



Trinity College Dublin

Coláiste na Tríonóide, Baile Átha Cliath

The University of Dublin

School of Statistics and Computer Science

Machine Learning to go *nyoom*

Using Machine Learning to evaluate rowing training and predict
training outcomes or performances

Liam Junkermann

Supervisor: Dr. Lucy Hederman

February 22, 2024

A Final Year Project submitted in partial fulfilment
of the requirements for the degree of
BA(Mod) in Science in Computer Science

Declaration

I hereby declare that this Final Year Project is entirely my own work and that it has not been submitted as an exercise for a degree at this or any other university.

I have read and I understand the plagiarism provisions in the General Regulations of the University Calendar for the current year, found at <http://www.tcd.ie/calendar>.

I have completed the Online Tutorial on avoiding plagiarism 'Ready Steady Write', located at <http://tcd-ie.libguides.com/plagiarism/ready-steady-write>.

I consent / do not consent to the examiner retaining a copy of the thesis beyond the examining period, should they so wish (EU GDPR May 2018).

Signed: _____

Date: _____

Abstract

A short summary of the problem investigated, the approach taken and the key findings. This should not be more than around 400 words.

This must be on a separate page.

Lay Abstract

Similar to the actual abstract in terms of the information, but written for a non-specialist. So no jargon, no acronyms. Explain to a member of the general public what this project entailed. Should be no longer than the actual abstract.

This must be on a separate page.

Acknowledgements

Thanks Everyone!

You should acknowledge any help that you have received (for example from technical staff), or input provided by, for example, a company.

Contents

1	Introduction	1
1.1	Motivations	1
1.2	Goals	1
1.3	The Report	1
2	Background	3
2.1	A brief introduction to rowing	3
2.1.1	Training Principles	3
2.1.2	Energy Systems	6
2.1.3	Training Considerations	7
2.1.4	Performance	7
2.1.5	Summary	7
2.2	A Review of Performance Modelling	8
2.2.1	Quantifying Training Load (Fatigue)	8
2.2.2	Impulse-Response Models	10
2.2.3	Alternative Models	12
3	Data Collection and Management	13
4	Data Analysis and Visualization	14
5	Machine Learning Applications	15
5.1	Model Considerations	15
5.2	The Implementation	15
6	Discussion	16
7	Conclusion	17
A1	Appendix	20
A1.1	Appendix numbering	20
A1.2	Lactate Test Results	21

List of Figures

2.1	Lactate Profiles from January and May 2023	5
2.2	Borg 20	9
2.3	Borg CR10	9

List of Tables

A1.1 Test 1 Interval Details	21
A1.2 Test 2 Interval Details	21

Vocabulary

Heart Rate Metrics

HRV	Heart Rate Variability
HR	Heart Rate
HR _{max}	Maximum Heart Rate

Training Zones

UT2	Basic Oxygen Utilisation Training
UT1	Oxygen Utilization Training
AT	Anaerobic Threshold Training
TR	Oxygen Transport Training
AN	Anaerobic Capacity Training
LT	Lactate Threshold

General Training Terms

Stroke Rate	The number of strokes completed per minute (like cadence in cycling or running)
Split	The standard split in rowing is time to complete 500m

Chapter 1

Introduction

This chapter will introduce the general approach and motivation for this project, the expected outcomes, and a structure for the remainder of the report.

1.1 Motivations

What motivated this project. Interest in rowing, rowing background, sports science is cool, blah blah blah

1.2 Goals

Target: ML to go fast. Data collection piece, data analysis piece, final model target.

1.3 The Report

How this reports structured, whats going to be covered.

Background - covering intro to sports science for rowing, and terms used for data analysis and modelling. Models which are used atm, why they work well, or don't.

Data Collection and management - Discussing how I collected data (made website and deployed bits, etc). How data was managed (s3 for raw data, dynamo & rds for user generated/processed data).

Data Analysis + visualization - I mean, it does what it says on the box Machine Learning Applications - what was the goal, considerations for the model, what'd we actually make

Discussion - Did we actually do the thing, how can we improve what we did, what stopped us from building the whole thing

Conclusion - uhm, some waffle to tie it all together

Chapter 2

Background

This chapter will cover the basic background of rowing, the sports science which guides rowing training, and how an athlete's body responds to training stimulus. Next, a review of performance modelling will explore the development of human performance modelling since the introduction of the basic Bannister model in 1975 [1]. The section will outline how training load and performance can be quantified, and explain the way these approximations are used in various performance models to date.

