the mental health clusters (**Figure 6**).

## Prediction of the mental health cluster assignment by demographic, socioeconomic and accident-related factors

Finally, we intended to model the mental health cluster assignment with demographic, socioeconomic, clinical and accident-related factors available during acute medical management of the patient (**Supplementary Table S10**) with seven popular machine learning algorithms: the random forests, neural networks, support vector machines, recursive partitioning, discriminant analysis, elastic net regression, and conditional random forests (**Supplementary Table S11**). Such models would enable early identification of mountain sport accident victims at risk of mental health problems.

The cluster assignment models employing early predictors demonstrated a moderate-to-excellent prediction performance in the training subset of the study cohort (training accuracy: 64% - 100%, Cohen’s : 0.45 - 1, Brier score: 0.018 - 0.57). However, their accuracy in cross-validation and in the test subset was poor (test accuracy: 34% - 47%, Cohen’s : 0.018 - 0.21, Brier score: 0.65 - 0.96). Concerning the early identification of the vulnerable PTS cluster patients, the conditional random forest, random forest and support vector machine models were able to differentiate between the PTS cluster and the pooled neutral and PTG clusters with the best sensitivity (test subset sensitivity 48%, **Figure 7**, **Supplementary Table S12**). Pre-existing mental disorders or physical illness, hospitalization and surgery, injury severity, age, rescue mode, education, annual income, and traumatic events in the past were among the most important explanatory factors employed by these models (**Supplementary Figure S8**). Of note, inclusion of follow-up-related variables such as received psychological or psychiatric support, self-reported need for psychological or psychiatric support after the accident, persisting physical health consequences or caution during mountain sport activity in the machine learning models did not improve their accuracy at prediction of the mental health cluster assignment in cross-validation or the test subset (test accuracy: 36% - 46%, Cohen’s : 0.049 - 0.2, Brier score: 0.64 - 0.92, **Supplementary Figures S9** - **S10**, **Supplementary Table S10**).

# Discussion

We characterized mental health outcomes in individuals treated at a tertiary trauma center for an accident in mountain sport at a median of 44 months following the incident. We identified three approximately equally sized subsets of participants with distinct mental health characteristic: (1) the neutral mental health cluster, (2) the PTG cluster characterized by a predominantly salutatory reaction to the accident manifested by post-traumatic growth, and (3) the PTS cluster hallmarked by symptoms of PTSD, anxiety, depression, panic disorder and somatization as well as low resilience, quality of life and sense of coherence. Individuals in the PTS cluster were characterized by younger age, the highest frequency of pre-existing mental disorders and persisting physical health consequences of the accident compared to individuals in the other two clusters. They reported a need for psychological or psychiatric support following the accident and more cautious behavior during mountain sports since the accident. However, even with robust machine learning algorithms we could not establish a reliable model to identify PTS cluster patients at risk of long-term mental health problems with data available during acute medical management of the mountain sport accident victim.

This study evaluates a broad palette of features of mental health including symptoms of anxiety and depression, somatization, resilience, sense of coherence and quality of life, post-traumatic growth and symptoms of PTSD (operationalized as positive scoring in at least on PTSD domain of the PCL-5) in survivors of mountain sport accidents. This population is of special interest because, on one hand, mountain sports are generally considered beneficial for mental health (2,3). On the other hand, mountain sports are associated with specific individual physical and mental traits, and bear accident and injury risks (5,6). Not many studies have assessed PTSD rates following accidents in mountain sports. Survivors of avalanches showed self-ratings in psychometric assessment suggestive of PTSD11% - 16% (25,26). This figure is comparable with 19% participants of our study screened positive for at least one of the PTSD symptom domains defined by the PCL-5 tool (28). PTSD frequencies of 11.8% - 26% have been reported at 3 - 12 months after acute trauma in setting outside of mountain sports (12,15,17). However, in our cohort, solely 1.3% of participants demonstrated concomitant PCL-5 domain B, C, D and E symptoms suggestive of fully manifest PTSD.

Our study did not specifically address professional mountain rescuers and only 7.2% of the sample reported as search and rescue profession and 5.2% a mountain sport profession. However, studies conducted on mountain guides and mountain rescue personnel, showed that 71% - 78% of them experienced a traumatic event (23,27), while the reported frequency of PTSD was very low (0.9% - 2.7%) (23,27). This is comparable to traumic event estimates for the general population (11,61,62) and, particularly, lower than in other potentially vulnerable groups such as emergency workers (63). By contrast, a recent study performed in France found a prevalence of PTSD in up to 22% of mountain rescuers once differentiating between clinical interviews and self-reporting of symptoms for the assessment of PTSD (24).

Two-thirds of our study participants were classified as highly resilient (32). Resilience has been proposed as a protective trait against negative mental health consequences following trauma (19) and high resilience levels in mountain rescuers may explain low prevalence of PTSD in this population despite trauma exposure (23). The high resilience may be a the results of protective effect of physical activity in an alpine environment on mental health and explain the the low rate of manifest PTSD as well as lower frequencies of clinically relevant symptoms of anxiety and depression in our cohort than estimated for the general Austrian population (64). The high resilience in the study cohort may also pertain to the high rating of quality of life (EUROHIS, median: 4.4, interquartile range: 4 to 4.6) comparable to the general western European population (34).

By semi-supervised clustering, we identified mountain sport accident victims with co-existing symptoms of PTSD (operationalized as scoring positive in at least on PTSD domain), depression, anxiety, somatization, as well as low sense of coherence and quality of life. Those individuals assigned to the PTS subset may represent a clinically relevant risk group of post-traumatic mental disorders. The psychometric characteristic of the PTS cluster corresponds to the multi-faceted mental symptoms described for individuals following a traumatic event (8,9,65). Regarding predictors of mental health problems in mountain accident victims, the PTS cluster was characterized by a significantly younger age as well as high rates of pre-existing physical and mental disorders.. Of note, female gender (17), young age at traumatization (15), and pre-existing mental disorders (14) have been proposed as risk factors of PTSD. PTSD risk has also been previously associated with injury severity, head injuries, hospitalization length and pain (67–69). In our study we could not establish any link between the mental health clusters and, in particular, the PTS cluster with objective injury severity measured with the AIS scale (39) or injury location. Intriguingly, subjective injury severity, which was not surveyed in our cohort, was shown to be a stronger predictor of PTSD symptoms than objective injury severity (70). This phenomenon together with the longer time interval between hospital admission and survey than in most other studies and differences study popultation, may explain the lacking effect of injury location or severety on subsequent mental disorders in our collective. Importantly, the percentage of participants having received psychological or psychiatric support after the accident was generally low in the study cohort (9.1%) and did not differ between the mental health clusters. Hence, it is unlikely that psychological or psychiatric intervention had any impact on classification of the participants.

