Optimizing Rugby Training Under Fatigue

Data Me Down

Justin Ferrara | justin.c.ferrara@asu.edu

Nomith Murari | nsmurari@asu.edu

Farhan Khera | Farhan.khera@asu.edu

W. P. Carey School of Business | Tempe Campus





Properly defining and measuring fatigue will lead to significant increase in player performance and well-being

Problem Statement

In our analysis, the Data Me Down team attempts to identify, using regression techniques:

- (1) the factors that allow us to **define** and **measure fatigue**, and
- (2) how we can **optimize** our **program** around **predicted fatigue** throughout the season to **maximize game performance**.

Solution

Our solution consists of a two-prong load management approach.

- (1) We have created an interactive line chart that shows the predicted fatigue of all players on each given day. This is used to **assign a training regime** that will allow a player to perform training that will push them **optimally.**
- (2) Graphic that shows **how many times** each training regime was prescribed for a **given player**. Based on this figure, we can identify ways to **improve certain types of training** or **identify players** that spend **greater than average time with high fatigue** and address their lifestyle habits to reduce time spent in fatigue.

Fatigue Over Time **Predicted Player Fatigue Each Day** Data for Player 5 Training 1. Recovery Predicted Fatigue 2. Skills 3. Conditioning 4. Speed 5. Strenath 6. Combat 7. Game Apr 2018 Oct 2017 Jan 2018 Jul 2018 Date

Training Regime Quantity/Player





Optimization of training regime will produce significant economic impact

| Regression Predicting Fatigue | | | | | | | | |
|-------------------------------|---------|---------|-----------------------|-------|---|--|--|--|
| Predicted Fatigue | Coef. | P-value | Model Characteristics | | Comments | | | |
| Soreness | 0.2601 | 0.000 | R-square | 0.961 | We ran a number of models to accurately predict | | | |
| Desire | 0.3215 | 0.000 | F-sig | 0.000 | and define "Fatigue"; this model was the most accurate for our purposes. All predictors are significant and 96.1% of the variation in y-hat is explained by our predictors. This model also fulfills all 4 key assumptions to validate our model. | | | |
| Irritability | 0.1616 | 0.000 | | | | | | |
| Illness | -0.1065 | 0.000 | | | | | | |
| SleepIndex | 0.0233 | 0.000 | | | · | | | |

Economic Impact

| Cost Savings | |
|----------------------------------|-------------|
| Injuries per Season | 22 |
| Expected Value of Cost per Injur | \$159,000 |
| Total Cost Savings on Injury | \$3,498,000 |

| Revenue Generation | | | |
|----------------------------|-----------|--|--|
| Sponsorships | \$539,000 | | |
| Sale of Merchandise | \$286,000 | | |
| Total Increase in Revenues | \$825,000 | | |

| Total Financial Benefit | \$4,323,000 |
|-------------------------|-------------|
|-------------------------|-------------|