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Supplementary files: de Ste Croix (2021), item 9259

Supplementary file 1

Supplementary file 1. TRIPOD Checklist: Prediction Model Development and Validation

Section/Topic Item Page			Checklist Item	
Title and abstract				
Title	1	D-V	Identify the study as developing and/or validating a multivariable prediction model, the target population, and the outcome to be predicted.	183
Abstract	2	D-V	Provide a summary of objectives, study design, setting, participants, sample size, predictors, outcome, statistical analysis, results, and conclusions.	183
Introduction				
Background and objectives	3a	D-V	Explain the medical context (including whether diagnostic or prognostic) and rationale for developing or validating the multivariable prediction model, including references to existing models	185
	3b	D-V	Specify the objectives, including whether the study describes the development or validation of the model or both	187
Methods				
Source of data	4a	D-V	Describe the study design or source of data (e.g., randomized trial, cohort, or registry data), separately for the development and validation data sets, if applicable	188
	4b	D-V	Specify the key study dates, including start of accrual; end of accrual; and, if applicable, end of follow-up	189
Participants	5a	D-V	Specify key elements of the study setting (e.g., primary care, secondary care, general population) including number and location of centres.	187
	5b	D-V	Describe eligibility criteria for participants.	188
	5c	D-V	Give details of treatments received, if relevant.	-
Outcome	6a	D-V	Clearly define the outcome that is predicted by the prediction model, including how and when assessed.	191
	6b	D-V	Report any actions to blind assessment of the outcome to be predicted.	
Predictors	7a	D-V	Clearly define all predictors used in developing or validating the multivariable prediction model, including how and when they were measured.	189
	7b	D-V	Report any actions to blind assessment of predictors for the outcome and other predictors.	-
Sample size	8	D-V	Explain how the study size was arrived at.	-
Missing data	9	D-V	Describe how missing data were handled (e.g., complete-case analysis, single imputation, multiple imputation) with details of any imputation method.	192
Statistical analysis	10a	D	Describe how predictors were handled in the analyses.	192
methods	10b	D	Specify type of model, all model-building procedures (including any predictor selection), and method for internal validation.	193
	10c	V	For validation, describe how the predictions were calculated.	193
	10d	D-V	Specify all measures used to assess model performance and, if relevant, to compare multiple models.	193
	10e	V	Describe any model updating (e.g., recalibration) arising from the validation, if done.	194
Risk groups	11	DV	Provide details on how risk groups were created, if done.	193

Development vs. validation	12	V	For validation, identify any differences from the development data in	193
			setting, eligibility criteria, outcome, and predictors.	
Results Participants	13a	D-V	Describe the flow of participants through the study, including the number	
Tarrespants	134	DV	of participants with and without the outcome and, if applicable, a summary of the follow-up time. A diagram may be helpful.	
	13b	D-V	Describe the characteristics of the participants (basic demographics, clinical features, available predictors), including the number of participants with missing data for predictors and outcome.	
	13c	V	For validation, show a comparison with the development data of the distribution of important variables (demographics, predictors and outcome).	
Model development	14a	D	Specify the number of participants and outcome events in each analysis.	187
	14b	D	If done, report the unadjusted association between each candidate predictor and outcome.	189
Model specification	15a	D	Present the full prediction model to allow predictions for individuals (i.e., all regression coefficients, and model intercept or baseline survival at a given time point).	
	15b	D	Explain how to the use the prediction model.	202
Model performance	16	D-V	Report performance measures (with CIs) for the prediction model.	appx.
Model-updating	17	V	If done, report the results from any model updating (i.e., model specification, model performance).	8-10- SInf
Discussion				
Limitations	18	D-V	Discuss any limitations of the study (such as non-representative sample, few events per predictor, missing data).	210
Interpretation	19a	V	For validation, discuss the results with reference to performance in the development data, and any other validation data.	207
	19b	D-V	Give an overall interpretation of the results, considering objectives, limitations, results from similar studies, and other relevant evidence.	207
Implications	20	D-V	Discuss the potential clinical use of the model and implications for future research.	207
Other information				
Supplementary information	21	D-V	Provide information about the availability of supplementary resources, such as study protocol, Web calculator, and data sets.	appx.
Funding	22	D-V	Give the source of funding and the role of the funders for the present study.	-
1.7				

^{*}Items relevant only to the development of a prediction model are denoted by D, items relating solely to a validation of a prediction model are denoted by V, and items relating to both are denoted D-V. We recommend using the TRIPOD Checklist in conjunction with the TRIPOD Explanation and Elaboration document.

Supplementary file 2. Description of the personal or individual injury risk factors recorded

Name	Labels					
Player position	Goalkeeper or outfield player					
Current level of play	1 st division or 2 nd division					
Dominant leg	Right, left or two-footed					
Sex	Male or female					
Age	Sub21, sub23, senior (23-30 y) or veteran (> 30y)					
Pody mass (kg)	<50, 50-54.1, >54.1-58.2, >58.2-62.3, >62.3-66.4, >66.4-					
Body mass (kg)	70.5 or > 70.5					
Statuma (ama)	<148.5, 148.5-156.1, >156.1-163.7, >163.7-171.2, >171.2-					
Stature (cm)	178.8, >178.8-186.4 or >186.4					
History of lower extremity soft tissue injury	Yes or no					
last season	1 es of 110					

Supplementary file 3. Description of the psychological risk factors recorded

Name	Labels								
Sleep quality	<2.46, 2.46-3.02, >3.02-3.58, >3.58-4.14 or >4.14								
Athlete Burnout									
a) Physical/emotional exhaustion	<1.5, 1.5-1.8, >1.8-2.1, >2.1-2.4 or >2.4								
b) Reduced sense of accomplishment	<2.1 or >2.1								
a) Sport develuation	<1.3, 1.3-1.6, >(1.6-1.9, >1.9-2.2, >2.2-2.5, >2.5-2.8, >2.8-3.1,								
c) Sport devaluation	>3.1-3.4, >3.4-3.7 or >3.7								
Psychological C	haracteristics Related to Sport Performance								
a) Stress control	<30.8, 30.8-42.6, >42.6-54.4, >54.4-66.2 or >66.2								
h) Inflyence of enert evaluation	<20.8, >20.8-23.6, >23.6-26.4, >26.4-29.2, >29.2-32, >32-34.8 or								
b) Influence of sport evaluation	>34.8								
c) Mental skills	<13, 13-15, >15-17, >17-19, >19-21, >21-23 or >23								
J) Matirostian	<13.1, 13.1-15.2, >15.2-17.3, >17.3-19.4, >19.4-21.5, >21.5-23.6,								
d) Motivation	>23.6-25.7 or >25.7								
e) Team cohesion	<17, 17-23 or >23								

Supplementary file 4. Description of the measures obtained from the isometric hip abduction and adduction strength test