In our hands, none of seven machine learning procedures using different mathematical principles, random forest, neural network, support vector machines, recursive partitioning, conditional random forest, discriminant analysis or elastic net regression, was able to reliably predict the mental health cluster assignment based on 40 demographic, socioeconomic, clinical and accident-related variables available during acute medical treatment of the patient. This phenomenon is likely attributed to the lacking strong non-psychometric explanatory variables differentiating between the mental health clusters and, of clinical importance, highly specific markers of post-traumatic stress patients. Early identification of trauma patients at risk of developing PTSD or further mental disorders has previously been tackled by diverse machine learning algorithms (13–15,69). Interestingly, inclusion of peri-traumatic mental symptoms like PTSD features, flashbacks, or hyperarousal, were shown to be crucial for optimal prediction of long-term mental disorders or PTSD during recovery (13–15,17,69,70). Unfortunately, such variables were not available in our cohort investigated at least six months after the accident.

# Limitations

Our study bears limitations. The overall analysis inclusion rate was low and, as indicated by comparison of the excluded and analyzed participants, the results were potentially affected by a selection bias towards individuals with higher social status and more severe injuries. Low response rates are unfortunately a problem of many PTSD studies, therefore results have to interpreted with caution (Zitat) It is unclear whether this lead to a bias of under- or overreporting of symptoms since both has been described in the literature. Therefor our results have to be interpreted with caution. Furthermore, the study survey did not record potentially important explanatory factors for distinction between the mental health clusters such as rehabilitation need or ability to work. Due to the cross-sectional design of the study, we were not able to assess peri-traumatic mental health, which has been proven crucial for identification of vulnerable patients. A longitudinal survey may have helped to assess individual trajectories of mental health reaction to the mountain sport accident. Finally, this analysis only included individuals who were treated for a mountain sport accident at the local trauma department. Individuals without contact to the medical system or bystanders, who can also be affected by PTSD, were not examined.

# Conclusions and outlook

Our study characterizes symptoms of PTSD and further facets of mental health following mountain sport accidents. We identified symptoms of mental health impairment in roughly one third of the investigated individuals. This is quite important because generally individuals perforoming mountain sports are generally regarded as “hardened and very healthy”. While the high reslieince and low PTSD rates observed here show that this might be partially true, it is important to consider that also individuals engaging in mountain sports can develop symptoms of PTSD, manifest PTSD or other mental disorders. Destigmatization of such symptoms and disorders is essential in the mountaineering community.

In the current analysis we could not identify robust predictors for impaired mental health at follow up among variables available during early routine medical treatment. Hence, until standardized robust predictor variables are available and screening protocols are in place for mental health outcomes following accidents in mountain sports low threshold access to psychological and psychiatric support as well as psychoeducational measures currently seems the most appropriate strategy for successful interdisciplinary management of individuals following accidents in mountain sport.

# Data and code availability

An R data (RData) file with anonymized patient data will be made available upon request to the corresponding author. The study analysis pipeline is available at <https://github.com/PiotrTymoszuk/mental_accident>.

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# Conflict of interests

The authors declare no conflict of interest.

# Authors’ contributions

Conceptualization: HS, MS, PP, HB, MF, BS-U, KH; Data Curation: HS, PT, MS, NK, KH; Formal Analysis: HS, PT; Funding Acquisition: HS, KH; Investigation: HS, NK, KH; Methodology: PT, KH; Project Administration: BS-U, KH; Resources: HS, NK; Software: PT; Supervision: MS, PP, HB, MF, BS-U, KH; Validation: all authors; Visualization: HS, PT, KH; Writing – Original Draft Preparation: HS, PT, KH; Writing – Review & Editing: all authors

# Ethics approval

The study protocol was approved by the ethics committee of the Medical University of Innsbruck (approval number: 1472/2020). The study was therefore performed in accordance with the ethical standards laid down in the 1964 Declaration of Helsinki and its later amendments. Informed consent was obtained from all participants.

# Tables

Table 1: Demographic and socioeconomic characteristic of the study cohort. Numeric variables are presented as medians with interquartile ranges (IQR). Categorical variables are presented as percentages and counts within the complete observation set.

| **Variablea** | **Statistic** |
| --- | --- |
| Participants, n | 307 |
| Hospital visit – survey time, days | 1300 [IQR: 800 - 1400] range: 390 - 1600 |
| Age at the accident, years | 51 [IQR: 33 - 60] range: 18 - 82 |
| Age at the accident, class, years | 16-30: 20% (n = 61) 31-65: 66% (n = 202) >65: 14% (n = 44) |
| Sex | female: 45% (n = 137) male: 55% (n = 170) |
| Residence in the Alps | 73% (n = 225) |
| Highest education grade | primary: 16% (n = 49) secondary: 38% (n = 115) tertiary: 45% (n = 136) |
| Employment at the accident | employed: 68% (n = 210) unemployed: 3.6% (n = 11) student: 10% (n = 32) retired: 18% (n = 54) |
| Mountain sport profession | 5.2% (n = 16) |
| Search and rescue profession | 7.2% (n = 22) |
| Income/year | no income: 21% (n = 63) < 30000 EUR: 18% (n = 56) 30000 - 45000 EUR: 19% (n = 59) ≥ 45000 EUR: 42% (n = 129) |
| Smoking | 7.8% (n = 24) |
| Problematic alcohol use (CAGE ≥2) | 9.4% (n = 29) |
| Pre-existing physical illness type | none: 85% (n = 260) CVD: 2.9% (n = 9) neurological: 1.3% (n = 4) metabolic: 1.3% (n = 4) pulmonary: 0.65% (n = 2) cancer: 0.65% (n = 2) rheumatoid: 0.33% (n = 1) skin: 0.33% (n = 1) other: 7.8% (n = 24) |
| Number of prior traumatic events/DIA-X | none: 60% (n = 183) 1: 30% (n = 92) 2: 7.5% (n = 23) 3+: 2.9% (n = 9) |
| Pre-existing diagnosed mental disorder | 5.2% (n = 16) |
| Type of pre-existing diagnosed mental disorder | affective disorder: 2.3% (n = 7) personality disorder: 0.33% (n = 1) post-traumatic stress disorder: 0.65% (n = 2) somatoform disorder: 1.6% (n = 5) anxiety disorder: 0.65% (n = 2) attention-deficit hyperactivity disorder: 0.33% (n = 1) addiction: 0.33% (n = 1) bulimia nervosa: 0.33% (n = 1) |
| aCAGE: Cut/Annoyed/Guilty/Eye substance abuse scale; DIA-X: Diagnostic Expert System, traumatic event score. | |