2.1 A brief introduction to rowing

Rowing is an Olympic sport, raced across a 2,000 metre course, typically lasting six to seven minutes. It is classed as a power-endurance sport, this means training is focused on building aerobic, anaerobic, and power while also developing rowing technique [2]. Most time is spent building endurance, next most time is spent building anaerobic capacity, finally building strength and power through Strength and Condition sessions [3]. The importance of power is more significant in rowing than in cycling given the relatively short duration of exertions, with longer distance racing typically only covering five to seven kilometers, or fifteen to twenty-five minutes. Road cycling, for example, tends to last for a longer period of time, where the shortest races might last two hours. There are many different approaches to how training is conducted and which energy systems are targeted. This section will discuss the basic training principles which guide training, the way athletes respond to different kinds of training loads, and how performance is evaluated in rowing.

2.1.1 Training Principles

Generally when a coach builds a training plan they have a few factors they can work with: volume, the amount of mileage or total time spent training, sessions intensity, how hard the given session is meant to be, and finally the frequency, or the time spent in different

intensity zones. There are various ways to measure intensity, including heart rate, blood lactate concentration, velocity at maximal oxygen uptake ($\text{VO}_2 \text{ max}$), and rate of perceived exertion (RPE) [4]. Rowers tend to use heart rate zones or blood lactate concentration depending on access to the equipment to test blood lactate. Typically when a rower uses calculated aerobic zones each zone will be a percentage of HR_{max} . Following this approach, zones are defined as follows:

Z1 "Very Light" intensity, 50% - 60% of HR_{max}

Z2 "Light" intensity, 60%-70% of HR_{max}

Z3 "Moderate" intensity, 70%-80% of HR_{max}

Z4 "Hard" intensity, 80%-90% of HR_{max}

Z5 "Maximum" intensity, 90%-100% of HR_{max}

The exact definition of these zones varies in the literature, as does the method to calculate HR_{max} without a specific test. However, most high level athletes will have completed some kind of stress test to determine their HR_{max} in order to train more effectively on their prescribed zones. A rower who uses lactate based training zones might use the following zones:

T1 basic oxygen utilization training (UT2) [lactate = 0-2 mmol/L]

T2 oxygen utilization training (UT1) [lactate = 2-3.5 mmol/L]

T3 anaerobic threshold training (AT) [lactate = 3.5-4.5 mmol/L]

T4 oxygen transport training (TR) [lactate = 4.5-6 mmol/L]

T5 anaerobic capacity training (AN) [lactate \geq 6 mmol/L] [5]

Depending on how rigorous the testing protocol was, Heart Rate zones may be calculated for each zone, these may vary from the aerobic zones calculated from HR_{max} .

Blood Lactate - A brief description

Lactate is constantly produced by the body during the day. The concentration of blood lactate, measured in millimoles per litre (mmol/L), does not increase until the rate of lactate production surpasses the rate of lactate removal. Many things can affect the rate of lactate removal, training can improve the rate of lactate removal, with certain sessions targeting that adaptation. Blood lactate acts as a biomarker used regularly to determine muscular fatigue during exercise. During an exercise lactate test a lactate profile is generated. Below is a chart showing two lactate profiles generated during two lactate tests about 14 weeks apart (98 days) in January and May of 2023. The test conducted saw the rower complete seven,