Name	Labels					
Name	Dominant Leg	Non-Dominant Leg				
	<1.64, 1.64-1.89, >1.89-2.14,	<1.85, 1.85-2.17, >2.17-2.5, >2.5-				
PT _{ISOM} -HipAbd-Normalized	>2.14-2.39, >2.39-2.63, >2.63-	2.83, >2.83-3.16, >3.16-3.48 or				
	2.88 or > 2.88	>3.48				
	<1.57, 1.57-1.84, >1.84-2.11,	-1.50 1.50 1.06 \ 1.06 2.14				
PT _{ISOM} -HipAdd- Normalized	>2.11-2.37, >2.37-2.63, >2.63-2.9	<1.58, 1.58-1.86, >1.86-2.14, >2.14-2.42 or >2.42				
	or >2.9	~2.14-2.42 OI ~2.42				
	<0.74, 0.74-0.82, >0.82-0.91,					
UnRatio-ISOM-HipAbd/HipAdd	>0.91-0.99, >0.99-1.08, >1.08-	<0.69, 0.69-0.83, >0.83-0.97,				
Ulikatio-ISOM-HipAdd/HipAdd	1.17, >1.17-1.25, >1.25-1.34,	>0.97-1.11, >1.11-1.24 or >1.24				
	>1.34-1.42 or >1.42					
BilaRatio-PT _{ISOM} -HipAbd	No Asymmetry or Asymmetry					
BilaRatio-PT _{ISOM} -HipAdd	No Asymmetry or Asymmetry					

Bila: bilateral; Uni: unilateral; ISOM: isometric; PT: peak torque; Abd: abduction; Add: adduction.

Supplementary file 5. Description of the measures obtained from the Y-Balance test

Name	Labels						
Name	Dominant L	eg	No Dominant Leg				
Y-Balance-Anterior	<50.9, 50.9-55.7, >	55.7-60.5,	<51.3, 51.3-56.7, >56.7-62.2, >62.2-				
1-Barance-America	>60.55-65.4, >65.4-70	0.2 or > 70.2	67.7, >67.7-73.1 or >73.1				
	<83.1, 83.1-88.7, >8	88.7-94.4,	<93.3, 93.3-97.6, >97.6-101.8,				
Y-Balance-PosteroMedial	>94.4-100.1, >100.1-10	05.8, >105.8-	>101.8-106.1, >106.1-110.4 or >110.4				
	111.4 or >11	1.4					
Y-Balance-PosteroLateral	<81.7, 81.7-91.4, >9	1.4-101.1,	<89.2, 89.2-97.0, >97.0-104.9,				
r-Barance-PosteroLaterar	>101.1-110.7 or	>110.7	>104.9-112.7 or >112.7				
BilaRatio-Y-Balance-	No Asymmetry or Asymmetry						
Anterior		No Asymmen	y of Asymmetry				
BilaRatio-Y-Balance-		No Asymmetr	ay or Acymmetry				
PosteroMedial		No Asymmen	ry or Asymmetry				
BilaRatio-Y-Balance-		No Agymmet	ay on A aymmatay				
PosteroLateral		No Asymmen	ry or Asymmetry				
Y-Balance-Composite	<78.4, 78.4-85.9,	<80.4, 80.4	-84.1, >84.1-87.8, >87.8-91.5 or >91.5				
1 -Datance-Composite	>85.9-93.3 or >93.3						

Supplementary file 6. Description of the measures obtained from the lower extremity range of motion assessment tests

Name	Labels						
Name	Dominant Leg	Non-Dominant Leg					
DOM HE	<117.5, 117.5-125, >125-132.5, >132.5-	<118.2, 118.2-126.3, >126.3-134.5,					
ROM - HF_{KF}	140, >140-147.5 or >147.5	>134.5-142.7, >142.7-150.8 or >150.8					
ROM-HF _{KE}	<63.7, 63.7-71.4, >71.4-79.1 or >79.1	<59, 59-68 or >68					
ROM-HE	<0.1, 0.1-3.8, >3.8-7.7, >7.7-11.6, >11.6-	<0.1, 0.1-4.2, >4.2-8.3, >8.3-12.4, >12.4-					
ком-пе	15.5 or >15.5	16.5, >16.5-20.6 or >20.6					
	<42.9, 42.9-48.8, >48.8-54.7, >54.7-60.6,						
ROM-HABD	>60.6-66.5, >66.5-72.4, >72.4-78.3 or	<46.5, 46.5-67, >67-87.5 or >87.5					
	>78.3						
ROM-HIR	<35, 35-50, >50-65 or >65	<30.9, 30.9-36.8, >36.8-42.7 or >42.7					
ROM-HER	<40.8, 40.8-50.6, >50.6-60.4, >60.4-70.2	<42.8, 42.8-54.6, >54.6-66.4, >66.4-78.2					
KOWI-IIEK	or >70.2	or >78.2					
	<106.4, 106.4-112.8, >112.8-119.2,	<98.4, 98.4-105.7, >105.7-113.1, >113.1-					
ROM-KF	>119.2-125.6, >125.6-132, >132-138.4,	120.5, >120.5-127.9, >127.9-135.2,					
	>138.4-144.8 or >144.8	>135.2-142.6 or >142.6					
ROM-AKDF _{KE}	<44.5 or >44.5	<24.4, 24.4-29.8, >29.8-35.2, >35.2-40.6					
KOWI-AKDI KE	\\\-44.5 01 \\/\-44.5	or >40.6					
	<24.9, 24.9-27.8, >27.8-30.7, >30.7-33.6,	<24, 24-27, >27-30, >30-33, >33-36, >36-					
ROM- $AKDF_{KF}$	>33.6-36.5, >36.5-39.4, >39.4-42.3 or	39 or >39					
	>42.3	57 01 7 37					
ROM-BIL-	No Asymmetry	y or Asymmetry					
HF_{KF}	110 715711111041	y or rusymmoury					
ROM-BIL-	No Asymmetry	y or Asymmetry					
HF _{KE}							
ROM-BIL- HE	No Asymmetry	y or Asymmetry					
ROM-BIL-	No Asymmetry	y or Asymmetry					
HABD							
ROM-BIL- HIR	No Asymmetry	y or Asymmetry					
ROM-BIL-	No Asymmetry	y or Asymmetry					
HER							
ROM-BIL- KF	No Asymmetry	y or Asymmetry					
ROM-BIL-	No Asymmetry	y or Asymmetry					
AKDF _{KE}							
ROM-BIL-	No Asymmetry	y or Asymmetry					
$AKDF_{KF}$	1.0 Hayimileti	,, ,					

ROM: range of motion; HF_{KF}: hip flexion with the knee flexed; HF_{KE}: hip flexion with the knee extended; HE: Hip extension; HABD: hip abduction at 90° of hip flexion; HIR: hip internal rotation; HER: hip external rotation; KF: knee flexion; AKDF_{KE}: ankle dorsi-flexion with the knee extended; AKDF_{KF}: ankle dorsi-flexion with the knee flexed; BIL: bilateral ratio.

Supplementary file 7. Descriptions of the resampling, ensemble and cost-sensitive algorithms applied to the base classifiers.

