

Estimating Power without Measuring it: a Machine Learning Approach

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Introduction

Context

Saris PowerTap PowerCal

- ▶ Low-cost power-meter based on heart-rate
- ✓ Low-cost device
- ✗ Not suitable to track small power changes

Strava

- ▶ Large amount of data
- ▶ Mathematical model based on mechanic

Problematic

Use machine learning to find more complex models using heterogeneous data sensor

Introduction

State-of-the-art

Mathematical model based on mechanics

$$P_{meca} = (0.5\rho S C_x V_a^2 + C_r m g \cos \alpha + m g \sin \alpha) V_d . \quad (1)$$

Model parameters

Name	Symbol	Unit
Air density	ρ	kg m^{-3}
Frontal surface	S	m^2
Drag coefficient	C_x	NA
Air speed	V_a	m s^{-1}
Rolling coefficient	C_r	NA
Mass of rider and bike	m	kg
Gravitational constant	g	m s^{-2}
Slope	α	rad
Rider speed	V_d	m s^{-1}

Power estimation using machine learning

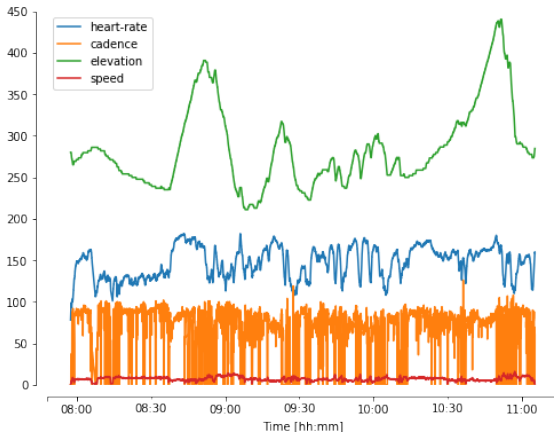


Figure: Original data

Power estimation using machine learning

Feature engineering

- ▶ Compute gradient for some data (acceleration, elevation, heart-rate)
- ▶ Compute derivative with different time periods (1 s — 5 s)
- ▶ Total of 48 features

Regressor

- ▶ Gradient boosting machine

Experiments Setup

Data set

- ▶ 5 riders
- ▶ 4 power meters: Saris PowerTap, Rotor Power LT, Power2Max, and SRM
- ▶ 417 rides

Model validation

- ▶ Group k -fold cross-validation with $k = 3$

Model evaluation

- ▶ Coefficient of determination R^2
- ▶ Median absolute error (MAE)

Experiments

Model hyper-parameters

Mathematical model

Parameter	Value
Rider weight	Specific
Bike weight	6.8 kg
Rolling coefficient C_r	0.0045
Atmospheric pressure	1013 hPa
SC_x	0.32 m ²
Temperature	15 °C

