# Package 'shorts'

May 2, 2024

Type Package

```
Title Short Sprints
Version 3.1.1
Description Create short sprint acceleration-velocity (AVP) and force-velocity (FVP) profiles
      and predict kinematic and kinetic variables using the timing-gate split times, laser or
      radar gun data, tether devices data, as well as the data provided by the GPS and LPS
      monitoring systems. The modeling method utilized in this package is based on the works of
     Furusawa K, Hill AV, Parkinson JL (1927) <doi:10.1098/rspb.1927.0035>,
     Greene PR. (1986) <doi:10.1016/0025-5564(86)90063-5>,
     Chelly SM, Denis C. (2001) <doi:10.1097/00005768-200102000-00024>,
     Clark KP, Rieger RH, Bruno RF, Stearne DJ. (2017) <doi:10.1519/JSC.000000000000002081>,
     Samozino P. (2018) <doi:10.1007/978-3-319-05633-3_11>,
     Samozino P. and Peyrot N., et al (2022) <doi:10.1111/sms.14097>,
     Clavel, P., et al (2023) <doi:10.1016/j.jbiomech.2023.111602>,
     Jovanovic M. (2023) <doi:10.1080/10255842.2023.2170713>, and
     Jovanovic M., et al (2024) <doi:10.3390/s24092894>.
URL https://mladenjovanovic.github.io/shorts/
BugReports https://github.com/mladenjovanovic/shorts/issues
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Author Mladen Jovanović [aut, cre, cph]
     (<https://orcid.org/0000-0002-4013-6530>)
Maintainer Mladen Jovanović <coach.mladen.jovanovic@gmail.com>
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```

coef.shorts\_model

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# Description

S3 method for extracting model parameters from shorts\_model object

# Usage

```
## S3 method for class 'shorts_model'
coef(object, ...)
```

# Arguments

```
object shorts_model object
... Extra arguments. Not used
```

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#### **Examples**

```
split_distances <- c(10, 20, 30, 40, 50)
split_times <- create_timing_gates_splits(
    gates = split_distances,
    MSS = 10,
    MAC = 9,
    FD = 0.25,
    TC = 0
)

# Simple model
simple_model <- model_timing_gates(split_distances, split_times)
coef(simple_model)</pre>
```

confint.shorts\_model S3 method for providing confidence intervals for the shorts\_model

# Description

S3 method for providing confidence intervals for the shorts\_model

# Usage

```
## S3 method for class 'shorts_model'
confint(object, ...)
```

# **Arguments**

```
## Not run:
split_distances <- c(10, 20, 30, 40, 50)
split_times <- create_timing_gates_splits(
    gates = split_distances,
    MSS = 10,
    MAC = 9,
    FD = 0,
    TC = 0,
    noise = 0.01
)

# Simple model
simple_model <- model_timing_gates(split_distances, split_times)
confint(simple_model)

## End(Not run)</pre>
```

4 convert\_FVP

convert_FVP	Convert Force-Velocity profile back to Acceleration-Velocity profile

# Description

This function converts back the Force-Velocity profile (FVP; F0 and V0 parameters) to Acceleration-Velocity profile (AVP; MSS and MAC parameters)

# Usage

```
convert_FVP(
  F0,
  V0,
  bodymass = 75,
  inertia = 0,
  resistance = 0,
  wind_velocity = 0,
  ...
)
```

# Arguments

F0, V0	Numeric vectors. FV profile parameters
bodymass	Body mass in kg. Used to calculate relative power and forwarded to $\verb get_air_resistance  $
inertia	External inertia in kg (for example a weight vest, or a sled). Not included in the air resistance calculation
resistance	External horizontal resistance in Newtons (for example tether device or a sled friction resistance)
wind_velocity	In meters per second (m/s). Use negative number as head wind, and positive number as back wind. Default is 0m/s (no wind)
	Forwarded to predict_power_at_distance

# Value

A list with calculated MSS and MAC parameters

```
FVP <- create_FVP(7, 8.3, inertia = 10, resistance = 50)
convert_FVP(FVP$F0, FVP$V0, inertia = 10, resistance = 50)</pre>
```

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|--|

# **Description**

Creates Force-Velocity Profile (FVP) modified using ideas by Pierre Samozino and JB-Morin, et al. (2016) and Pierre Samozino and Nicolas Peyror, et al (2021).

# Usage