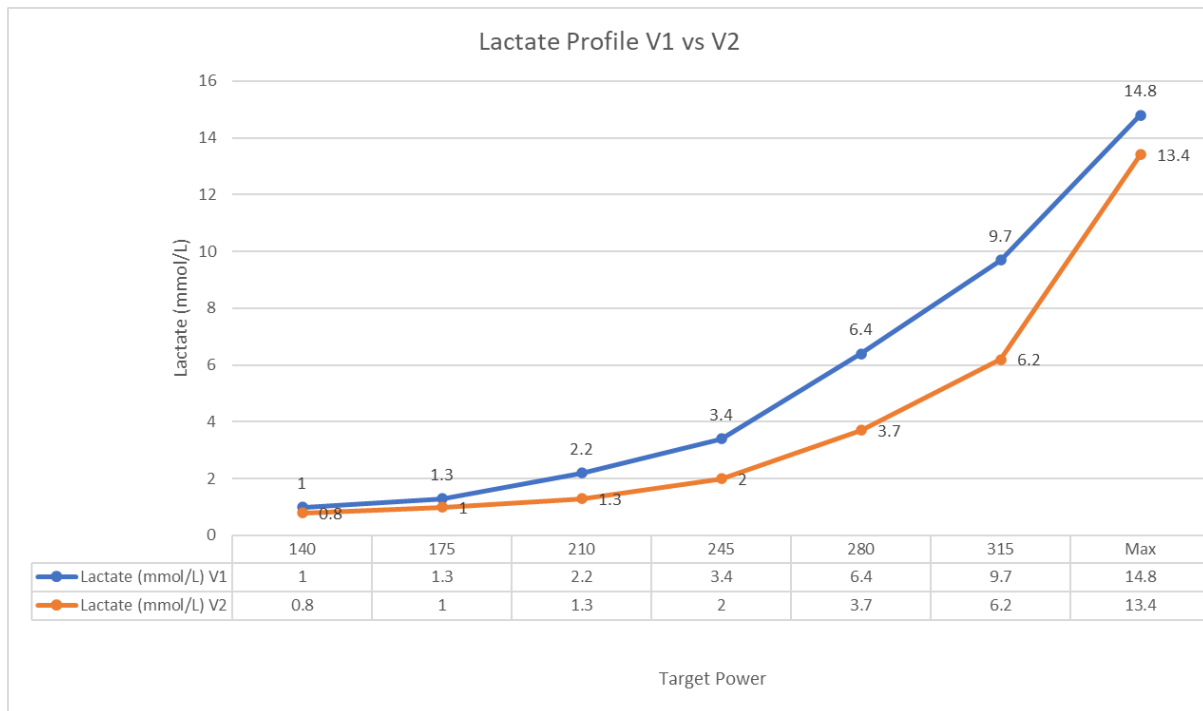


Figure 2.1: Two lactate profiles completed 98 days apart following the protocol described above. Test one completed January 24, 2023. Test Two completed May 2, 2023

four minute, efforts, at prescribed wattage targets. Details of each set of intervals can be found in A1.2.

The most basic zone approximation approach uses three zones based around certain physiological thresholds, like, lactate thresholds (LT_1 and LT_2) and ventilatory thresholds. Cyclists may use critical power to determine these three basic zones, although this practice has not become popular in rowing training. The zones become simply, low-intensity, moderate-intensity, and high-intensity.

There are a few different approaches for distributing intensity for endurance training. The three main methods are: polarised training, sweet spot or threshold training, and pyramidal training. This guides the final factor a coach considers when building a general training plan, frequency. For the purposes of comparing polarized training (POL), threshold training (THR), and pyramidal training (PYR), the more basic three zones of intensity will be used. The breakdown per zone for each training method is as follows:

Polarised Training Far more time spent in the low-intensity zone [6].

Low-Intensity 75%-85% of total training volume

Medium-Intensity 5%-10% of total training volume

High-Intensity 5%-10% of total training volume

Threshold Training More time spent in the medium-intensity zone [6].

Low-Intensity 45%-55% of total training volume

Medium-Intensity 35%-55% of total training volume

High-Intensity 15%-20% of total training volume

Pyramidal Training Most time spent in low-intensity zone with progressively less time spent in higher zones [7].

Low-Intensity 75%-85% of total training volume

Medium-Intensity 15%-20% of total training volume

High-Intensity 5%-10% of total training volume

This report will not compare the effectiveness of different training distributions. Different distributions tend to be used by different sports, or depending on which energy system is being targeted. The use of polarized training is most common in rowing [4], although other training approaches have been used, especially around competition time.