Machine learning model

Parameter	Value
Number of decision tree	200
Depth of each decision tree	8

Results

Quantitative results

R^2 and MAE scores

Metric	Math	ML
R^2	-0.55	0.76
MAE	61.09	21.95

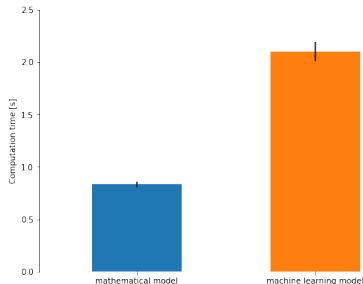


Figure: Computation time for estimation around 1 million samples

Results

Zoom-in

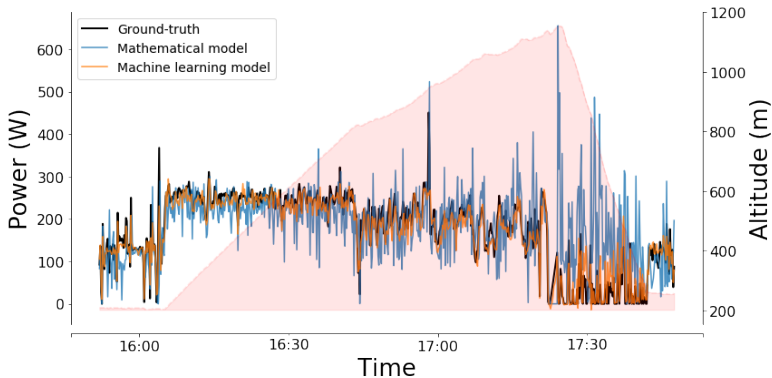


Figure: Power estimation for uphill and downhill

Results

Zoom-in

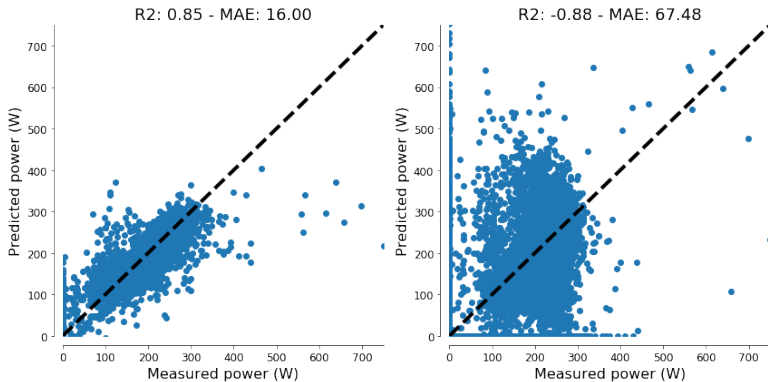


Figure: Left: Machine learning model — right: mathematical model

Results

Zoom-in

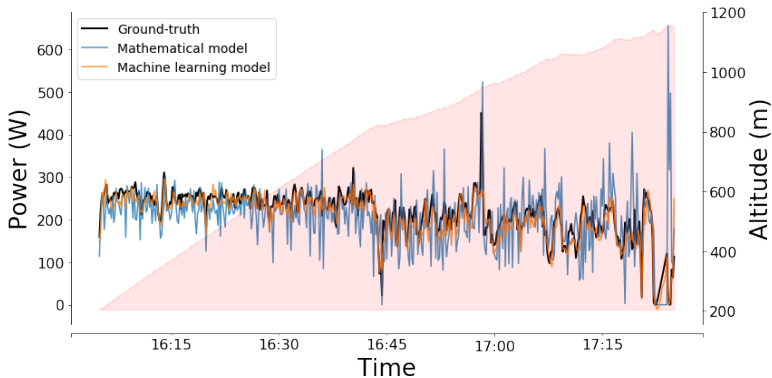


Figure: Power estimation for uphill

Results

Zoom-in

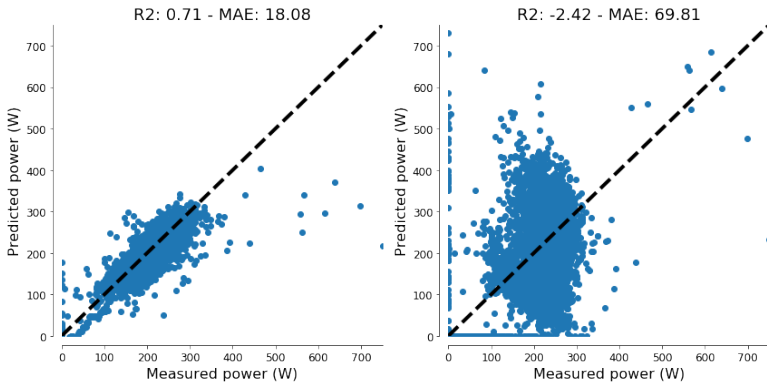


Figure: Left: Machine learning model — right: mathematical model

Results Summary

Mathematical model

- ✓ Fast prediction
- ✗ Too much unknown parameters
- ✗ Too much variation in the estimation

Machine learning model

- ✓ Fast prediction
- ✓ Better prediction
- ✗ Difficulty to predict short power peak

Results

Analysis of the machine learning model

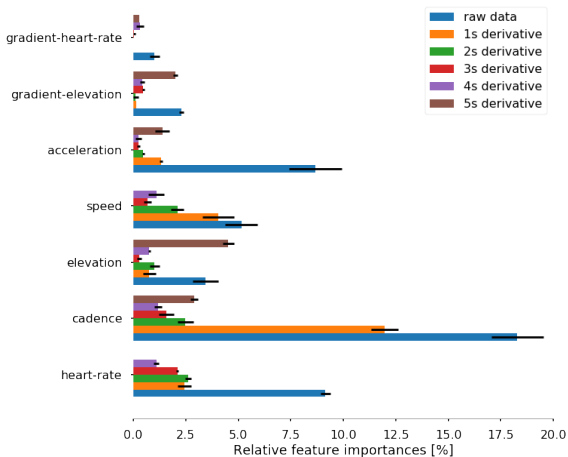


Figure: Feature importances of the different features used in the model

Future work

Extension of the current work

- ✓ Convolutional neural-network
- ✗ Larger data set

Open-source initiative via GoldenCheetah

- ▶ OpenData collection
- ▶ Development of `scikit-sports`

Purpose

- ✓ Reproducibility of analysis and methods
- ✓ Use data science tools to solve different problematic in cycling performance