

Financial Disclosure Statement

No Financial Disclosures

Background

- Severe injuries in high school sports have both physical and emotional ramifications.¹⁻³
- **Sport-specific guidelines** must continue to evolve to protect athletes from these injuries.
 - Growing role of machine learning (ML)⁴



Objective

To apply machine and deep learning techniques to predict shoulder injury severity in a national high school sports injury database.

Methods



Return to Sport (RTS) and predictor variables selected.

ML models tested and feature importance analysis performed.⁵

- Balanced random forest (RF)
- Elastic-net regression (ENet)
- Gradient boosted tree (GBM)
- Neural net (NN)

Results

Injury Demographics

- A total of 2,405 injuries were studied (mean age = 16.1 ± 1.2 years).
- Level of play was categorized as freshman (8.6%), junior varsity (21.5%), varsity (66.1%), and other (3.8%).
- Time of season was categorized by preseason (24.8%), in-season (70.6%), postseason (4.5%), and other (0.8%).
- 355 injuries (14.8%) experienced a prolonged return to sport (RTS).

Results

Performance of Machine Learning Models

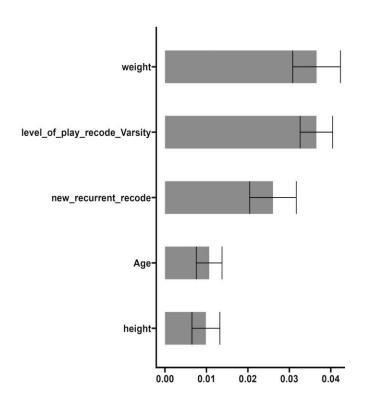
The <u>Gradient Boosted Tree (GBM)</u> had the best performance, marginally outperforming the other models.

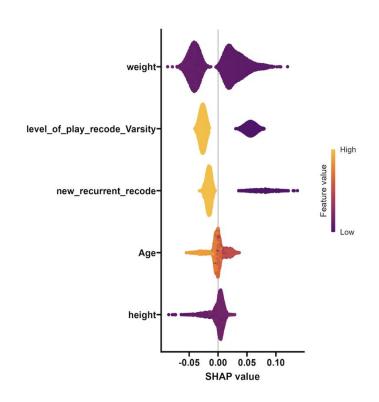
Outcome	Benchmark AUC (± 95% CI)	Gradient Boosted Tree AUC (± 95% CI)	Elastic Net AUC (± 95% CI)	Random Forest AUC (± 95% CI)	Neural Network AUC (± 95% CI)
Prolonged Return to Sport (n = 355)	0.5 ± 0.00	0.61 ± 0.01	0.60 ± 0.01	0.60 ± 0.02	0.57 ± 0.02

Results

Feature Importance Analysis of GBM

Advanced <u>level of play</u>, heavier <u>weight</u>, and <u>recurrent injury status</u> were the top three injury predictors.





Conclusion

- Given the results of the feature importance analysis, we suggest that injury prevention efforts focus on high school football players with <u>recurrent injuries</u> and those of <u>higher weights</u> and <u>levels of play</u>.
- This study suggests the <u>utility of machine learning</u> models to help generate guidelines for athlete injury prevention.

References

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