2.1.2 Energy Systems

In the human body there are two major types of skeletal muscle fibres, fast twitch, and slow twitch. Slow twitch muscles are used for longer, slower contractions where relative strength is low. Conversely, fast twitch muscles are shorter, faster contractions where strength is relatively higher. These types of fibres store energy differently and respond to training differently. Slow twitch muscles can be adapted to longer, lower effort, sessions and become resistant to fatigue, while fast twitch muscles fatigue easily, due to their lower glycogen capacity. Slow twitch muscle fibres have a low anaerobic capacity, while fast twitch muscle fibres have high anaerobic capacities. These two kinds of fibres draw on two different energy systems: anaerobic and aerobic systems. These systems provide muscles with Adenosine Triphosphate (ATP) which is used by the mitochondria in the muscle fibre cells to produce energy, allowing the muscles to contract [8].

The anaerobic system is used to provide energy to muscles to produce power without the use of oxygen. The anaerobic system typically provides energy for shorter periods of time. An immediate energy system can provide energy for 1-2 seconds of maximal work, this is typically used for resistance based strength training. More commonly in rowers, the short term energy system is used to provide energy to muscle fibres under significant strain. However, each time the energy system produces (ATP), lactate is also produced. This system is limited by an athletes ability to flush lactate from muscles. If the muscles are unable to flush the lactate quickly enough, the muscles will fatigue until failure forcing a stop to exercise.

The aerobic, or long term energy, system is the slowest at providing energy to muscles. This

system uses sugars, fats and oxygen to produce ATP. This system only works when large amounts of oxygen are available, which for rowers is typically during lower intensity sessions. Longer "steady state" sessions rely on the aerobic to provide energy to muscles, with these sessions also being used to build the aerobic system. By spending time in the appropriate training zones the aerobic system builds efficiency by creating new capillaries to support the slow twitch fibres.

Rowers develop both their anaerobic and aerobic systems. Much of the winter season (typically September-March) is spent building the aerobic system, this is due to the longer length of any races done during this period, and to build a larger base on which to build a strong anaerobic system. As the sprint racing season begins, or shortly before, more anaerobic sessions will be introduced to build the system. Rowers will build their lactate tolerance, and spend more time at just-below-maximal efforts to prepare for the shorter race format.

2.1.3 Training Considerations

Overtraining and burnout

Tapering

2.1.4 Performance

There are two primary ways to measure a performance in rowing. First, on the rowing machine, a 2,000 meter, 6,000 meter, or 30 minute at stroke rate 20 (30 r20) test is normally used, depending on when during the season the test is done. This can be used to determine a rowers fitness and can be used to track progress for an athlete. These sessions are typically maximum effort sessions with some degree of taper (typically 4-7 days maximum) to illicit a peak performance. On the water, times can also be used to judge an entire crew, with external factors like wind direction and intensity, current flow, and temperature considered. Some boats and squads may use on the water telemetry to quantify the impact and power output of each rower. On the water performances may not always be peak efforts depending on a squads seasonal focus or an event's progression (eg. a maximal effort will rarely make an appearance in the heat stage of a heat-semi-final progression). Ergometer scores tend to be considered as the purest way of quantifying rowers performance, however a combination of approaches can be used to quantify on the water performance.

2.1.5 Summary

Rowing training is quite time consuming for athletes. Many rowers will learn much of what has been discussed in this section to make better decisions themselves, or understand why training is done in a given way. The focus for a majority of the training cycle is to build a strong aerobic base in order to allow for higher "peak", ideally coinciding with the peak

event for a season of training. For international athletes this will be a yearly cycle targetting World Championships, which is part of a four-yearly cycle targetting the Olympics. For national level athletes, national championships are typically the target, with Henley Royal Regatta featuring as a seasons target. Coaches will normally take into account season targets and rower proficiency and base fitness when generating training plans. The target of this project is to make it easier for coaches and more accessible for athletes who may be more independant.

2.2 A Review of Performance Modelling

The first wide spread approach to modelling performance was developed by Banister *et al.* [1] with the Banister Fitness - Fatigue model, and further refined by Morton *et al.* [9] when defining training impulse to determine fitness and fatigue. The use of machine learning, specifically artifical neural networks has since been introduced in approaches to model performance. In preparation for the 2000 Olympics in Sydney Edelmann-nusser *et al.* [10] successfully predicted olympic performances within an error of 0.05 seconds across a total time of 2:12.64 (min:sec) for the 200m backstroke event. Edelmann-nusser *et al.* [10] specifically consider the limitation of using a linear model, such as the model proposed by Banister *et al.* [1], on training adaptation and performance; adaptation to training is inherently a complex non-linear process.