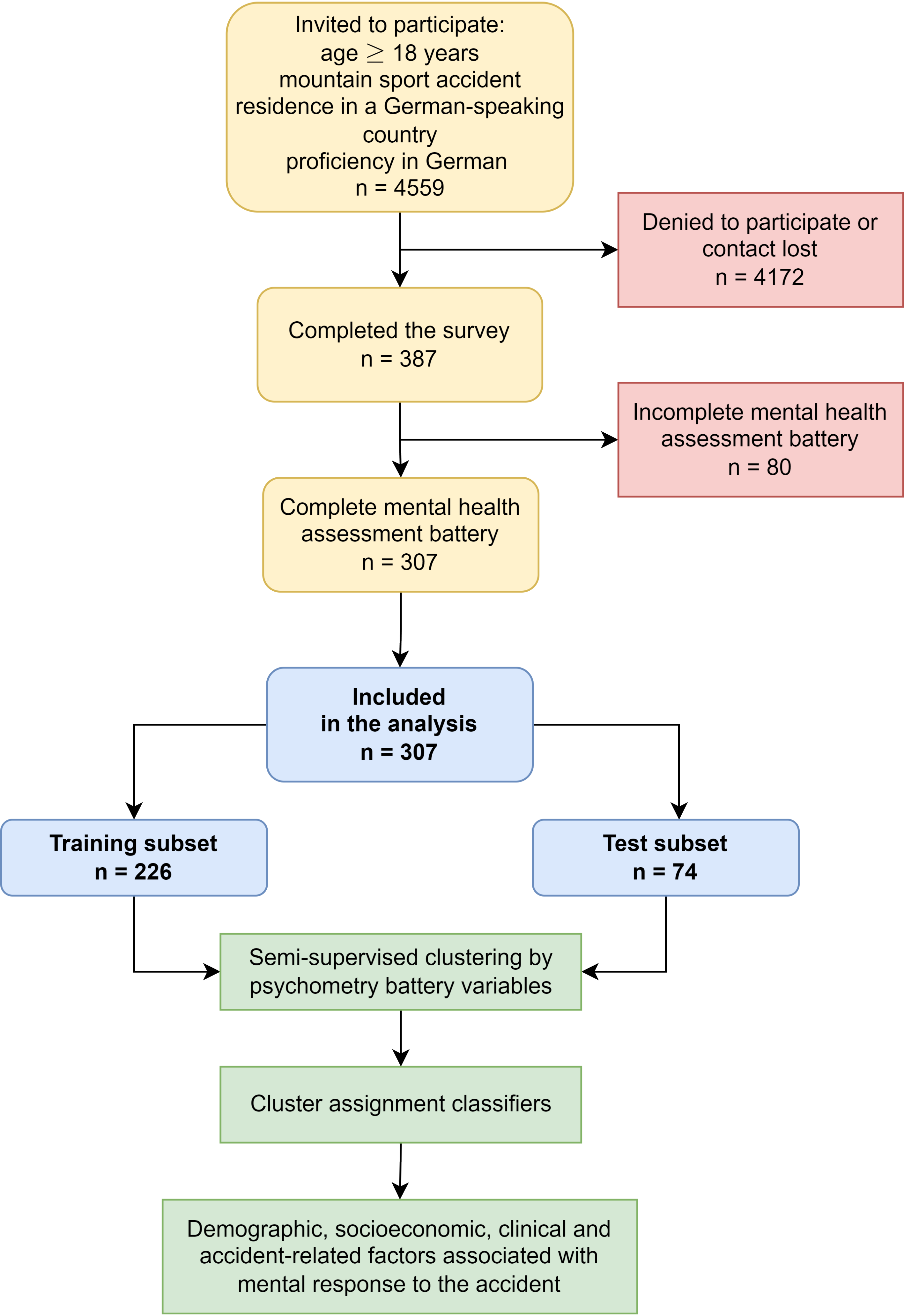
Table 2: Characteristic of the sport accident, injury, psychological management and accident consequences. Numeric variables are presented as medians with interquartile ranges (IQR). Categorical variables are presented as percentages and counts within the complete observation set.

| **Variable** | **Statistic** |
| --- | --- |
| Prior mountain sport accidents | 38% (n = 118) n = 307 |
| Mountain sport typea | ski/snowboard/cross-country: 64% (n = 197) sledding: 3.9% (n = 12) climbing/hiking/mountaineering/skitour: 14% (n = 42) biking: 16% (n = 48) other: 2.6% (n = 8) n = 307 |
| Alone during the accident | 32% (n = 97) n = 307 |
| Responsible for the accident | self: 77% (n = 237) non-self: 23% (n = 70) n = 307 |
| Number of injured persons | only self: 64% (n = 195) self and partner: 3.6% (n = 11) 3+ persons: 1.3% (n = 4) no information: 32% (n = 97) n = 307 |
| Rescue mode | self: 50% (n = 155) companion: 21% (n = 63) rescue team: 29% (n = 89) n = 307 |
| Injury severity class, AIS | 1: 37% (n = 108) 2: 35% (n = 103) 3+: 28% (n = 83) n = 294 |
| Hospitalizedb | 26% (n = 80) n = 307 |
| Surgical therapy | 14% (n = 43) n = 307 |
| Psychological/psychiatric support post accident | 9.1% (n = 28) n = 307 |
| Psychological/psychiatric support need post accident | 7.5% (n = 23) n = 307 |
| Physical health consequences of the accident | 37% (n = 115) n = 307 |
| Returned to same mountain sport post accident | 85% (n = 262) n = 307 |
| Caution during mountain sport post accident | no change: 35% (n = 106) more cautious: 65% (n = 199) less cautious: 0.65% (n = 2) n = 307 |
| Flashback frequency during mountain sport | none: 60% (n = 185) > 1/year: 22% (n = 68) > 1/month: 18% (n = 54) n = 307 |
| aski/snowdoard/cross-country: alpine skiing, snowboarding and cross-country skiing sledding: sledding or bobsled climbing/hiking/mountaineering: hiking, rock and ice climbing, mountaineering, skitouring biking: mountainbike, tour and road cycling. | |
| bAIS: abbreviated injury scale. | |