With regard to the resampling techniques, four (two oversampling and two undersampling algorithms) of the most popular methodologies were selected, which are the synthetic minority oversampling technique (SMOTE)²⁵⁷, random oversampling (ROS), random undersampling (RUS) and Wilson's edited nearest neighbor rule (ENN)²⁵⁸. In the four resampling techniques selected, a level of balance in the training data near the 40/60 was attempted. In addition, the interpolations that are computed to generate new synthetic data are made considering the k-5-nearest neighbors of minority class instances using the Euclidean distance.

Regarding ensemble learning algorithms, classic ensembles such as Bagging²⁵⁹, AdaBoost²⁶⁰ and AdaBoot.M1²⁶¹ were included in this study. Furthermore, the algorithm families designed to deal with skewed class distributions in data sets were also included: Boosting-based and Bagging-based. The Boosting based ensembles that were considered in the current study were SMOTEBoost²⁶² and RUSBoost²⁶³. Concerning Bagging based ensembles, it was included from the OverBagging group, OverBagging (which uses ROS)²⁶⁴, UnderBagging (which uses RUS)²⁶⁴ and SMOTEBagging²⁶⁴. The number of internal classifiers used within each ensemble learning algorithm was set 100 (always the same) base classifiers (C4.5, ADTree, SVM and KNN) by default.

Concerning the cost-sensitive learning algorithms, two different algorithms were used, namely MetaCost²⁶⁵ and cost-sensitive classifier. Cost-sensitive learning solutions incorporating both the data (external) and algorithmic level (internal) approaches assume higher misclassification costs for samples in the minority class and seek to minimize the high cost errors. For the both cost-sensitive algorithms selected, the cox matrix set-up was to:

$$c = \begin{cases} 0 & 2 \\ 1 & 0 \end{cases}$$
 where a false negative has a cost of 2 and false positive had a cost of 1.

The behavior of some specific combinations of class-balanced ensembles with cost-sensitive base classifiers was also studied. Finally, the algorithm Random Forest²⁶⁶ in isolation and in combination with the resampling techniques was also explored due to its good results showed in previous studies²⁶⁷.

For the sake of brevity and the lack of space, the code of the algorithms used in this study has not been written here. Instead, we have only specified the names and refer the reader to their original sources. Furthermore, all the classification algorithms used are available in Weka Data Mining software.

Supplementary file 8. AUC results (mean and standard deviation) of the personal or individual characteristics data set (DS 1) for the five base classifiers in isolation and after applying in them the resampling, ensemble (Classic, Boosting-based, Bagging-based and Class-balanced ensembles) and cost-sensitive learning techniques selected

	Base classifiers							
Technique	C4.5	ADTree	SMO	KNN	RF			
	AUC	AUC	AUC	AUC	AUC			
None	0.47 ± 0.09	0.57 ± 0.11	0.49 ± 0.01	0.50 ± 0.12	$0.492 \begin{array}{c} \pm 0.11 \\ 1 \end{array}$			
None	5 1	9 7	4 6	4 6	1			
SMOTE	0.47 ± 0.13	0.56 ± 0.12	0.48 ± 0.09	0.48 ± 0.12	0.522 ± 0.11			
SMOTE	4 4	1 3	8 6	7 0	3			
ROS	0.45 ± 0.11	0.57 ± 0.12	0.49 ± 0.10	0.48 ± 0.12	0.497 ± 0.11			
ROS	4 7	0 9	6 0	8 1	4			
RUS	0.49 ± 0.10	0.56 ± 0.13	0.50 ± 0.10	0.51 ± 0.12	0.490 ± 0.12			
ROS	5 3	5 1	5 9	7 9	4			
ENN	0.50 ± 0.00	0.56 ± 0.12	0.49 ± 0.02	0.50 ± 0.13	0.496 ± 0.12			
	0 6	3 8	1 7	5 7	1			
		C	lassic Ensemb	les				
ADB1	0.43 ± 0.11	0.47 ± 0.11	0.50 ± 0.10	0.47 ± 0.13				
ADBI	5 7	2 5	1 1	6 5				
M1	0.45 ± 0.11	0.47 ± 0.12	0.51 ± 0.12	0.46 ± 0.10				
1711	4 3	5 0	1 4	9 9				
BAG	0.49 ± 0.11	0.57 ± 0.10	0.51 ± 0.11	0.50 ± 0.12				
DitG	6 7	9 9	2 8	2 0				
Decorate	0.42 ± 0.12	0.50 ± 0.12		0.43 ± 0.11				
	2 4	1 0	4 6	3 3				
-			ing-based Ens					
SBO	0.48 ± 0.12	0.51 ± 0.12	0.50 ± 0.12	0.48 ± 0.11				
БВС	3 1	3 2	9 9	2 8				
RUSB	0.46 ± 0.12	0.48 ± 0.11	0.48 ± 0.12	0.45 ± 0.11				
	4 4	6 4	5 8	8 9				
		Baggi	ng-based Ense	embles				
OBAG	0.49 ± 0.11	0.57 ± 0.10	0.55 ± 0.11					
OBAG	2 2	3 7	4 6	3 1				
UBAG	0.52 ± 0.11	0.57 ± 0.10	0.56 ± 0.11	0.52 ± 0.11				
CB/1G	8 9	9 6	8 4	8 9				
SBAG	0.53 ± 0.11	0.58 ± 0.10	0.55 ± 0.11	0.52 ± 0.11				
	3 2	3 5	1 6	4 2				
		Cost-se	ensitive Classi	fication				
MetaCost	0.49 ± 0.01		0.48 ± 0.03	0.50 ± 0.13				
Miciacost	9 3	0 7	5 6	8 5				
CS-	0.48 ± 0.06	0.57 ± 0.12	0.47 ± 0.06	0.50 ± 0.12				
Classifier	0 0	4 2	4 1	5 5				

Class-balanced Ensembles with a Cost-sensitive Classifier									
CS-OBAG	0.52 ± 0.11	0.57 ± 0.10	0.56 ± 0.11	0.48 ± 0.11	-	-			
CS-OBAG	1 1			5 3					
CS-UBAG	0.53 ± 0.11	0.58 ± 0.10	0.57 ± 0.11	0.52 ± 0.12	-	-			
CS-UBAG	8 2	1 8		8 1					
CS-SBAG	0.54 ± 0.10	0.58 ± 0.10	0.55 ± 0.11	0.52 ± 0.11	-	-			
CS-SBAG	5 9	4 4	1 6	3 3					

Supplementary file 9. AUC results (mean and standard deviation) of the sleep quality data set (DS 2) for the four base classifiers in isolation and after applying in them the resampling. ensemble (Classic, Boosting-based, Bagging-based and Class-balanced ensembles) and cost-sensitive learning techniques selected

