## A step forward the democratization of screening tools for ankle sprain risk

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# **Summary**

Ankle sprains are common injuries affecting both sports and daily-life activities. A traditional screening tool, relying on expensive instruments, was transformed into a low-cost solution, using machine learning and smartphone data making injury risk assessment accessible to more athletes in the field. A NuSVC binary classifier, fed with anthropometric data and smartphone features of 126 participants, achieved 70% test phase accuracy to predict ankle risk of injury.

# Introduction

Injury risk is a major factor that influences both individual players and team performance [1]. When the injury affects the lower limbs, the limitations affect both sports and daily life. Methods and tools are developed that may assess injury risk in sport contexts. Among these, a proof-of-concept able to analyze the single leg drop jump landing (SLDJ) test using a combination of a smartphone accelerometer and a machine learning classifier was developed for ankle sprain injury risk [2]. This approach could aid in identifying individuals at risk while also guiding preventive strategies for vulnerable athletes. In this work, we want to verify if the democratic approach proposed in [2] is still valid with wider population.

## Methods

A group of 126 physically active participants (Table 1) was tested while performing the SLDJ procedure proposed by Fransz [3], jumping on a force plate (FP, Bertec, Bertec Corporation, Columbus, OH, USA; fs = 1000 samples/s; size =  $40 \times 40$  cm), while hand—holding a smartphone (SP, Samsung Galaxy S9+, Samsung Group, Seoul, Republic of Korea; fs = 500 samples/s; full-scale range: accelerometer =  $\pm$  8 g; gyroscope =  $\pm$  500 deg/s), as proposed in [2]. Each participant performed the SLDJ once per leg.

**Table 1:** Anthropometric data of males (M) and females (F) participants (mean  $\pm$  standard deviation).

|   | Number | Age (y)    | Mass (kg)   | Height (m)      |
|---|--------|------------|-------------|-----------------|
| M | 78     | $22 \pm 5$ | $72 \pm 13$ | $1.79\pm0.08$   |
| F | 48     | 25 ± 4     | $60 \pm 8$  | $1.64 \pm 0.07$ |

After data acquisition, the two risk factors (RMS ML, HOR GRF late) defined in [3] were calculated from both FP and SP signals through MATLAB R2022a (The MathWorks Inc., Natick, MA, USA). Risk factors computed with FP and reference values [3] were used for classification. Participants were considered at risk (class 1) if the combination of their FP

risk factors resulted in a risk-level higher than one relative to a participant with average scores, or belonging to class 0, if otherwise.

Age, height, mass and SP risk factors (calculated from the accelerometer, as in [2]) are used as input variables of a binary NuSVC classifier developed in Jupyterlite Notebook (v. 0.5.0) to predict the classification obtained using FP risk factors. After dividing train/test sets (80-20% of the participants) and a z-score normalization step, hyperparameter tuning was performed using GridSearchCV to find the best binary classifier, considering GroupKFold approach for validation to ensure data leakage.

#### **Results and Discussion**

A final dataset of 246 trials (49 classified as 0, 197 as 1) was used to develop the classifier (Table 2).

**Table 2:** Hyperparameter options and ranges for NuSVC with the corresponding best model hyperparameters values and accuracy.

| Hyperparameters       | Options/ranges                                       | Model   |
|-----------------------|--|---------|
| Kernel                | 'linear', 'poly', 'rbf', 'sigmoid',<br>'precomputed' | 'poly'  |
| Degree                | [1,2,3,4,5]  | 5       |
| Gamma                 | ['scale', 'auto']                                    | 'scale' |
| Nu                    | [0.1:0.1:1]  | 0.3     |
|                       |  |         |
| Accuracy (train-test) | -  | 74-70%  |

The accuracy of the model is promising because obtained from a bigger population relative to [2]. Since current reference values and the risk assessment model in [3] only included a male young population, longitudinal studies including females and older male population are required to revise current classification and further develop this classifier.

### **Conclusions**

Smartphone-based solutions could help in democratizing the screening process of ankle sprains injury risk.

### References

- [1] Hägglund M et al. (2013). Br. J. Sports Med., 47:738–742
- [2] De Lazzari B et al. (2023). In *Proc XXIII SIAMOC 2023*, pp. 52 [Online].
- [3] Fransz DP et al (2018). Am. J. Sports Med., 46, 3454–3462.