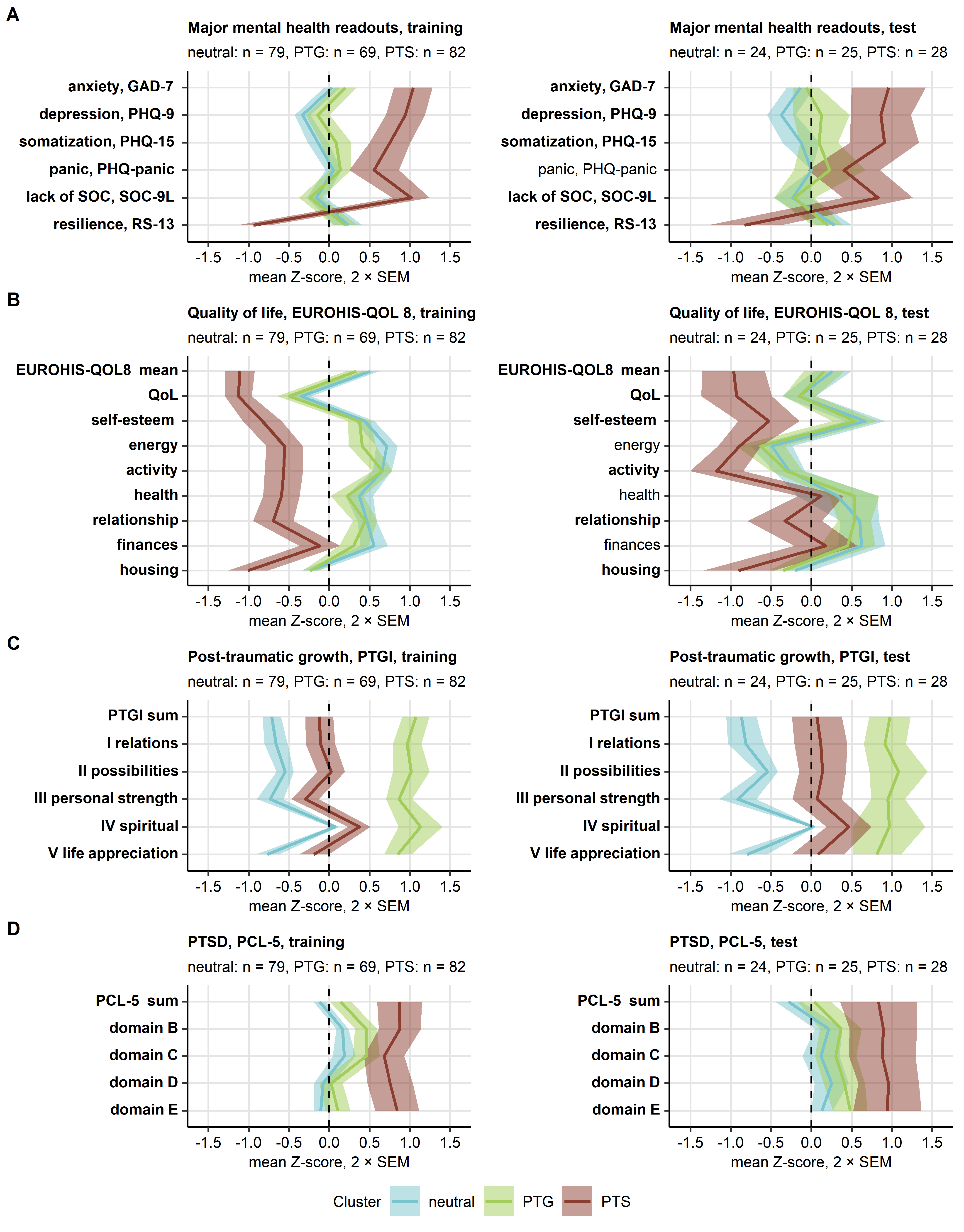
Table 3: Mental health characteristic of the study participants at survey completion. Numeric variables are presented as medians with interquartile ranges (IQR). Categorical variables are presented as percentages and counts within the complete observation set.

| **Variablea** | **Statistic** |
| --- | --- |
| Participants, n | 307 |
| GAD-7 score | 1 [IQR: 0 - 3], range: 0 - 15 |
| Clinically relevant anxiety symptoms (GAD-7 ≥10) | 2.3% (n = 7) |
| PHQ-9 score | 2 [IQR: 1 - 5], range: 0 - 16 |
| Clinically relevant depression symptoms (PHQ-9 ≥11) | 5.5% (n = 17) |
| PHQ-15 score | 2 [IQR: 1 - 4], range: 0 - 23 |
| Clinically relevant somatizaton symptoms (PHQ-15 ≥11) | 4.9% (n = 15) |
| EUROHIS-QOL 8 mean score | 4.4 [IQR: 4 - 4.6], range: 2 - 5 |
| SOC-9L score | 19 [IQR: 16 - 25], range: 10 - 49 |
| RS13 score | 78 [IQR: 70 - 85], range: 15 - 91 |
| RS13 resilience class | low: 18% (n = 56) moderate: 14% (n = 42) high: 68% (n = 209) |
| PTGI score | 32 [IQR: 16 - 48], range: 0 - 100 |
| PCL-5 score | 3 [IQR: 1 - 7], range: 0 - 44 |
| PTSD symptoms (at least one PCL-5 domain positive) | 19% (n = 58) |
| aGAD-7: 7-item general anxiety disorder scale; PHQ-9: 9-item patient health questionnaire for depressive symptoms; PHQ-15: 15-item patient health questionnaire for common somatic symptoms as a substitute for somatization, EUROHIS-QOL 8: 8-item EUROHIS project quality of life scale; SOC-9L: Leipzig 9-item sense of coherence questionnaire; RS13: 13-item resilience scale; PCL-5: PTSD checklist for DSM-5; PTGI: post-traumatic growth inventory; PTSD: post-traumatic stress disorder. | |