	Base classifiers									
Technique	C	4.5	ADTree		SMO		KNN		RF	
	A	UC	A	UC	Αl	JC	AUC		AUC	
None	0.500	±0.000	0.458	±0.123	0.500	±0.000	0.461	±0.124	0.454	±0.122
				Resa	mpling	Techniq	ues			
SMOTE	0.410	±0.127	0.409	±0.131	0.451	±0.092	0.409	±0.130	0.407	±0.131
ROS	0.475	±0.068	0.452	±0.131	0.492	±0.065	0.455	±0.128	0.444	±0.133
RUS	0.491	±0.044	0.459	±0.132	0.490	±0.074	0.460	±0.134	0.458	±0.134
ENN	0.500	±0.000	0.466	±0.132	0.498	±0.011	0.467	±0.134	0.463	±0.133
				С	lassic Er	nsembles	6			
ADB1	0.452	±0.111	0.458	±0.123	0.473	±0.088	0.458	±0.122	-	-
M1	0.454	±0.093	0.459	±0.122	0.459	±0.120	0.458	±0.122	-	-
BAG	0.485	±0.062	0.425	±0.117	0.523	±0.091	0.455	±0.122	-	-
Decorate	0.497	±0.032	0.433	±0.126	0.500	±0.000	0.451	±0.124	-	-
				Boost	ing-base	ed Ensen	nbles			
SBO	0.421	±0.126	0.421	±0.126	0.444	±0.106	0.422	±0.128	-	-
RUSB	0.461	±0.100	0.462	±0.129	0.456	±0.122	0.474	±0.126	-	-
				Baggi	ng-base	d Ensem	ıbles			
OBAG	0.415	±0.119	0.407	±0.120	0.411	±0.118	0.416	±0.120	-	-
UBAG	0.477	±0.129	0.444	±0.120	0.509	±0.121	0.454	±0.122	-	-
SBAG	0.378	±0.119	0.376	±0.117	0.413	±0.117	0.375	±0.118	-	-
				Cost-se	ensitive	Classific	cation			
MetaCost	0.500	±0.000	0.503	±0.106	0.498	±0.012	0.576	±0.122	-	-
CS- Classifier	0.500	±0.000	0.458	±0.122	0.484	±0.030	0.461	±0.124	-	-
		Class	-balanc	ed Ensen	nbles wi	th a Cos	t-sensit	ive Clas	sifier	
CS-OBAG	0.415	±0.118	0.407	±0.120	0.426	±0.118	0.416	±0.118	-	-
CS-UBAG	0.431	±0.125	0.438	±0.121	0.431	±0.121	0.433	±0.121	-	-
CS-SBAG	0.370	±0.117	0.374	±0.118	0.365	±0.115	0.373	±0.118	-	-

Supplementary file 10. AUC results (mean and standard deviation) of the Athlete Burnout data set (DS 3) for the four base classifiers in isolation and after applying in them the resampling. ensemble (Classic, Boosting-based, Bagging-based and Class-balanced ensembles) and cost-sensitive learning techniques selected

				Base cl	assifiers				
Technique	C4.5	ADTr	ee	SN	ИΟ	KI	NN]	RF
_	AUC	AUC	2	A	UC	Al	UC	A	UC
None	0.500 ±0.00	00 0.558 ±	0.127	0.495	±0.024	0.642	±0.117	0.633	±0.121
			Res	ampling	g Techniqu	ıes			
SMOTE	0.543 ±0.12	22 0.537 ±	0.126	0.511	±0.102	0.614	±0.114	0.598	±0.114
ROS	0.542 ± 0.12	.3 0.568 ±	0.121	0.532	±0.102	0.642	±0.118	0.630	±0.120
RUS	0.494 ± 0.04	4 0.558 ±	0.123	0.525	±0.097	0.604	±0.121	0.592	±0.127
ENN	0.500 ± 0.00	00 0.553 ±	0.125	0.502	±0.038	0.619	±0.127	0.618	±0.128
			(Classic E	Insembles				
ADB1	0.577 ±0.12	25 0.617 ±	0.126	0.523	±0.099	0.627	±0.127	-	-
M1	0.564 ±0.12	3 0.615 ±	0.126	0.560	±0.122	0.630	±0.118	-	-
BAG	0.506 ±0.10	06 0.579 ±	0.128	0.530	v0.118	0.636	±0.120	-	-
Decorate	0.521 ±0.12	22 0.588 ±	0.133	0.495	±0.024	0.610	±0.124	-	-
			Boos	sting-bas	ed Ensem	bles			
SBO	0.596 ±0.12	.3 0.594 ±	0.126	0.570	±0.119	0.619	±0.122	-	-
RUSB	0.591 ±0.12	22 0.612 ±	0.126	0.572	±0.122	0.624	±0.121	-	-
			Bagg	ging-bas	ed Ensem	bles		•	
OBAG	0.610 ±0.12	.4 0.583 ±	0.126	0.588	±0.121	0.636	±0.120	-	-
UBAG	0.562 ± 0.13	3 0.577 ±	0.125	0.568	±0.119	0.617	±0.123	-	-
SBAG	0.585 ± 0.12	.4 0.581 ±	0.126	0.570	±0.119	0.622	±0.116	-	-
			Cost-	sensitive	e Classific	ation			
MetaCost	0.500 ±0.00	00 0.555 ±0	0.125	0.512	±0.048	0.562	±0.138	-	-
CS- Classifier	0.500 ±0.00	00 0.562 ±0	0.125	0.523	±0.063	0.643	±0.118	-	-
	C	lass-balance	d Ense	mbles w	rith a Cost	t-sensitiv	e Classif	fier	
CS-OBAG	0.592 ±0.12	.8 0.581 ±	0.128	0.580	±0.122	0.635	±0.119	-	-
CS-UBAG	0.564 ±0.12	22 0.578 ±	0.127	0.568	±0.124	0.616	±0.125	-	-
CS-SBAG	0.583 ±0.11	9 0.579 ±	0.127	0.565	±0.121	0.624	±0.116	-	-

Supplementary file 11. AUC results (mean and standard deviation) of the psychological characteristics related to sport performance data set (DS 4) for the four base classifiers in isolation and after applying in them the resampling. ensemble (Classic, Boosting-based, Bagging-based and Class-balanced ensembles) and cost-sensitive learning techniques selected