This section will review basic approaches to systems modelling, exploring metrics which are commonly used to guide models and some non-linear systems models which have been explored.

2.2.1 Quantifying Training Load (Fatigue)

Rate of Perceived Exertion (RPE)

Rate of perceived exertion (RPE) is a scale, originally introduced by Gunnar Borg, to measure an athletes effort. The original scale ranged from 6-20, where 6 would be no exertion at all, and 20 is maximal effort. This scale ranges from 6-20 to correspond more easily with heart rate, as the scale is used beyond just athletic settings. Therefore another scale, ranging from 0-10, was developed by Borg as well for use with "extreme intensity of activity", this is the scale normally used in athletic studies. The second scale is called a Category-Ratio (CR) scale where the reference anchor is RPE CR10 of 10, meaning maximal effort/pain. Session RPE (sRPE), using CR10, can be used, in conjunction with duration to determine session load [11].

Borg RPE	
Score	Level of exertion
6	No exertion at all
7	
7.5	Extremely light
8	
9	Very light
10	
11	Light
12	
13	Somewhat hard
14	
15	Hard (heavy)
16	
17	Very hard
18	
19	Extremely hard
20	Maximal exertion

Figure 2.2: The original Borg 6-20 RPE Scale [11]

Borg CR10 scale	
Score	Level of exertion
0	No exertion at all
0.5	Very, very slight (just noticeable)
1	Very slight
2	Slight
3	Moderate
4	Somewhat severe
5	Severe
6	
7	Very severe
8	
9	Very, very severe (almost maximal)
10	Maximal

Figure 2.3: The Borg Category-Ratio 10 RPE Scale [11]

The two Borg RPE scales

Training Impulse (TRIMP)

Training Impulse (TRIMP) is a method to calculate training load and is defined as $\text{TRIMP} = \text{Training Volume} \times \text{Training Intensity}$ [12]. There are many methods to calculate TRIMP, using different metrics to calculate training volume and training intensity. For the purposes of this project volume will simply be minutes, and intensity will be average heart rate (bpm), as the simplest TRIMP method outlined. Other modifications may apply a weighting against a heart rate metric to normalise longer sessions completed at a lower heart rate – considering the noted difference in heart rate responses to training in male and female athletes [9], alternatively load can be calculated by using the product of RPE and duration (ie. $\text{RPE} \times \text{Duration (minutes)}$) .

Acute Chronic Workload Ratio (ACWR)

Acute Chronic Workload Ratio (ACWR) can be used to monitor load in an athlete. It compares training load accumulated, in arbitrary units (AU), over the last seven days (acute workload) to the training load accumulated over the last twenty-eight days (chronic workload). The exact calculations used to determine acute and chronic workload depends on what type of ACWR model is selected. Additionally, the exact time periods considered for acute and chronic load can change depending on the application [13]. Typically ACWR is used for injury management in team sports, but some articles have found it may not be effective for preventing injury as it is inaccurate way of approximating load [14]. Training load estimation in rowing tends to be more objective given the objective internal load measurement of heart rate measured during a session. There is no clear consensus in

literature about the effectiveness of using ACWR in injury prevention. As a relatively new concept, first published in 2016 more research into its efficacy and to provide validation needs to be completed [15]. Regardless of its effectiveness in reducing injury, ACWR charts are easy to produce and provide easy feedback to rowers to see how training load can change week-to-week and ensuring not load is not increased too quickly to begin to risk overtraining or injury.

2.2.2 Impulse-Response Models

Impulse-Response models are a group of models, built on the model initially introduced by Banister *et al.* [1]. These models assume that a single exercise, or training session, produces two responses: Fitness, or a positive performance response, and Fatigue, or a negative performance response. A simplified version of this model is defined as

$$Performance = Fitness - Fatigue \quad (2.1)$$

This simplified model is commonly called the Fitness-Fatigue impulse response model (FFM) and research since its initial introduction in 1975 has sought to refine how the *Fitness* and *Fatigue* components of this equation are defined.