# Figures



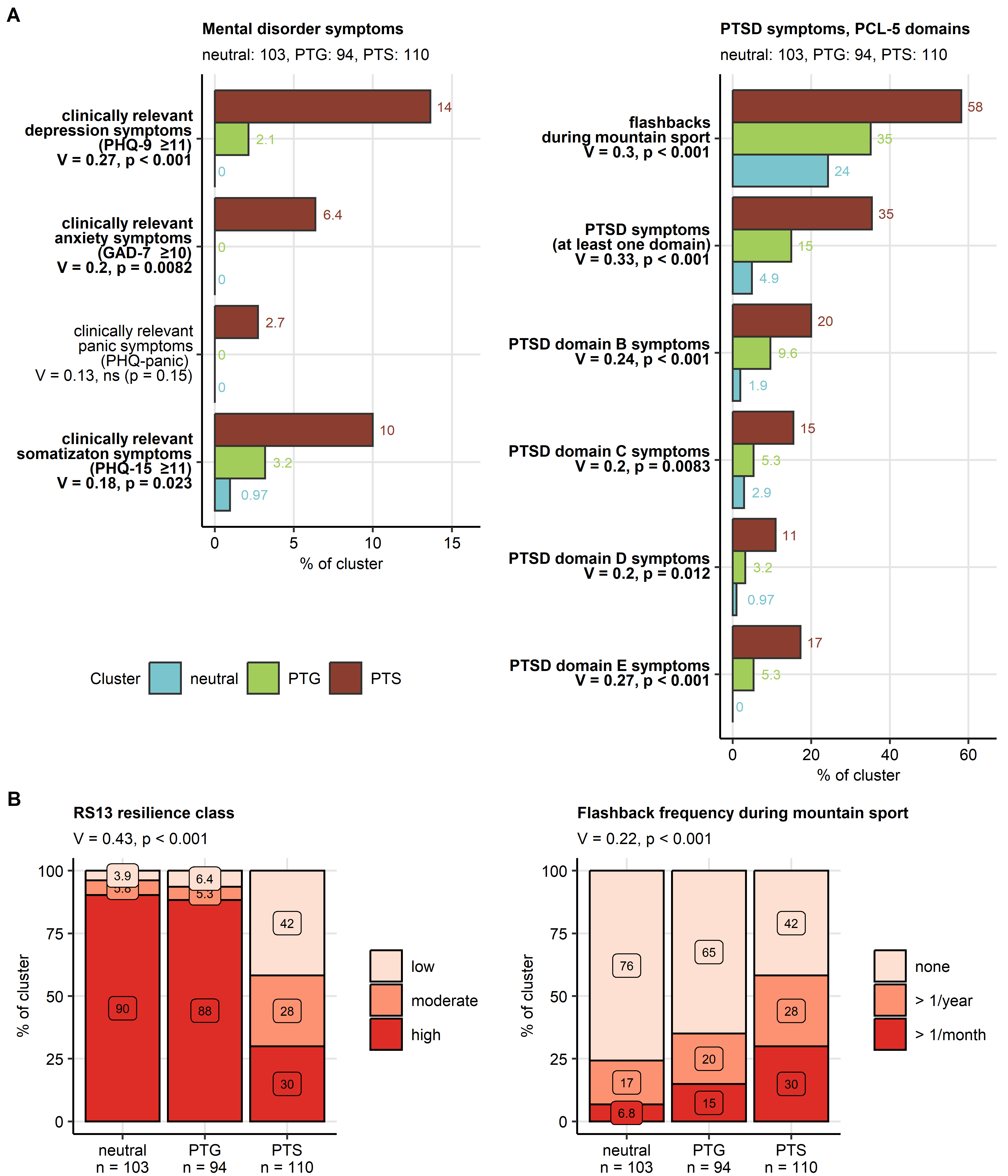
**Figure 1. Flow diagram of the analysis inclusion process.**



**Figure 2. Cluster-defining scores of symptoms of mental disorders, sense of coherence, resilience, quality of life, post-traumatic growth, and post-traumatic stress disorder in the mental health clusters.**

*Study participants in the training subset of the cohort were assigned to the neutral, PTG (post-traumatic growth) and PTS (post-traumatic stress) mental health clusters defined in respect to psychometric scores by the partition around medoids (PAM) algorithm. Study participants in the test subset of the cohort were assigned to the mental health clusters by the inverse distance weighted 27-nearest neighbor classifier. Differences in the cluster-defining psychometric scores between the clusters were assessed by Kruskal-Wallis test with effect size statistic. P values were adjusted for multiple testing with the false discovery rate method. Mean normalized scores of major mental health readouts, sense of coherence and resilience (A), quality of life (B), post-traumatic growth (C) and post-traumatic stress disorder (PTSD, D) in the mental health clusters of the training and test subsets of the study cohort are depicted as solid lines. Tinted regions represent 2 SEM (standard error of the mean). Numbers of observations in the clusters are displayed in the plot captions. Significant effects are highlighted in bold.*

*GAD-7: 7-item general anxiety disorder scale; PHQ-9: 9-item patient health questionnaire for depressive symptoms; PHQ-15: 15-item patient health questionnaire for common somatic symptoms as a substitute for somatization; EUROHIS-QOL 8: 8-item EUROHIS project quality of life scale; SOC-9L: Leipzig 9-item sense of coherence questionnaire; RS13: 13-item resilience scale; PCL-5: PTSD checklist for DSM-5; PTGI: post-traumatic growth inventory; PTSD: post-traumatic stress disorder.*