	Base classifiers									
Technique	C	4.5	AD	Tree	SI	ON	KNN		1	RF
	A	UC	A	UC	A	UC	A	UC	A	UC
None	0.500	±0.000	0.435	±0.122	0.492	±0.015	0.457	±0.105	0.379	±0.101
				Res	ampling	g Techni	ques			
SMOTE	0.458	±0.126	0.471	±0.135	0.490	±0.102	0.448	±0.116	0.417	±0.126
ROS	0.422	±0.122	0.441	±0.128	0.451	±0.090	0.458	±0.107	0.384	± 0.104
RUS	0.494	±0.050	0.448	±0.132	0.450	±0.102	0.474	±0.126	0.408	±0.120
ENN	0.500	±0.000	0.450	±0.131	0.490	±0.023	0.477	±0.116	0.403	±0.111
				(Classic I	Ensemble	es			
ADB1	0.419	±0.121	0.458	±0.114	0.463	±0.103	0.487	±0.105	-	-
M1	0.427	±0.125	0.446	±0.119	0.440	±0.121	0.414	±0.095	-	-
BAG	0.455	±0.115	0.431	±0.116	0.405	±0.112	0.468	±0.110	-	-
Decorate	0.487	±0.137	0.467	±0.121	0.492	±0.015	0.383	±0.120	-	-
				Boos	ting-bas	sed Ense	mbles			
SBO	0.451	±0.126	0.449	±0.123	0.452	±0.128	0.467	±0.122	-	-
RUSB	0.427	±0.121	0.435	±0.121	0.439	±0.128	0.464	±0.126	-	-
	•			Bagg	ing-bas	ed Ensei	nbles		•	
OBAG	0.417	±0.109	0.434	±0.117	0.440	±0.121	0.456	±0.113	-	-
UBAG	0.429	±0.113	0.430	±0.118	0.412	±0.119	0.474	±0.117	-	-
SBAG	0.436	±0.115	0.457	±0.119	0.459	±0.120	0.445	±0.115	-	-
				Cost-s	sensitiv	e Classifi	ication			
MetaCost	0.500	±0.000	0.417	±0.118	0.480	±0.029	0.465	±0.105	-	-
CS-Classifier	0.500	±0.000	0.433	±0.121	0.463	±0.047	0.457	±0.105	-	-
		Class	s-balano	ced Ense	mbles v	vith a Co	st-sensi	itive Clas	sifier	
CS-OBAG	0.426	±0.109	0.436	±0.118	0.434	±0.121	0.456	±0.113	-	-
CS-UBAG	0.437	±0.115	0.427	±0.117	0.427	±0.120	0.471	±0.115	-	-
CS-SBAG	0.447	±0.118	0.456	±0.120	0.448	±0.120	0.443	±0.116	-	

Supplementary file 12. AUC results (mean and standard deviation) of the self-perceived chronic ankle instability data set (DS 5) for the four base classifiers in isolation and after applying in them the resampling. ensemble (Classic, Boosting-based, Bagging-based and Class-balanced ensembles) and cost-sensitive learning techniques selected

	Base classifiers									
Technique	С	4.5	AD	Tree	SI	ON	K	NN	F	RF
	A	UC	A	AUC AUC AU		AUC		A	UC	
None	0.500	±0.000	0.596	±0.108	0.497	±0.014	0.596	±0.109	0.598	±0.111
				Res	ampling	g Technic	lues			
SMOTE	0.572	±0.108	0.564	±0.107	0.520	±0.085	0.552	±0.108	0.556	±0.108
ROS	0.551	±0.100	0.597	±0.115	0.532	±0.079	0.592	±0.118	0.596	±0.118
RUS	0.517	±0.075	0.582	±0.118	0.530	±0.087	0.582	±0.120	0.588	±0.122
ENN	0.500	±0.000	0.590	±0.116	0.500	±0.019	0.589	±0.120	0.589	±0.120
				(Classic I	insemble	s			
ADB1	0.595	±0.108	0.597	±0.109	0.526	±0.091	0.596	±0.110	-	-
M1	0.599	±0.113	0.595	±0.109	0.605	±0.115	0.595	±0.108	-	-
BAG	0.583	±0.111	0.600	±0.112	0.543	±0.085	0.597	±0.112	-	-
Decorate	0.519	±0.122	0.508	±0.117	0.497	±0.014	0.509	±0.118	-	-
				Boos	ting-bas	ed Enser	nbles			
SBO	0.558	±0.114	0.551	±0.112	0.559	±0.116	0.541	±0.110	-	-
RUSB	0.584	±0.111	0.593	±0.113	0.579	±0.123	0.590	±0.114	-	-
	•	•	•	Bagg	ing-bas	ed Ensen	nbles	•	•	
OBAG	0.588	±0.116	0.604	±0.114	0.604	±0.111	0.597	±0.115	-	-
UBAG	0.612	±0.118	0.599	±0.113	0.595	±0.123	0.594	±0.112	-	-
SBAG	0.567	±0.113	0.576	±0.113	0.606	±0.116	0.566	±0.115	-	-
				Cost-s	sensitive	e Classifi	cation			
MetaCost	0.499	±0.007	0.518	±0.123	0.498	±0.024	0.478	±0.126	-	-
CS-Classifier	0.501	±0.030	0.596	±0.109	0.532	±0.054	0.596	±0.110	-	-
		Class	s-balano	ced Ense	mbles w	rith a Cos	st-sensit	tive Class	sifier	
CS-OBAG	0.589	±0.116	0.604	±0.113	0.604	±0.113	0.597	±0.115	-	-
CS-UBAG	0.608	±0.117	0.601	±0.113	0.599	±0.113	0.594	±0.114	-	-
CS-SBAG	0.555	±0.111	0.574	±0.113	0.602	±0.112	0.556	±0.113	-	-

Supplementary file 13. AUC results (mean and standard deviation) of the lower extremity joint ranges of motion data set (DS 6) for the five base classifiers in isolation and after applying in them the resampling, ensemble and cost-sensitive learning techniques selected

Base classifiers										
Technique	C	4.5	.5 ADTree SMO KNN		NN		RF			
	A	UC	A	UC	A	UC	A	UC	1	AUC
None	0.629	±0.115	0.754	±0.122	0.567	±0.098	0.591	±0.125	0.690	±0.125
				Res	amplin	g Techni	ques			
SMOTE	0.614	±0.121	0.710	±0.126	0.563	±0.101	0.601	±0.117	0.679	±0.117
ROS	0.620	±0.115	0.745	±0.126	0.567	±0.097	0.592	±0.120	0.710	±0.111
RUS	0.640	±0.122	0.692	±0.130	0.595	±0.117	0.624	±0.122	0.688	±0.121
ENN	0.602	±0.113	0.695	±0.130	0.561	±0.102	0.601	±0.126	0.674	±0.125
				(Classic	Ensembl	es			
ADB1	0.602	±0.088	0.750	±0.112	0.575	±0.099	0.530	±0.121	-	-
M1	0.614	±0.092	0.726	±0.121	0.575	±0.099	0.556	±0.115	-	-
BAG	0.742	±0.105	0.755	±0.110	0.677	±0.111	0.609	±0.115	-	-
Decorate	0.681	±0.125	0.738	±0.113	0.569	±0.098	0.609	±0.124	-	-
				Boos	ting-ba	sed Ense	embles			
SBO	0.652	±0.113	0.669	±0.129	0.573	±0.098	0.577	±0.143	-	-
RUSB	0.672	±0.113	0.675	±0.128	0.616	±0.104	0.628	±0.126	-	-
				Bagg	ing-bas	ed Ense	mbles			
OBAG	0.758	±0.088	0.755	±0.109	0.677	±0.110	0.611	±0.114	-	-
UBAG	0.758	±0.088	0.735	±0.107	0.685	±0.107	0.652	±0.108	-	-
SBAG	0.736	±0.092	0.735	±0.106	0.681	±0.110	0.630	±0.116	-	-
				Cost-	sensitiv	e Classif	ication			
MetaCost	0.620	±0.115	0.728	±0.125	0.564	±0.096	0.605	±0.129	-	-
CS-Classifier	0.641	±0.112	0.757	±0.124	0.567	±0.098	0.500	±0.000	-	-
-		Class	s-balano	ced Ense	mbles v	vith a Co	st-sens	sitive Cla	ssifier	
CS-OBAG	0.746	±0.083	0.755	±0.108	0.677	±0.111	0.607	±0.113	-	-
CS-UBAG	0.755	±0.086	0.737	±0.106	0.686	±0.113	0.643	±0.114	-	-
CS-SBAG	0.733	±0.089	0.735	±0.107	0.681	±0.110	0.629	±0.116	-	-

In bold are highlighted those learning techniques that built prediction models with AUC scores >0.7.