Banister Fitness-Fatigue Model

In the FFM model originally introduced in 1975 [1], and built upon by Morton *et al.* [9] in 1990, the functions for fitness and fatigue were defined, considering a cumulative training load and a decay factor for each response. This decay factor is different for both the fitness and fatigue responses, resulting in a longer decay period for fitness than fatigue, but also differs from person to person, offering a parameter to be tuned for each athlete. First, a session needs to be quantified. A session is defined as:

$$w(t) = D(\Delta HR \text{ ratio}) \quad (2.2)$$

where

$$\Delta HR \text{ ratio} = \frac{HR_{\text{exercise}} - HR_{\text{rest}}}{HR_{\text{max}} - HR_{\text{rest}}}$$

Morton *et al.* [9] acknowledged the disproportionate influence longer sessions at a low heart rate could have on resulting sessions $w(t)$, in arbitrary units (AU), so introduced a weighting factor Y . They determined $Y = e^{bx}$, where b is a preselected value determined for men and women, and $x = \Delta HR \text{ ratio}$. So each training session is calculated as:

$$w(t) = D(\Delta HR \text{ ratio})Y \quad (2.3)$$

Next, Fitness, $g(t)$, and Fatigue, $h(t)$, can be calculated. These two functions are defined as

$$g(t) = g(t-1)e^{\frac{-i}{\tau_1}} + w(t) \quad (2.4)$$

and

$$h(t) = h(t-1)e^{\frac{-i}{\tau_2}} + w(t) \quad (2.5)$$

where each function states the fitness and fatigue at the end of day t , i is the number of days between the training session being added and the last training session, and τ_1 and τ_2 are the decay time constants for fitness and fatigue respectively. Finally, to model performance, two more constants are introduced: k_1 and k_2 . These constants are weighting factors for fitness and fatigue, they have "no direct physiological interpretation" [9] but can be used to adjust the model for athletes who recover more quickly to heavy training load or require more time to recover from sessions across a taper period. Finally, performance is modelled as

$$p(t) = k_1g(t) - k_2h(t) \quad (2.6)$$

The model was originally developed using a swimmer and when evaluated, using linear regression for the athlete specific parameters (τ_1 , τ_2 , k_1 , and k_2). The model reasonably estimates the performance outcome of a swimmer with relative confidence, no r^2 value is provided for the swimming application in 1975 [1]. In the 1990 model from Morton *et al.* [9], two of the researchers participated in a training and testing regimen and a "good degree of fit" was observed, with r^2 values of 0.71 and 0.96 calculated for a slightly trained and an untrained runner respectively, with the former being given a starting 1000 AU points given some pre-existing running fitness. The impulse-response model, if trained appropriately can help guide the structure and timing of a taper period for each athlete [9]. Versions of this kind of Impulse-Response model are used in basic analysis across many consumer fitness apps today, such as Training Peaks and Strava.

Limitations to the Impulse-Response Model

The Banister FFM simplifies human physiology quite effectively. However, in order to tune the parameters of the model, regular "criterion" testing needs to occur. These are maximal efforts done at regular intervals. During the testing of the Morton *et al.* [9] model refinements, both participants tested roughly once per week, something which is not feasible for high-level endurance athletes. This dependency on regular testing to tune parameters makes it particularly difficult to implement effectively as a wholistic model for rowers. Furthermore, the use of data to adjust model parameters results in less data being available to model training behaviour.

The model only considers training load, specifically endurance training load, as an indicator for performance. Given the power dominance in rowing, a training block with seemingly less

training load, as a result of more time and energy spent doing strength training, this can result in a bigger performance when tested next which may not be explained by changes in endurance training load. Additionally, the model does not consider other factors such as recovery and non-training strain an athlete might experience.

Finally, if the model is not constantly updated, it cannot predict future performances effectively. For the purposes of this project, a model would require an understanding of training behaviours in order to model the physiological response, predicting future performances. By modelling physiological response, suggestions can be made to improve certain energy system efficiency or effectiveness if needed.

Impulse-Response models are still widely used to model training and adaptation for athletes. Their simplicity being a key factor for this. To more effectively model performances, without the need for constant testing, alternative models can and should be used. Impulse-Response models can still be used in machine learning solutions, though, but as part of a larger ensemble approach [16].