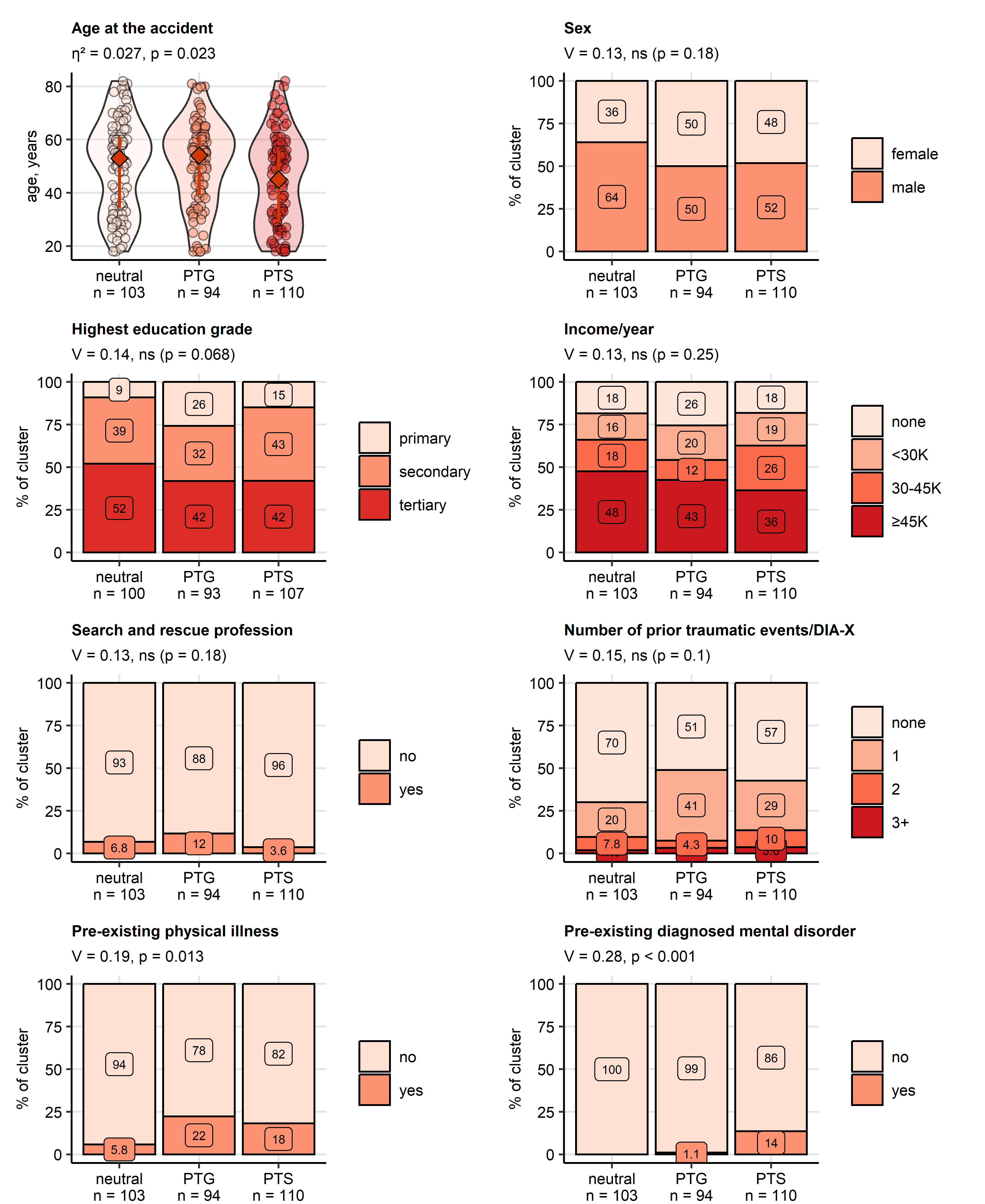
**Figure 3. Symptoms of mental disorders and resilience classes in the mental health clusters.**

*Differences in frequency of symptoms of mental disorder and flashbacks, and distribution of resilience classes between the mental health clusters (neutral, post-traumatic growth [PTG] and post-traumatic stress [PTS]) were investigated in the entire study cohort by test with Cramer’s V effect size statistic. P values were adjusted for multiple testing with the false discovery rate method.*

*(A) Symptoms of mental disorders, presence of flashbacks during mountain sport activity and symptoms of post-traumatic stress disorder (PTSD). Percentages of affected individuals in the mental health clusters are presented in bar plots. Effect sizes and p values are indicated in the Y axes. Significant effects are highlighted in bold. Numbers of observations in the clusters are indicated in the plot captions.*

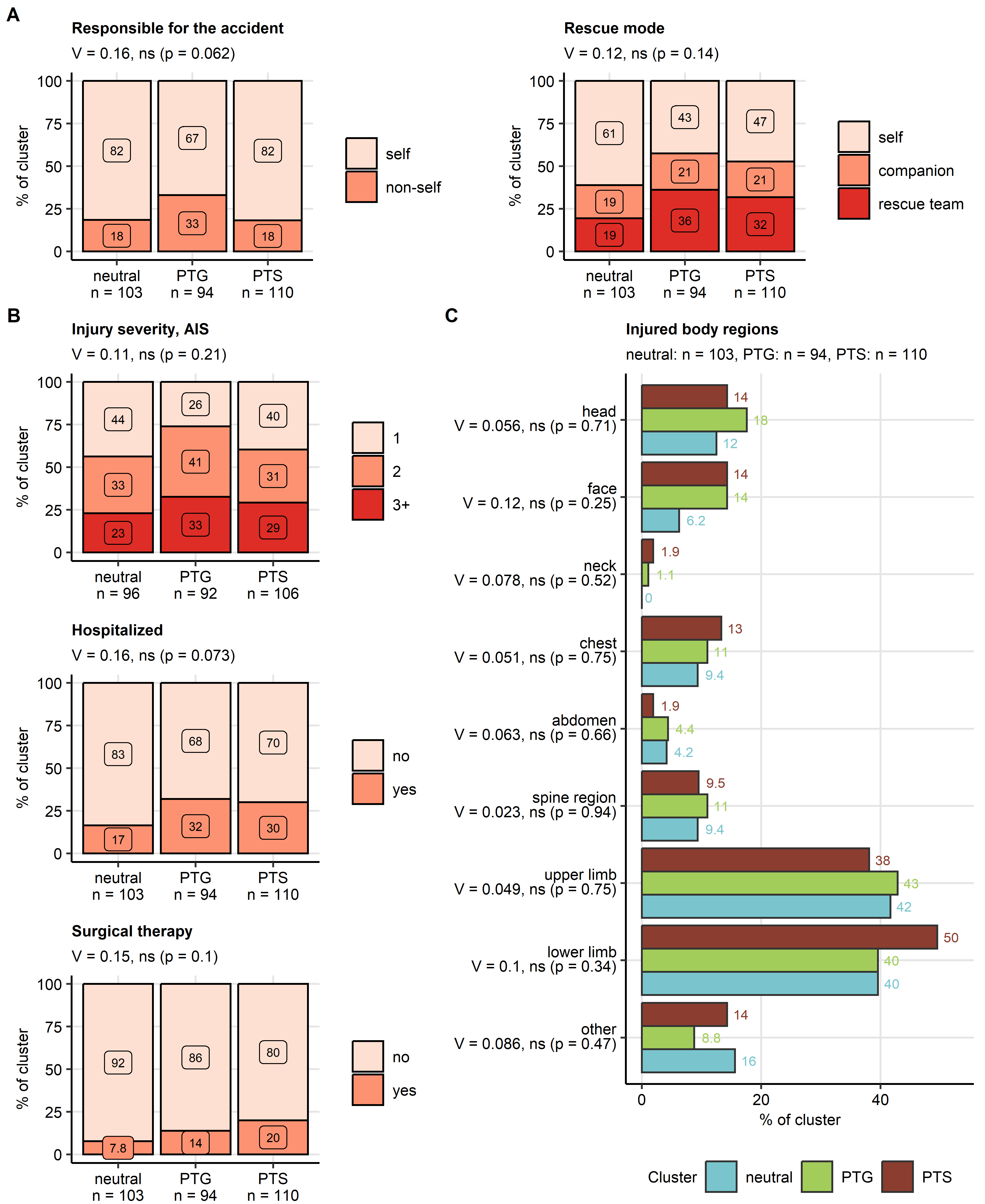
*(B) Resilience classes and frequency of flashbacks during sport activity in the mental health clusters presented in stack plots. Effect sizes and p values are displayed in the plot captions. Numbers of observations in the clusters are indicated in the X axes.*