```
create_FVP(
   MSS,
   MAC,
   bodymass = 75,
   inertia = 0,
   resistance = 0,
   wind_velocity = 0,
   ...
)
```

# Arguments

MSS, MAC	Numeric vectors. Model parameters
bodymass	Body mass in kg. Used to calculate relative power and forwarded to $\texttt{get\_air\_resistance}$
inertia	External inertia in kg (for example a weight vest, or a sled). Not included in the air resistance calculation
resistance	External horizontal resistance in Newtons (for example tether device or a sled friction resistance)
wind_velocity	In meters per second (m/s). Use negative number as head wind, and positive number as back wind. Default is 0m/s (no wind)
	Forwarded to predict_power_at_distance

# Value

List containing the following elements:

```
bodymass Returned bodymass used in FV profiling
F0 Horizontal force when velocity=0
F0_rel F0 divided by bodymass
V0 Velocity when horizontal force=0
Pmax Maximal horizontal power
Pmax_rel Pmax divided by bodymass
FV_slope Slope of the FV profile. See References for more info
```

6 create\_sprint\_trace

#### References

Samozino P, Rabita G, Dorel S, Slawinski J, Peyrot N, Saez de Villarreal E, Morin J-B. 2016. A simple method for measuring power, force, velocity properties, and mechanical effectiveness in sprint running: Simple method to compute sprint mechanics. Scandinavian Journal of Medicine & Science in Sports 26:648–658. DOI: 10.1111/sms.12490.

Samozino P, Peyrot N, Edouard P, Nagahara R, Jimenez-Reyes P, Vanwanseele B, Morin J. 2022. Optimal mechanical force-velocity profile for sprint acceleration performance. Scandinavian Journal of Medicine & Science in Sports 32:559–575. DOI: 10.1111/sms.14097.

# **Examples**

```
data("jb_morin")
m1 <- model_radar_gun(time = jb_morin$time, velocity = jb_morin$velocity)

fv_profile <- create_FVP(
   MSS = m1$parameters$MSS,
   MAC = m1$parameters$MAC,
   bodyheight = 1.72,
   bodymass = 120,
)</pre>
fv_profile
```

# Description

This function creates sprint trace either using time or distance vectors

```
create_sprint_trace(
  MSS,
  MAC,
  time = NULL,
  distance = NULL,
  TC = 0,
  DC = 0,
  FD = 0,
  remove_leading = FALSE
)
```

#### **Arguments**

MSS, MAC	Numeric vector. Model parameters
time	Numeric vector.
distance	Numeric vector.
TC	Numeric vector. Time-shift added to sprint times. Default is 0
DC	Numeric vector. Distance-shift added to sprint distance. Default is $\boldsymbol{0}$
FD	Numeric vector. Flying start distance. Default is 0
remove_leading	Should trace leading to sprint be removed? Default is FALSE

# Value

Data-frame with following 6 columns

```
time Measurement-scale time vector in seconds. Equal to parameter time
distance Measurement-scale distance vector in meters. Equal to parameter distance
velocity Velocity vector in m/s
acceleration Acceleration vector in m/s/s
sprint_time Sprint scale time vector in seconds. Sprint always start at t=0s
sprint_distance Sprint scale distance vector in meters. Sprint always start at d=0m
```

#### See Also

```
create_timing_gates_splits
```

# **Examples**

```
df <- create_sprint_trace(8, 7, time = seq(0, 6, by = 0.01))
df <- create_sprint_trace(8, 7, distance = seq(0, 40, by = 1))
```

# **Description**

This function is used to generate timing gates splits with predetermined parameters

8 dynaspeed

#### Usage

```
create_timing_gates_splits(
   MSS,
   MAC,
   gates = c(5, 10, 20, 30, 40),
   FD = 0,
   TC = 0,
   noise = 0
)
```

# Arguments

MSS, MAC Numeric vectors. Model parameters
gates Numeric vectors. Distances of the timing gates
FD Numeric vector. Flying start distance. Default is 0
TC Numeric vector. Time-correction added to split times (e.g., reaction time). Default is 0
noise Numeric vector. SD of Gaussian noise added to the split times. Default is 0

#### See Also

```
create_sprint_trace
```

# **Examples**

```
create_timing_gates_splits(
  gates = c(10, 20, 30, 40, 50),
  MSS = 10,
  MAC = 9,
  FD = 0.5,
  TC = 0
)
```

dynaspeed

DynaSpeed Single Sprint Data

# **Description**

DynaSpeed(TM) data collected for a single athlete (female, 177cm, 64kg) and a single sprint over 40m. Sampling frequency is 1,000Hz. Additional time and distance shift is added to the dataset to provide a sandbox for potential issues during the analysis