2.2.3 Alternative Models

There are models which employ more complicated mathematical approaches. This section will explore one of the most developed models, Performance Potential (PerPot), and explore Artificial Neural Network approaches used by a number of researchers in recent years.

Performance Potential (PerPot)

Artificial Neural Network (ANN) approaches

Chapter 3

Data Collection and Management

Discussing approaches to data collection, key decisions in generating the model, eg. building a model which can be run with minimal effort by the users.

Chapter 4

Data Analysis and Visualization

Basic data analysis done, visualizations generated, comment on feedback given to users.

Chapter 5

Machine Learning Applications

What was the target, what did I actually make, maybe some results

5.1 Model Considerations

What led to the final model, difficulties, and data Considerations

5.2 The Implementation

Details about the model, including some examples of outputs (on my own data?)

Chapter 6

Discussion

Evaluation of methods and model, and a discussion of potential further step.

Chapter 7

Conclusion

Did we do what we set out to do.

Bibliography

- [1] E. W. Banister, T. W. Calvert, M. V. Savage, and T. Bach, "A systems model of the effects of training on physical performance," *IEEE Transactions on Systems, Man, and Cybernetics*, vol. SMC-6, no. 2, pp. 94–102, Feb. 1976. DOI: 10.1109/tsmc.1976.5409179.
- [2] J. Mäestu, J. Jürimäe, and T. Jürimäe, "Monitoring of performance and training in rowing," *Sports Medicine*, vol. 35, no. 7, pp. 597–617, Jul. 2005. DOI: 10.2165/00007256-200535070-00005.
- [3] K. S. Seiler and G. Ø. Kjerland, "Quantifying training intensity distribution in elite endurance athletes: is there evidence for an "optimal" distribution?" *Scandinavian Journal of Medicine & Science in Sports*, vol. 16, no. 1, pp. 49–56, 2006, ISSN: 0905-7188. DOI: 10.1111/j.1600-0838.2004.00418.x.
- [4] M. A. Rosenblat, A. S. Perrotta, and B. Vicenzino, "Polarized vs. threshold training intensity distribution on endurance sport performance: A systematic review and meta-analysis of randomized controlled trials," *Journal of Strength and Conditioning Research*, vol. 33, no. 12, pp. 3491–3500, Dec. 2019. DOI: 10.1519/jsc.0000000000002618.
- [5] A. Das, U. S. Kaniganti, S. J. Shenoy, P. Majumdar, and A. K. Syamal, "Monitoring training load, muscle damage, and body composition changes of elite indian rowers during a periodized training program," *Journal of Science in Sport and Exercise*, vol. 5, no. 4, pp. 348–359, Nov. 2022. DOI: 10.1007/s42978-022-00197-7.
- [6] S. Seiler and G. Ø. Kjerland, "Quantifying training intensity distribution in elite endurance athletes: Is there evidence for an "optimal" distribution?" *Scandinavian Journal of Medicine and Science in Sports*, vol. 16, no. 1, pp. 49–56, Oct. 2004. DOI: 10.1111/j.1600-0838.2004.00418.x.
- [7] S. Selles-Perez, J. Fernández-Sáez, and R. Cejuela, "Polarized and pyramidal training intensity distribution: Relationship with a half-ironman distance triathlon competition," *Journal of Sports Medicine and Science*, vol. 18, no. 4, pp. 708–715, Nov. 2019.
- [8] A. S. Göktepe, "Energy systems in sport," in *Amputee Sports for Victims of Terrorism*. IOS Press, 2007, pp. 24–31.