*GAD-7: 7-item general anxiety disorder scale; PHQ-9: 9-item patient health questionnaire for depressive symptoms; PHQ-15: 15-item patient health questionnaire for common somatic symptoms as a substitute for somatization; RS13: 13-item resilience scale; PCL-5: PTSD checklist for DSM-5; PTSD: post-traumatic stress disorder.*



**Figure 4. Sociodemographic and medical history characteristic of the mental health clusters.**

*Differences in age at the time of the accident between the mental health clusters (neutral, post-traumatic growth [PTG], post-traumatic stress [PTS]) were assessed in the entire cohort by Kruskal-Wallis test with effect size statistic. Differences in genders, education, annual income classes (K: 1000 Euro), percentages of participants with a search and rescue profession, traumatic events in the past (DIA-X: Diagnostic Expert System, traumatic event score), as well as pre-existing physical illness and pre-existing mental disorders diagnosed by a physician between the mental health clusters were investigated in the entire cohort by test with Cramer’s V effect size statistic. P values were adjusted for multiple testing with the false discovery rate method. Age is presented in violin plots with single observations visualized as points and medians with interquartile ranges depicted by red diamonds and whiskers. The remaining variables are presented as percentages of the clusters in stack plots. Effect sizes and p values are displayed in the plot captions. Numbers of observations in the clusters are indicated in the X axes.*

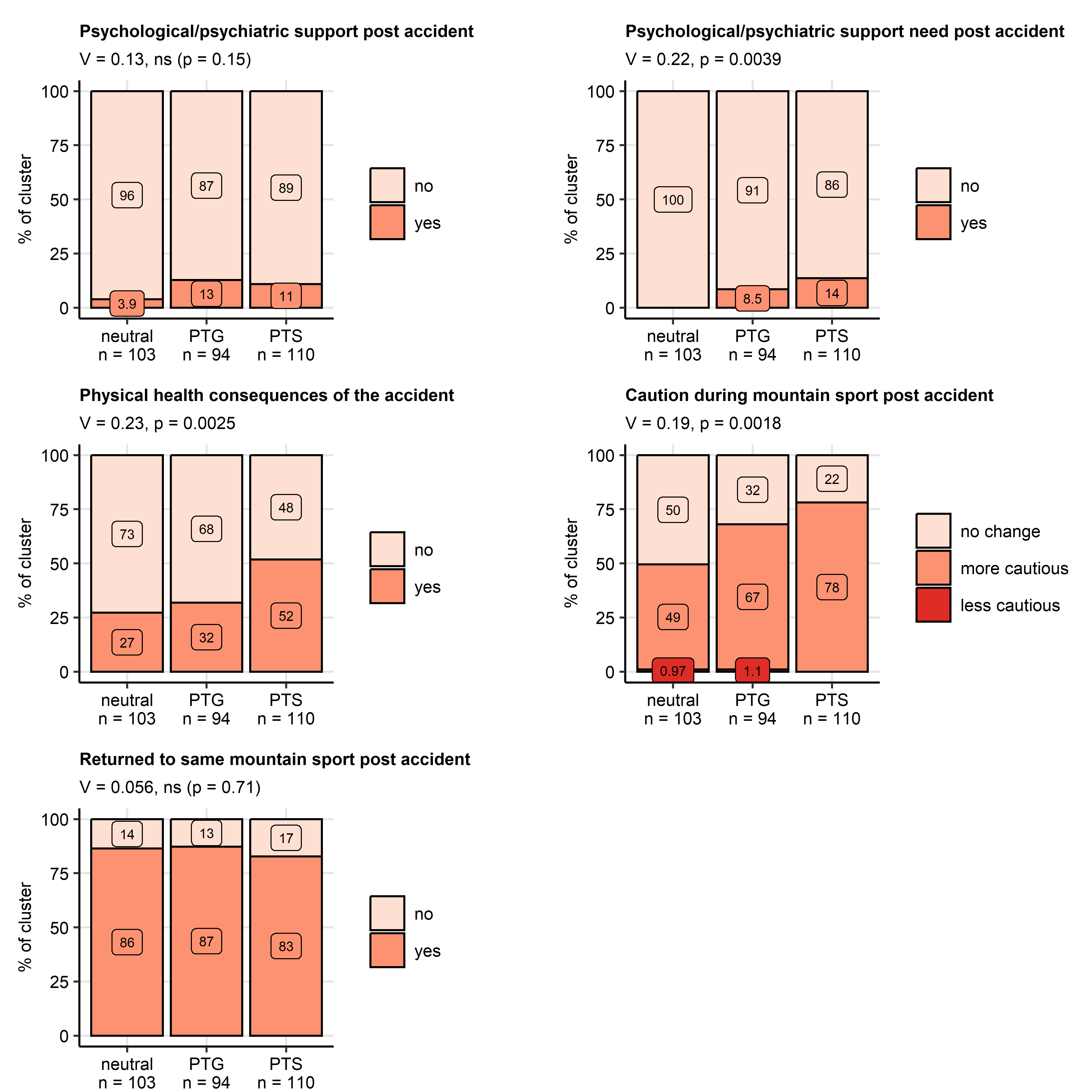


**Figure 5. Responsibility for the accident, accident rescue, injury severity and injured body parts in the mental health clusters..**

*Differences in responsibility for the accident, rescue mode, injury severity, hospitalization and surgery rates, and injured body regions between the mental health clusters (neutral, post-traumatic growth [PTG], post-traumatic stress [PTS]) were investigated in the entire cohort by test with Cramer’s V effect size statistic. P values were adjusted for multiple testing with the false discovery rate method.*