Supplementary file 14. AUC results (mean and standard deviation) of the isometric hip abduction and adduction strength data set (DS 7) for the five base classifiers in isolation and after applying in them the resampling, ensemble (Classic, Boosting-based, Bagging-based and Class-balanced ensembles) and cost-sensitive learning techniques selected

	Base classifiers									
Technique	C4.5 ADTree SMO KNN					RF				
	A	UC	A	UC	A	UC	A	UC	AUC	
None	0.520	±0.095	0.510	±0.130	0.491	±0.040	0.614	±0.122	0.567	±0.123
				Re	samplir	ng Techn	iques			
SMOTE	0.563	±0.132	0.527	±0.135	0.479	±0.095	0.605	±0.119	0.562	±0.125
ROS	0.534	±0.117	0.522	±0.139	0.495	±0.104	0.621	±0.122	0.566	±0.123
RUS	0.539	±0.122	0.521	±0.141	0.498	±0.112	0.557	±0.139	0.558	±0.137
ENN	0.507	±0.096	0.512	±0.133	0.493	±0.055	0.591	±0.134	0.556	±0.130
					Classic	Ensemb	les			
ADB1	0.578	±0.133	0.524	±0.131	0.530	±0.118	0.600	±0.119	-	-
M1	0.569	±0.131	0.531	±0.132	0.524	±0.120	0.563	±0.122	-	-
BAG	0.501	±0.116	0.531	±0.128	0.496	±0.121	0.635	±0.124	-	-
Decorate	0.553	±0.124	0.572	±0.128	0.491	±0.040	0.568	±0.133	-	-
				Boos	sting-ba	sed Ens	embles			
SBO	0.540	±0.131	0.501	±0.132	0.521	±0.130	0.614	±0.128	-	-
RUSB	0.542	±0.134	0.533	±0.133	0.524	±0.131	0.568	±0.136	-	-
				Bagg	ging-ba	sed Ense	mbles			
OBAG	0.570	±0.124	0.535	±0.131	0.505	±0.118	0.638	±0.124	-	-
UBAG	0.538	±0.135	0.543	±0.129	0.501	±0.117	0.608	±0.132	-	-
SBAG	0.563	±0.122	0.531	±0.130	0.508	±0.118	0.626	±0.122	-	-
				Cost	-sensitiv	e Classi	fication	l		
MetaCost	0.501	±0.093	0.500	±0.135	0.494	±0.066	0.585	±0.129	-	-
CS-Classifier	0.522	±0.100	0.514	±0.130	0.492	±0.074	0.614	±0.123	-	-
		Clas	s-balan	ced Ense	embles	with a C	ost-sen	sitive Cla	assifier	
CS-OBAG	0.574	±0.125	0.535	±0.130	0.523	±0.118	0.637	±0.124	-	-
CS-UBAG	0.545	±0.123	0.526	±0.125	0.525	±0.119	0.608	±0.132	-	-
CS-SBAG	0.571	±0.127	0.533	±0.130	0.522	±0.117	0.628	±0.122	-	-

Supplementary file 15. AUC results (mean and standard deviation) of the dynamic postural control data set (DS 6) for the five base classifiers in isolation and after applying in them the resampling, ensemble and cost-sensitive learning techniques selected

					Base cl	assifiers				
Technique	C4.5		AD	Tree	SN	МО	K	NN	R	RF.
	AU	UC	A	UC	A	UC	A	UC	AUC	
None	0.606	±0.127	0.644	±0.119	0.527	±0.091	0.587	±0.132	0.564	±0.133
				Re	esampling	g Techniq	ues			
SMOTE	0.634	±0.129	0.652	±0.115	0.623	±0.115	0.590	±0.138	0.571	±0.142
ROS	0.590	±0.123	0.640	±0.119	0.607	±0.117	0.564	±0.132	0.560	±0.141
RUS	0.619	±0.130	0.623	±0.127	0.601	±0.124	0.602	±0.136	0.610	±0.134
ENN	-	-	0.638	±0.128	0.533	±0.097	0.579	±0.143	0.575	±0.138
					Classic I	Ensemble	5			
ADB1	0.618	±0.125	0.609	±0.130	0.578	±0.121	0.544	±0.127	-	-
M1	0.633	±0.125	0.674	±0.130	0.606	±0.121	0.564	±0.124	-	-
BAG	0.624	±0.123	0.675	±0.118	0.582	±0.127	0.591	±0.135	-	-
Decorate	0.508	±0.132	0.616	±0.133	0.518	±0.079	0.521	±0.139	-	-
				Вос	osting-bas	sed Ensen	nbles			
SBO	0.580	±0.135	0.574	±0.160	0.662	±0.139	0.571	±0.136	-	-
RUSB	0.594	±0.125	0.605	±0.132	0.600	±0.134	0.591	±0.136	-	-
				Вая	ging-bas	ed Ensem	bles			
OBAG	0.642	±0.124	0.674	±0.122	0.630	±0.128	0.586	±0.134	-	-
UBAG	0.677	±0.115	0.677	±0.119	0.641	±0.129	0.619	±0.137	-	-
SBAG	0.641	±0.133	0.671	±0.120	0.628	±0.131	0.592	±0.140	-	-
				Cos	t-sensitiv	e Classific	cation			
MetaCost	0.569	±0.113	0.659	±0.122	0.541	±0.101	0.585	±0.146	-	-
CS-Classifier	0.592	±0.126	0.644	±0.117	0.540	±0.105	0.591	±0.134	-	-
-		C	lass-bala	nced Ens	sembles w	vith a Cos	t-sensitiv	e Classifie	er	
CS-OBAG	0.663	±0.125	0.674	±0.120	0.647	±0.131	0.582	±0.134	-	-
CS-UBAG	0.701	±0.114	0.680	±0.117	0.657	±0.128	0.605	±0.139	-	-
CS-SBAG	0.663	±0.130	0.674	±0.120	0.638	±0.130	0.592	±0.138	-	-

In bold are highlighted those learning techniques that built prediction models with AUC scores > 0.7

Supplementary file 16. AUC results (mean and standard deviation) of the measures obtained through questionnaires data set (DS 6) for the five base classifiers in isolation and after applying in them the resampling, ensemble (Classic, Boosting-based, Bagging-based and Class-balanced ensembles) and cost-sensitive learning techniques selected.