- [9] R. H. Morton, J. R. Fitz-Clarke, and E. W. Banister, "Modeling human performance in running," *Journal of Applied Physiology*, vol. 69, no. 3, pp. 1171–1177, Sep. 1990. DOI: 10.1152/jappl.1990.69.3.1171.
- [10] J. Edelmann-nusser, A. Hohmann, and B. Henneberg, "Modeling and prediction of competitive performance in swimming upon neural networks," *European Journal of Sport Science*, vol. 2, no. 2, pp. 1–10, Apr. 2002. DOI: 10.1080/17461390200072201.
- [11] N. Williams, "The borg rating of perceived exertion (rpe) scale," *Occupational Medicine*, vol. 67, no. 5, pp. 404–405, Jul. 2017. DOI: 10.1093/occmed/kqx063.
- [12] M. Kent, *Trimp method*, 2007. DOI: 10.1093/acref/9780198568506.013.7418. [Online]. Available: <https://www.oxfordreference.com/view/10.1093/acref/9780198568506.001.0001/acref-9780198568506-e-7418>.
- [13] R. White, *Acute:chronic workload ratio*, Oct. 2023. [Online]. Available: <https://www.scienceforsport.com/acutechronic-workload-ratio/>.
- [14] F. M. Impellizzeri, M. S. Tenan, T. Kempton, A. Novak, and A. J. Coutts, "Acute:chronic workload ratio: Conceptual issues and fundamental pitfalls," *International Journal of Sports Physiology and Performance*, vol. 15, no. 6, pp. 907–913, Jul. 2020. DOI: 10.1123/ijsp.2019-0864.
- [15] H. Zouhal, D. Boulosa, R. Ramirez-Campillo, A. Ali, and U. Granacher, "Editorial: Acute: Chronic workload ratio: Is there scientific evidence?" *Frontiers in Physiology*, vol. 12, May 2021. DOI: 10.3389/fphys.2021.669687.
- [16] F. Imbach, N. Sutton-Charani, J. Montmain, R. Candau, and S. Perrey, "The use of fitness-fatigue models for sport performance modelling: Conceptual issues and contributions from machine-learning," *Sports Medicine - Open*, vol. 8, no. 1, Mar. 2022. DOI: 10.1186/s40798-022-00426-x.

Appendix A1

Appendix

You may use appendices to include relevant background information, such as calibration certificates, derivations of key equations or presentation of a particular data reduction method. You should not use the appendices to dump large amounts of additional results or data which are not properly discussed. If these results are really relevant, then they should appear in the main body of the report.

A1.1 Appendix numbering

Appendices are numbered sequentially, A1, A2, A3. . . The sections, figures and tables within appendices are numbered in the same way as in the main text. For example, the first figure in Appendix A1 would be Figure A1.1. Equations continue the numbering from the main text.

A1.2 Lactate Test Results

Table A1.1: Test 1 Interval Details

Stage	Target Power (Watts)	Target Time	Relative VO2 (ml/kg/min)	HR (BPM)	RER	RPE	Actual Power (Watts)	Actual Time	Lactate (mmol/L)	Stroke Rate	Distance (m)	Notes
1	140	02:15.9	38.25	132	0.87	1	148	02:13.3	1	17	900	
2	175	02:06.2	39.88	141	0.87	2	178	02:05.3	1.3	19	957	
3	210	01:58.7	56.64	157	0.93	4	212	01:58.1	2.2	19	1016	
4	245	01:52.7	58.86	165	0.94	5	250	01:51.9	3.4	21	1072	
5	280	01:47.8	64.3	175	0.99	6	287	01:46.8	6.4	24	1123	
6	315	01:43.7	68.54	183	1.03	7	323	01:42.7	9.7	29	1168	
7	Max	Max	76.47	189	1.01	9	354	01:39.6	14.8	57	1204	
									16.3			1 minute post
									17.4			2 minutes post

Table A1.2: Test 2 Interval Details

Stage	Target Power (Watts)	Target Time	Relative VO2 (ml/kg/min)	HR (BPM)	RER	RPE	Actual Power (Watts)	Actual Time	Lactate (mmol/L)	Stroke Rate	Distance (m)	Notes
1	140	02:15.9	35.1	130	0.82	0	143	02:14.8	0.8	16	890	
2	175	02:06.2	41.9	142	0.81	1	178	02:05.2	1	17	958	
3	210	01:58.7	47.8	153	0.88	3	212	01:58.2	1.3	20	1015	
4	245	01:52.7	54.7	162	0.88	4	247	01:52.2	2	22	1069	
5	280	01:47.8	59.8	172	0.92	5	285	01:47.1	3.7	24	1120	
6	315	01:43.7	64.2	178	0.99	6	318	01:43.2	6.2	28	1162	
7	Max	Max	74.4	192	1.1	10	398	01:35.7	13.4	43	1253	
									13.3			2 mins post