```
data(dynaspeed)
```

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#### **Format**

Data frame with 4 variables and 7,251 observations:

time time in seconds

distance Distance in meters

velocity Smoothed velocity in meters per second

raw\_velocity Velocity in meters per second

#### Author(s)

Håkan Andersson The High-Performance Center Växjö, Sweden <hakan.andersson@hpcsweden.com>

find\_functions

Find functions

#### **Description**

Family of functions that serve a purpose of finding maximal value and critical distances and times at which power, acceleration or velocity drops below certain threshold.

find\_peak\_power\_distance finds peak power and distance at which peak power occurs

find\_peak\_power\_time finds peak power and time at which peak power occurs

find\_velocity\_critical\_distance finds critical distance at which percent of MSS is achieved

find\_velocity\_critical\_time finds critical time at which percent of MSS is achieved

find\_acceleration\_critical\_distance finds critical distance at which percent of MAC is reached

find\_acceleration\_critical\_time finds critical time at which percent of MAC is reached

find\_power\_critical\_distance finds critical distances at which peak power over percent is achieved

find\_power\_critical\_time finds critical times at which peak power over percent is achieved

```
find_peak_power_distance(MSS, MAC, inertia = 0, resistance = 0, ...)
find_peak_power_time(MSS, MAC, inertia = 0, resistance = 0, ...)
find_velocity_critical_distance(MSS, MAC, percent = 0.9)
find_velocity_critical_time(MSS, MAC, percent = 0.9)
find_acceleration_critical_distance(MSS, MAC, percent = 0.9)
```

find\_functions

```
find_acceleration_critical_time(MSS, MAC, percent = 0.9)

find_power_critical_distance(
   MSS,
   MAC,
   inertia = 0,
   resistance = 0,
   percent = 0.9,
   ...
)

find_power_critical_time(
   MSS,
   MAC,
   inertia = 0,
   resistance = 0,
   percent = 0.9,
   ...
)
```

# Arguments

MSS, MAC	Numeric vectors. Model parameters
inertia	External inertia in kg (for example a weight vest, or a sled). Not included in the air resistance calculation
resistance	External horizontal resistance in Newtons (for example tether device or a sled friction resistance)
	Arguments passed on to get_air_resistance
	velocity Instantaneous running velocity in meters per second (m/s)
	bodymass In kilograms (kg). Default is 75kg
	bodyheight In meters (m). Default is 1.75m
	barometric_pressure In Torrs. Default is 760Torrs
	air_temperature In Celzius (C). Default is 25C
	wind_velocity In meters per second (m/s). Use negative number as head wind, and positive number as back wind. Default is 0m/s (no wind)
percent	Numeric vector. Used to calculate critical distance. Default is 0.9

#### Value

find\_peak\_power\_distance returns list with two elements: peak\_power and distance at which peak power occurs

find\_peak\_power\_time returns list with two elements: peak\_power and time at which peak power occurs

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#### References

Haugen TA, Tønnessen E, Seiler SK. 2012. The Difference Is in the Start: Impact of Timing and Start Procedure on Sprint Running Performance: Journal of Strength and Conditioning Research 26:473–479. DOI: 10.1519/JSC.0b013e318226030b.

Samozino P. 2018. A Simple Method for Measuring Force, Velocity and Power Capabilities and Mechanical Effectiveness During Sprint Running. In: Morin J-B, Samozino P eds. Biomechanics of Training and Testing. Cham: Springer International Publishing, 237–267. DOI: 10.1007/978-3-319-05633-3 11.

```
dist \leftarrow seq(0, 40, length.out = 1000)
velocity <- predict_velocity_at_distance(</pre>
  distance = dist,
 MSS = 10,
  MAC = 9
)
acceleration <- predict_acceleration_at_distance(</pre>
  distance = dist,
  MSS = 10,
  MAC = 9
)
# Use ... to forward parameters to the shorts::get_air_resistance
pwr <- predict_power_at_distance(</pre>
  distance = dist,
 MSS = 10,
 MAC = 9
  # bodyweight = 100,
  # bodyheight = 1.9,
  # barometric_pressure = 760,
  # air_temperature = 25,
  # wind_velocity = 0
)
# Find critical distance when 90% of MSS is reached
plot(x = dist, y = velocity, type = "l")
abline(h = 10 * 0.9, col = "gray")
abline(v = find\_velocity\_critical\_distance(MSS = 10, MAC = 9), col = "red")
# Find critical distance when 20% of MAC is reached
plot(x = dist, y = acceleration, type = "l")
abline(h = 9 * 0.2, col = "gray")
abline(v = find_acceleration_critical_distance(MSS = 10, MAC = 9, percent = 0.2), col = "red")
# Find peak power and location of peak power
plot(x = dist, y = pwr, type = "l")
peak_pwr <- find_peak_power_distance(</pre>
```