	Base classifiers									
Technique	C4.5	Al	DTree	SI	МО	K	NN	1	RF	
	AUC	I	AUC		AUC		AUC		UC	
None	0.460 ±0.0	0.506	±0.133	0.518	±0.096	0.496	±0.136	0.443	±0.131	
			Res	sampling	g Techni	ques				
SMOTE	0.508 ±0.1	137 0.528	3 ±0.137	0.517	±0.100	0.458	±0.130	0.445	±0.135	
ROS	0.451 ±0.3	113 0.510	±0.133	0.527	±0.100	0.485	±0.134	0.446	±0.124	
RUS	0.480 ±0.2	125 0.515	±0.135	0.527	±0.125	0.517	±0.139	0.469	±0.131	
ENN	0.474 ±0.0	0.505	±0.131	0.518	±0.102	0.498	±0.140	0.467	±0.131	
			(Classic I	Ensembl	es			_	
ADB1	-	0.505	±0.105	0.524	±0.113	0.489	±0.126	-	-	
M1	0.479 ±0.0	0.497	±0.107	0.527	±0.111	0.483	±0.121	-	-	
BAG	0.489 ±0.2	128 0.515	±0.130	0.548	±0.133	0.502	±0.133	-	-	
Decorate	0.468 ±0.7	135 0.494	±0.138	0.530	±0.099	0.455	±0.138	-	-	
			Boos	sting-bas	sed Ense	mbles			_	
SBO	0.504 ±0.7	112 0.506	±0.122	-	-	0.470	±0.139	-	-	
RUSB	0.495 ±0.2	115 0.508	±0.104	0.530	±0.127	0.518	±0.134	-	-	
			Bagg	ging-bas	ed Ense	mbles				
OBAG	0.468 ±0.7	126 0.516	±0.129	0.549	±0.133	0.490	±0.130	-	-	
UBAG	0.509 ±0.1	134 0.529	±0.128	0.558	±0.136	0.519	±0.133	-	-	
SBAG	0.537 ±0.3	124 0.532	±0.128	0.544	±0.133	0.498	±0.134	-	-	
			Cost-	sensitiv	e Classif	ication				
MetaCost	0.466 ±0.0	0.500	±0.128	0.533	±0.105	0.478	±0.129	-	-	
CS-Classifier	0.450 ±0.7	102 0.507	±0.130	0.530	±0.102	0.496	±0.138	-	-	
	(Class-balar	nced Ense	embles v	vith a Co	st-sensi	itive Clas	ssifier		
CS-OBAG	0.477 ±0.2	125 0.518	3 ±0.128	0.550	±0.135	0.486	±0.132	-	-	
CS-UBAG	0.515 ±0.7	127 0.530	±0.131	0.556	±0.137	0.516	±0.135	-	-	
CS-SBAG	0.537 ±0.7	123 0.532	2 ±0.128	0.548	±0.133	0.499	±0.135	-	-	

Supplementary file 17. AUC results (mean and standard deviation) of the field-based tests of neuromuscular performance data set (DS 6) for the five base classifiers in isolation and after applying in them the resampling, ensemble and cost-sensitive learning techniques selected

					Base cl	assifiers	3			
Technique	C	C4.5 ADTree		SI	МО	K	KNN		RF	
	A	UC	A	UC	A	UC	A	UC	A	UC
None	0.598	±0.097	0.758	±0.084	0.563	±0.075	0.747	±0.098	0.742	±0.100
				Res	sampling	g Techni	ques			
SMOTE	0.718	±0.105	0.753	±0.088	0.685	±0.112	0.740	±0.101	0.737	±0.105
ROS	0.704	±0.110	0.760	±0.090	0.685	±0.126	0.749	±0.101	0.745	±0.100
RUS	0.679	±0.118	0.749	±0.093	0.675	±0.124	0.745	±0.100	0.742	±0.105
ENN	0.584	±0.098	0.756	±0.091	0.559	±0.075	0.747	±0.102	0.738	±0.105
				(Classic I	Ensemble	es			
ADB1	0.756	±0.094	0.763	±0.086	0.776	±0.088	0.738	±0.101	-	-
M1	0.759	±0.086	0.751	±0.093	0.757	±0.091	0.748	±0.101	-	-
BAG	0.727	±0.088	0.763	±0.087	0.661	±0.127	0.756	±0.094	-	-
Decorate	0.710	±0.102	0.732	±0.095	0.564	±0.075	0.708	±0.108	-	-
				Boos	sting-bas	sed Ense	mbles			
SBO	0.739	±0.104	0.747	±0.104	0.749	±0.102	0.735	±0.102	-	-
RUSB	0.751	±0.091	0.759	±0.089	0.758	±0.089	0.745	±0.097	-	-
				Bagg	ging-bas	ed Enser	nbles			
OBAG	0.753	±0.089	0.766	±0.087	0.750	±0.099	0.759	±0.096	-	-
UBAG	0.747	±0.084	0.755	±0.087	0.752	±0.094	0.758	±0.092	-	-
SBAG	0.769	±0.099	0.776	±0.092	0.771	±0.101	0.769	±0.100	-	-
				Cost-	sensitiv	e Classif	ication			
MetaCost	0.539	±0.081	0.724	±0.110	0.500	±0.000	0.519	±0.200	-	-
CS-Classifier	0.641	±0.112	0.756	±0.087	0.500	±0.000	0.751	±0.099	-	-
		Clas	ss-balan	ced Ense	embles v	vith a Co	st-sensi	tive Clas	sifier	
CS-OBAG	0.759	±0.095	0.767	±0.088	0.760	±0.103	0.763	±0.097	-	-
CS-UBAG	0.748	±0.089	0.757	±0.088	0.767	±0.096	0.761	±0.095	-	-
CS-SBAG	0.770	±0.104	0.776	±0.092	0.768	±0.100	0.772	±0.101	-	

In bold are highlighted those learning techniques that built prediction models with AUC scores >0.7.