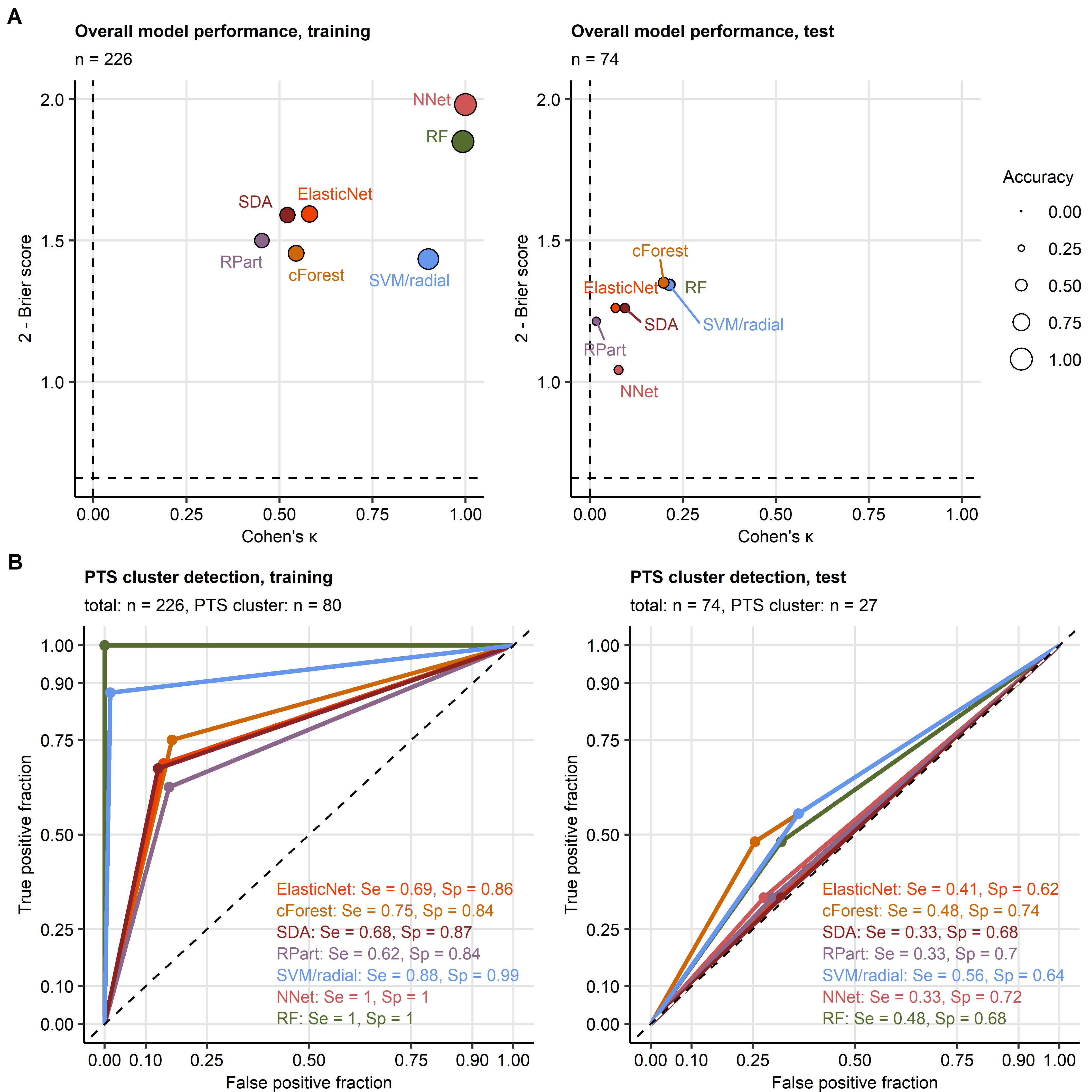
*(A, B) Accident responsibility, distribution of the rescue modes and injury severity classes (AIS: abbreviated injury scale), rates of hospitalization and surgery expressed as percentages of the mental health clusters were presented in stack plots. Effect sizes and p values are displayed in the plot captions. Numbers of observations in the clusters are indicated in the X axes.*

*(C) Percentages of injured body regions were presented in a bar plot. Effect sizes and p values are indicated in the Y axis. Numbers of observations in the clusters are displayed in the plot captions.*



**Figure 6. Psychological support and consequences of the accident in the mental health clusters.**

*Differences in rates of received psychological or psychiatric support, self-reported need for psychological or psychiatric support after the accident, persistent physical health consequences of the accident, cautious behavior during mountain sport and percentages of participants having returned to the same mountain sport between the mental health clusters (neutral, post-traumatic growth [PTG], post-traumatic stress [PTS]) were assessed in the entire cohort by test with Cramer V effect size statistic. P values were adjusted for multiple testing with the false discovery rate method. Percentages within the clusters are presented in stack plots. Effect sizes and p values are displayed in the plot captions. Numbers of observations in the clusters are indicated in the X axes.*



**Figure 7. Assignment of accident victims to the mental health clusters based on explanatory factors available during acute medical management of the patient.**

*The mental health cluster assignment was modeled with demographic, medical history and accident-related explanatory factors available during acute medical management of the patient. Psychometric variables used for cluster definition, mental disorder symptoms, resilience classes, presence and frequency of flashbacks and variables concerning recovery and long term consequences of the accident were excluded from the explanatory factor set.*

*(A) Accuracy of the predicted Cluster assignment and predictive performance of the modeling algorithms was assessed by overall cluster assignment accuracy, Cohen’s inter-rater accuracy metric (high values indicate good accuracy) and Brier score (low values indicate good performance) in the training and test subsets of the study cohort. Performance metrics are presented in scatter plots. Point size codes for the overall cluster assignment accuracy. Point color codes for the modeling algorithm. Numbers of complete observations are displayed in the plot captions.*

*(B) Sensitivity (Se) and specificity (Sp) of detection of participants assigned to the PTS cluster (post-traumatic stress) investigated by receiver-operating characteristic in the training and test subset of the study cohort. Sensitivity and specificity values are indicated in the plots. Line color codes for the modeling algorithm. Numbers of complete observations and observations in the PTS cluster are indicated in the plot captions.*

*RF: random forest; NNet: neural network with a single hidden layer; SVM/radial: support vector machines with radial kernel; RPart: recursive partitioning; SDA: shrinkage discriminant analysis; cForest: conditional random forest; Elastic Net: elastic net multinomial regression.*

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