```
MSS = 10,
MAC = 9
# Use ... to forward parameters to the shorts::get_air_resistance
)
abline(h = peak_pwr$peak_power, col = "gray")
abline(v = peak_pwr$distance, col = "red")

# Find distance in which relative power stays over 75% of PMAX'
plot(x = dist, y = pwr, type = "l")
abline(h = peak_pwr$peak_power * 0.75, col = "gray")
pwr_zone <- find_power_critical_distance(MSS = 10, MAC = 9, percent = 0.75)
abline(v = pwr_zone$lower, col = "blue")
abline(v = pwr_zone$upper, col = "blue")</pre>
```

find\_optimal\_distance Function that finds the distance at which the sprint, probe, or FV profile is optimal (i.e., equal to 100 perc)

# Description

Function that finds the distance at which the sprint, probe, or FV profile is optimal (i.e., equal to 100 perc)

# Usage

```
find_optimal_distance(..., optimal_func = optimal_FV, min = 1, max = 60)
```

#### **Arguments**

```
... Forwarded to selected optimal_func

optimal_func Selected profile optimization function. Default is optimal_FV

min, max Distance over which to find optimal profile distance
```

#### Value

Distance

```
MSS <- 10
MAC <- 8
bodymass <- 75

fv <- create_FVP(MSS, MAC, bodymass)

find_optimal_distance(
  F0 = fv$F0,
    V0 = fv$V0,</pre>
```

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```
bodymass = fv$bodymass,
  optimal_func = optimal_FV,
  method = "max"
)

find_optimal_distance(
  MSS = MSS,
  MAC = MAC,
  optimal_func = optimal_MSS_MAC
)

find_optimal_distance(
  MSS = MSS,
  MAC = MAC,
  optimal_func = probe_MSS_MAC
)
```

fitted.shorts\_model

S3 method for returning predictions of shorts\_model

### **Description**

S3 method for returning predictions of shorts\_model

# Usage

```
## S3 method for class 'shorts_model'
fitted(object, ...)
```

# **Arguments**

```
object shorts_model object
... Extra arguments. Not used
```

```
split_distances <- c(10, 20, 30, 40, 50)
split_times <- create_timing_gates_splits(
   gates = split_distances,
   MSS = 10,
   MAC = 9,
   FD = 0.25,
   TC = 0
)

# Simple model
simple_model <- model_timing_gates(split_distances, split_times)
fitted(simple_model)</pre>
```

14 format\_splits

format\_splits

Format Split Data

# **Description**

Function formats split data and calculates split distances, split times and average split velocity

# Usage

```
format_splits(distance, time)
```

# **Arguments**

distance Numeric vector time Numeric vector

#### Value

Data frame with the following columns:

```
split Split number
split_distance_start Distance at which split starts
split_distance_stop Distance at which split ends
split_distance Split distance
split_time_start Time at which distance starts
split_time_stop Time at which distance ends
split_time Split time
split_mean_velocity Mean velocity over split distance
split_mean_acceleration Mean acceleration over split distance
```

```
data("split_times")
john_data <- split_times[split_times$athlete == "John", ]
format_splits(john_data$distance, john_data$time)</pre>
```

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get\_air\_resistance

Get Air Resistance

# **Description**

get\_air\_resistance estimates air resistance in Newtons

# Usage

```
get_air_resistance(
  velocity,
  bodymass = 75,
  bodyheight = 1.75,
  barometric_pressure = 760,
  air_temperature = 25,
  wind_velocity = 0
)
```

# Arguments

velocity Instantaneous running velocity in meters per second (m/s)

bodymass In kilograms (kg). Default is 75kg bodyheight In meters (m). Default is 1.75m

barometric\_pressure

In Torrs. Default is 760Torrs

air\_temperature

In Celzius (C). Default is 25C

wind\_velocity In meters per second (m/s). Use negative number as head wind, and positive

number as back wind. Default is 0m/s (no wind)

#### Value

Air resistance in Newtons (N)

#### References

Arsac LM, Locatelli E. 2002. Modeling the energetics of 100-m running by using speed curves of world champions. Journal of Applied Physiology 92:1781–1788. DOI: 10.1152/japplphysiol.00754.2001.