Supplementary file 18. AUC results (mean and standard deviation) of the global data set (DS 11) for the five base classifiers in isolation and after applying in them the resampling, ensemble (Classic, Boosting-based, Bagging-based and Class-balanced ensembles) and cost-sensitive learning techniques selected

					Base cl	assifiers				
Technique	C	C4.5 ADTree SMO KNN		NN	I	RF				
•	A	UC	A	UC	AUC		AUC		A	UC
None	0.642	±0.124	0.741	±0.119	0.568	±0.086	0.704	±0.131	0.713	±0.135
	Resampling Techniques									
SMOTE	0.709	±0.130	0.738	±0.121	0.651	±0.128	0.700	±0.129	0.711	±0.139
ROS	0.694	±0.130	0.738	±0.122	0.659	±0.127	0.704	±0.131	0.712	±0.136
RUS	0.663	±0.131	0.720	±0.126	0.645	±0.129	0.698	±0.120	0.708	±0.137
ENN	0.637	±0.123	0.731	±0.124	0.567	±0.093	0.697	±0.130	0.707	±0.136
					Classic I	Ensemble	s			_
ADB1	0.746	±0.124	0.769	±0.131	0.722	±0.138	0.691	±0.135	-	-
M1	0.754	±0.110	0.742	±0.144	0.797	±0.131	0.690	±0.136	-	-
BAG	0.740	±0.115	0.743	±0.116	0.694	±0.131	0.716	±0.127	-	-
Decorate	0.709	±0.127	0.720	±0.124	0.569	±0.087	0.676	±0.141	-	-
				Воо	sting-bas	sed Enser	nbles			
SBO	0.715	±0.138	0.749	±0.061	0.740	±0.102	0.707	±0.132	-	-
RUSB	0.736	±0.121	0.748	±0.138	0.752	±0.118	0.710	±0.128	-	-
				Bag	ging-base	ed Ensen	nbles			
OBAG	0.744	±0.112	0.741	±0.116	0.742	±0.125	0.720	±0.126	-	-
UBAG	0.742	±0.111	0.739	±0.119	0.737	±0.121	0.719	±0.120	-	-
SBAG	0.751	±0.118	0.745	±0.119	0.750	±0.124	0.724	±0.125	-	-
				Cost	-sensitive	e Classifi	cation			
MetaCost	0.572	±0.120	0.698	±0.134	0.500	±0.000	0.604	±0.147	-	-
CS-Classifier	0.685	±0.129	0.739	±0.124	0.500	±0.000	0.706	±0.128	-	-
		(Class-bala	nced Ens	embles w	vith a Cos	st-sensitiv	ve Classif	ier	
CS-OBAG	0.751	±0.107	0.742	±0.115	0.747	±0.121	0.715	±0.126	-	-
CS-UBAG	0.749	±0.105	0.741	±0.119	0.747	±0.116	0.722	±0.124	-	-
CS-SBAG	0.755	±0.115	0.746	±0.119	0.750	±0.121	0.719	±0.127	_	-

In bold are highlighted those learning techniques that built prediction models with AUC scores >0.7.

Supplementary file 19: schemes of the algorithms selected in data sets (DS) 6, 8, 10 and 11

Lower extremity joint ranges of motion (DS – 6)

CS-Classifier [ADTree]

weka.classifiers.meta.MultiSearch -E FM -search "weka.core.setupgenerator.MathParameter -property classifier.numOfBoostingIterations -min 5.0 -max 50.0 -step 1.0 -base 10.0 -expression I" -class-label 1 - algorithm "weka.classifiers.meta.multisearch.DefaultSearch -sample-size 100.0 -initial-folds 2 - subsequent-folds 10 -initial-test-set . -subsequent-test-set . -num-slots 1" -log-file /Applications/weka-3-8-3 -S 1 -W weka.classifiers.meta.CostSensitiveClassifier -- -cost-matrix "[0.0 2.0; 1.0 0.0]" -S 1 -W weka.classifiers.trees.ADTree -- -B 10 -E -3 -S 1

Dynamic postural control (DS - 8)

CS-UBAG [C4.5]

weka.classifiers.meta.MultiSearch -E FM -search "weka.core.setupgenerator.MathParameter -property classifier.classifier.classifier.confidenceFactor -min 0.05 -max 0.75 -step 0.05 -base 10.0 -expression I" - class-label 1 -algorithm "weka.classifiers.meta.multisearch.DefaultSearch -sample-size 100.0 -initial-folds 2 -subsequent-folds 10 -initial-test-set . -subsequent-test-set . -num-slots 1" -log-file /Applications/weka-3-8-3 -S 1 -W weka.classifiers.meta.Bagging -- -P 100 -S 1 -num-slots 1 -I 100 -W weka.classifiers.meta.FilteredClassifier -- -F "weka.filters.supervised.instance.RUS -P 60.0" -S 1 -W weka.classifiers.meta.CostSensitiveClassifier -- -cost-matrix "[0.0 2.0; 1.0 0.0]" -S 1 -W weka.classifiers.trees.J48 -- -C 0.25 -M 2

Neuromuscular measures from field-based tests (DS – 10)

CS-UBAG [SMO]

weka.classifiers.meta.AttributeSelectedClassifier - E "weka.attributeSelection.CfsSubsetEval - P 1 - E 1" - S "weka.attributeSelection.GreedyStepwise -B -T -1.7976931348623157E308 -N -1 -num-slots 1" -W weka.classifiers.meta.MultiSearch -- -E AUC -search "weka.core.setupgenerator.MathParameter property classifier.classifier.calibrator.ridge -min -10.0 -max 5.0 -step 1.0 -base 10.0 -expression pow(BASE,I)" -class-label 1 -algorithm "weka.classifiers.meta.multisearch.DefaultSearch -sample-size 100.0 -initial-folds 2 -subsequent-folds 10 -initial-test-set . -subsequent-test-set . -num-slots 1" -log-file /Applications/weka-3-8-3 -S 1 -W weka.classifiers.meta.Bagging -- -P 100 -S 1 -num-slots 1 -I 100 -W weka.classifiers.meta.FilteredClassifier -- -F "weka.filters.supervised.instance.RUS -P 60.0" -S 1 -W weka.classifiers.meta.CostSensitiveClassifier -- -cost-matrix "[0.0 2.0; 1.0 0.0]" weka.classifiers.functions.SMO -- -C 1.0 -L 0.001 -P 1.0E-12 -N 0 -V -1 -W 1 -K "weka.classifiers.functions.supportVector.PolyKernel -C 250007" -calibrator "weka.classifiers.functions.Logistic -R 1.0E-8 -M -1 -num-decimal-places 4"

Global (DS - 11)

CS-UBAG [C4.5]

weka.classifiers.meta.AttributeSelectedClassifier -E "weka.attributeSelection.CfsSubsetEval -P 1 -E 1" -S "weka.attributeSelection.GreedyStepwise -B -T -1.7976931348623157E308 -N -1 -num-slots 1" -W weka.classifiers.meta.MultiSearch -E FM -search "weka.core.setupgenerator.MathParameter -property classifier.classifier.classifier.confidenceFactor -min 0.05 -max 0.75 -step 0.05 -base 10.0 -expression I" - class-label 1 -algorithm "weka.classifiers.meta.multisearch.DefaultSearch -sample-size 100.0 -initial-folds 2 -subsequent-folds 10 -initial-test-set . -subsequent-test-set . -num-slots 1" -log-file /Applications/weka-3-8-3 -S 1 -W weka.classifiers.meta.Bagging -- -P 100 -S 1 -num-slots 1 -I 100 -W weka.classifiers.meta.FilteredClassifier -- -F "weka.filters.supervised.instance.RUS -P 60.0" -S 1 -W weka.classifiers.meta.CostSensitiveClassifier -- -cost-matrix "[0.0 2.0; 1.0 0.0]" -S 1 -W weka.classifiers.trees.J48 -- -C 0.25 -M 2