Samozino P, Rabita G, Dorel S, Slawinski J, Peyrot N, Saez de Villarreal E, Morin J-B. 2016. A simple method for measuring power, force, velocity properties, and mechanical effectiveness in sprint running: Simple method to compute sprint mechanics. Scandinavian Journal of Medicine & Science in Sports 26:648–658. DOI: 10.1111/sms.12490.

van Ingen Schenau GJ, Jacobs R, de Koning JJ. 1991. Can cycle power predict sprint running performance? European Journal of Applied Physiology and Occupational Physiology 63:255–260. DOI: 10.1007/BF00233857.

jb\_morin

#### **Examples**

```
get_air_resistance(
  velocity = 5,
  bodymass = 80,
  bodyheight = 1.90,
  barometric_pressure = 760,
  air_temperature = 16,
  wind_velocity = -0.5
)
```

jb\_morin

JB Morin Sample Dataset

# **Description**

Sample radar gun data provided by Jean-Benoît Morin on his website. See https://jbmorin.net/2017/12/13/a-spreadsheet-for-sprint-acceleration-force-velocity-power-profiling/for more details.

# Usage

```
data(jb_morin)
```

#### Format

Data frame with 2 variables and 232 observations:

```
time Time in secondsvelocity Velocity in m/s
```

#### **Details**

This dataset represents a sample data provided by Jean-Benoît Morin on a single individual running approximately 35m from a stand still position that is measured with the radar gun. Individual's body mass is 75kg, height is 1.72m. Conditions of the run are the following: air temperature 25C, barometric pressure 760mmHg, wind velocity 0m/s.

The purpose of including this dataset in the package is to check the agreement of the model estimates with Jean-Benoît Morin Microsoft Excel spreadsheet.

# Author(s)

```
Jean-Benoît Morin
Inter-university Laboratory of Human Movement Biology
Saint-Étienne, France https://jbmorin.net/
```

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#### References

Morin JB. 2017.A spreadsheet for Sprint acceleration Force-Velocity-Power profiling. Available at https://jbmorin.net/2017/12/13/a-spreadsheet-for-sprint-acceleration-force-velocity-power-profiling/(accessed October 27, 2020).

laser\_gun\_data

Laser Gun Data

### **Description**

Performance of 35m sprint by a youth basketball player done using standing start. Sample was collected by laser gun (CMP3 Distance Sensor, Noptel Oy, Oulu, Finland) and was sampled at a rate of 2.56 KHz. A polynomial function modeling the relationship between distance and time was employed and subsequently resampled at a frequency of 1,000 Hz using Musclelab<sup>TM</sup> v10.232.107.5298, a software developed by Ergotest Technology AS located in Langesund, Norway. Data was further modified by calculating raw acceleration using dv/dt (using smoothed velocity provided by the system), and then smoothed out using 4th-order Butterworth filter with a cutoff frequency of 1 Hz.

#### Usage

data(laser\_gun\_data)

#### Format

Data frame with 6 variables and 4805 observations:

time Time vector in seconds

distance Distance vector in meters

velocity Smoothed velocity vector in m/s; this represent step-averaged velocity

raw\_velocity Raw velocity vector in m/s

**raw\_acceleration** Raw acceleration vector in m/s/s; calculated using difference in smoothed velocity divided by time difference (i.e., dv/dt method of derivation)

butter\_acceleration Smoothed acceleration vector in m/s/s; smoothed out using 4th-order Butter-worth filter with a cutoff frequency of 1 Hz

LPS\_session

LPS Basketball Session Dataset

#### Description

LPS Basketball Session Dataset

# Usage

data(LPS\_session)

#### **Format**

Data frame with 5 variables and 91,099 observations:

time Time in seconds from the start of the session

x x-coordinate in meters provided by the LPS

y y-coordinate in meters provided by the LPS

velocity Velocity provided by LPS in m/s

acceleration Acceleration provided by LPS in m/s

#### **Details**

This dataset represents a sample data provided by Local Positioning System (LPS) on a single individual performing a single basketball practice session (aprox. 90min). Sampling frequency is 20Hz.

model\_functions

Model functions

#### **Description**

Family of functions that serve a purpose of estimating short sprint parameters

model\_in\_situ estimates short sprint parameters using velocity-acceleration trace, provided by the monitoring systems like GPS or LPS. See references for the information

model\_radar\_gun estimates short sprint parameters using time-velocity trace, with additional parameter TC serving as intercept

model\_laser\_gun alias for model\_radar\_gun

model\_tether estimates short sprint parameters using distance-velocity trace (e.g., tether devices).

model\_tether\_DC estimates short sprint parameters using distance-velocity trace (e.g., tether devices) with additional distance correction DC parameter

model\_time\_distance estimates short sprint parameters using time distance trace

model\_time\_distance\_FD estimates short sprint parameters using time-distance trace with additional flying distance correction parameter FD

model\_time\_distance estimates short sprint parameters using time distance trace with additional time correction parameter TC

model\_time\_distance estimates short sprint parameters using time distance trace with additional distance correction parameter DC

model\_time\_distance estimates short sprint parameters using time distance trace with additional time correction TC and distance correction TC parameters

model\_timing\_gates estimates short sprint parameters using distance-time trace (e.g., timing gates/photo cells)

model\_timing\_gates\_TC estimates short sprint parameters using distance-time trace (e.g., timing gates/photo cells), with additional time correction parameter TC

model\_timing\_gates\_FD estimates short sprint parameters using distance-time trace (e.g., timing gates/photo cells), with additional flying distance correction parameter FD

model\_timing\_gates\_DC estimates short sprint parameters using distance-time trace (e.g., timing gates/photo cells), with additional distance correction parameter DC

model\_timing\_gates\_TC\_DC estimates short sprint parameters using distance-time trace (e.g., timing gates/photo cells), with additional time correction TC and distance correction DC parameters

```
model_in_situ(
  velocity,
  acceleration,
  weights = 1,
  velocity_threshold = NULL,
  velocity_step = 0.2,
  n_{observations} = 2,
  CV = NULL,
  na.rm = FALSE.
)
model_radar_gun(
  time,
  velocity,
  weights = 1,
  CV = NULL,
  use_observed_MSS = FALSE,
  na.rm = FALSE,
)
model_laser_gun(
  time,
  velocity,
```

```
weights = 1,
 CV = NULL,
  use_observed_MSS = FALSE,
  na.rm = FALSE,
)
model_tether(
 distance,
 velocity,
 weights = 1,
 CV = NULL,
 use_observed_MSS = FALSE,
  na.rm = FALSE,
)
model_tether_DC(
 distance,
  velocity,
 weights = 1,
 CV = NULL,
  use_observed_MSS = FALSE,
 na.rm = FALSE,
)
model_time_distance(time, distance, weights = 1, CV = NULL, na.rm = FALSE, ...)
model_time_distance_FD(
  time,
 distance,
 weights = 1,
 FD = NULL,
 CV = NULL,
 na.rm = FALSE,
)
model_time_distance_TC(
  time,
 distance,
 weights = 1,
 CV = NULL,
 na.rm = FALSE,
)
```

```
model_time_distance_DC(
  time,
  distance,
 weights = 1,
 CV = NULL,
 na.rm = FALSE,
)
model_time_distance_TC_DC(
  time,
 distance,
 weights = 1,
 CV = NULL,
 na.rm = FALSE,
)
model_timing_gates(distance, time, weights = 1, CV = NULL, na.rm = FALSE, ...)
model_timing_gates_TC(
 distance,
 time,
 weights = 1,
 CV = NULL,
 na.rm = FALSE,
)
model_timing_gates_FD(
  distance,
  time,
 weights = 1,
 FD = NULL,
 CV = NULL,
 na.rm = FALSE,
)
model_timing_gates_DC(
  distance,
  time,
 weights = 1,
 CV = NULL,
 na.rm = FALSE,
)
```

```
model_timing_gates_TC_DC(
   distance,
   time,
   weights = 1,
   CV = NULL,
   na.rm = FALSE,
   ...
)
```

# **Arguments**

weights

velocity\_threshold

Velocity cutoff. If NULL (default), velocity of the observation with the fastest acceleration is taken as the cutoff value

velocity\_step Velocity increment size for finding max acceleration. Default is 0.2 m/s

n\_observations Number of top acceleration observations to keep in velocity bracket. Default is 2

CV Should cross-validation be used to estimate model fit? Default is NULL. Otherwise use integer indicating number of folds

na.rm Logical. Default is FALSE

Forwarded to n.l.s.l. M. function

... Forwarded to nlsLM function time, velocity, distance, acceleration Numeric vector use\_observed\_MSS

Should observed peak velocity be used as MSS parameter? Default is FALSE

FD Use this parameter if you do not want the FD parameter to be estimated, but

rather take the provided value

Numeric vector. Default is 1

#### Value

List object with the following elements:

data Data frame used to estimate the sprint parameters

model\_info Extra information regarding model used

model Model returned by the nlsLM function

parameters List with the following estimated parameters: MSS, MAC, TAU, and PMAX

correction List with additional model correcitons

predictions Data frame with .predictor, .observed, .predicted, and .residual columns

model\_fit List with multiple model fit estimators

CV If cross-validation is performed, this will included the data as above, but for each fold

#### References

Samozino P. 2018. A Simple Method for Measuring Force, Velocity and Power Capabilities and Mechanical Effectiveness During Sprint Running. In: Morin J-B, Samozino P eds. Biomechanics of Training and Testing. Cham: Springer International Publishing, 237–267. DOI: 10.1007/978-3-319-05633-3\_11.

Clavel, P., Leduc, C., Morin, J.-B., Buchheit, M., & Lacome, M. (2023). Reliability of individual acceleration-speed profile in-situ in elite youth soccer players. Journal of Biomechanics, 153, 111602. https://doi.org/10.1016/j.jbiomech.2023.111602

Morin, J.-B. (2021). The "in-situ" acceleration-speed profile for team sports: testing players without testing them. JB Morin, PhD – Sport Science website. Accessed 31. Dec. 2023. https://jbmorin.net/2021/07/29/the-in-situ-sprint-profile-for-team-sports-testing-players-without-testing-them/

```
# Model In-Situ (Embedded profiling)
data("LPS_session")
m1 <- model_in_situ(</pre>
  velocity = LPS_session$velocity,
  acceleration = LPS_session$acceleration,
  # Use specific cutoff value
  velocity_threshold = 4)
plot(m1)
# Model Radar Gun (includes Time Correction)
df <- create_sprint_trace(MSS = 8, MAC = 6, time = seq(0, 6, 0.1))</pre>
# Add some noise
df$velocity <- df$velocity + rnorm(n = nrow(df), 0, 10^-2)
m1 <- model_radar_gun(time = df$time, velocity = df$velocity)</pre>
plot(m1)
# Model Laser Gun (includes Time Correction)
df <- create_sprint_trace(MSS = 8, MAC = 6, time = seq(0, 6, 0.1))</pre>
# Add some noise
df$velocity <- df$velocity + rnorm(n = nrow(df), 0, 10^-2)</pre>
m1 <- model_laser_gun(time = df$time, velocity = df$velocity)</pre>
plot(m1)
# Model Tether
df \leftarrow create\_sprint\_trace(MSS = 8, MAC = 6, time = seq(0, 6, 0.5))
m1 <- model_tether(distance = df$distance, velocity = df$velocity)</pre>
```

```
plot(m1)
# Model Tether with Distance Correction (DC)
df \leftarrow create\_sprint\_trace(MSS = 8, MAC = 6, time = seq(0.001, 6, 0.5), DC = 5)
m1 <- model_tether_DC(distance = df$distance, velocity = df$velocity)</pre>
plot(m1)
# Model Time-Distance trace (simple, without corrections)
df <- create_sprint_trace(MSS = 8, MAC = 6, time = seq(0, 5, by = 0.5))
m1 <- model_time_distance(time = df$time, distance = df$distance)</pre>
m1
plot(m1)
# Model Time-Distance trace (with Flying Distance Correction)
df \leftarrow create\_sprint\_trace(MSS = 8, MAC = 6, time = seq(0, 5, by = 0.5), FD = 0.5)
m1 <- model_time_distance_FD(time = df$time, distance = df$distance)</pre>
m1
plot(m1)
# Model Time-Distance trace (with Time Correction)
df \leftarrow create\_sprint\_trace(MSS = 8, MAC = 6, time = seq(0, 5, by = 0.5), TC = 1.5)
m1 <- model_time_distance_TC(time = df$time, distance = df$distance)</pre>
m1
plot(m1)
# Model Time-Distance trace (with Distance Correction)
df \leftarrow create\_sprint\_trace(MSS = 8, MAC = 6, time = seq(0, 5, by = 0.5), DC = -5)
m1 <- model_time_distance_DC(time = df$time, distance = df$distance)</pre>
m1
plot(m1)
# Model Time-Distance trace (with Time and Distance Corrections)
df \leftarrow create\_sprint\_trace(MSS = 8, MAC = 6, time = seq(0, 5, by = 0.5), TC = -1.3, DC = 5)
m1 <- model_time_distance_TC_DC(time = df$time, distance = df$distance)</pre>
m1
plot(m1)
# Model Timing Gates (simple, without corrections)
df \leftarrow create\_sprint\_trace(MSS = 8, MAC = 6, distance = c(5, 10, 20, 30, 40))
m1 <- model_timing_gates(distance = df$distance, time = df$time)</pre>
m1
plot(m1)
```

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```
# Model Timing Gates (with Time Correction)
df <- create_sprint_trace(MSS = 8, MAC = 6, distance = c(5, 10, 20, 30, 40), TC = 0.2)
m1 <- model_timing_gates_TC(distance = df$distance, time = df$time)</pre>
plot(m1)
# Model Timing Gates (with Flying Distance Correction)
df <- create_sprint_trace(MSS = 8, MAC = 6, distance = c(5, 10, 20, 30, 40), FD = 0.5)
m1 <- model_timing_gates_FD(distance = df$distance, time = df$time)</pre>
plot(m1)
# Model Timing Gates (with Distance Correction)
df <- create_sprint_trace(MSS = 8, MAC = 6, distance = c(5, 10, 20, 30, 40), DC = 1.5)
m1 <- model_timing_gates_DC(distance = df$distance, time = df$time)</pre>
plot(m1)
# Model Timing Gates (with Time and Distance Corrections)
df \leftarrow create\_sprint\_trace(MSS = 8, MAC = 6, distance = c(5, 10, 20, 30, 40), TC = 0.25, DC = 1.5)
m1 <- model_timing_gates_TC_DC(distance = df$distance, time = df$time)</pre>
m1
plot(m1)
```

optimal\_functions

Optimal profile functions

#### Description

Family of functions that serve a purpose of finding optimal sprint or force-velocity profile optimal\_FV finds "optimal" F0 and V0 where time at distance is minimized, while keeping the power the same

optimal\_MSS\_MAC finds "optimal" MSS and MAS where time at distance is minimized, while keeping the Pmax the same

```
optimal_FV(
  distance,
  F0,
  V0,
  bodymass = 75,
  inertia = 0,
  resistance = 0,
  method = "max",
```

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```
optimal_MSS_MAC(distance, MSS, MAC)
```

#### **Arguments**

distance Numeric vector

F0, V0 Numeric vectors. FV profile parameters

bodymass in kg

inertia External inertia in kg (for example a weight vest, or a sled). Not included in the

air resistance calculation

resistance External horizontal resistance in Newtons (for example tether device or a sled

friction resistance)

method Method to be utilized. Options are "peak" and "max" (default)

... Arguments passed on to get\_air\_resistance

velocity Instantaneous running velocity in meters per second (m/s)

bodyheight In meters (m). Default is 1.75m

barometric\_pressure In Torrs. Default is 760Torrs air\_temperature In Celzius (C). Default is 25C

wind\_velocity In meters per second (m/s). Use negative number as head

wind, and positive number as back wind. Default is 0m/s (no wind)

MSS, MAC Numeric vectors. Model parameters

#### Value

optimal\_FV returns s data frame with the following columns

F0 Original F0

V0 Original F0

bodymass Bodymass

inertia Inertia

resistance Resistance

Pmax Maximal power estimated using F0 \* V0 / 4

Pmax\_rel Relative maximal power

slope FV profile slope

distance Distance

time Time to cover distance

Ppeak Peak power estimated quantitatively

**Ppeak\_rel** Relative peak power

**Ppeak\_dist** Distance at which peak power is manifested

Ppeak\_time Time at which peak power is manifested

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**F0\_optim** Optimal F0

F0\_coef Ratio between F0\_optim an F0

V0\_optim Optimal V0

**V0\_coef** Ratio between V0\_optim an V0

Pmax\_optim Optimal maximal power estimated F0\_optim \* V0\_optim / 4

Pmax\_rel\_optim Optimal relative maximal power

slope\_optim Optimal FV profile slope

profile\_imb Percent ratio between slope and optimal slope

time\_optim Time to cover distance when profile is optimal

time\_gain Difference in time to cover distance between time\_optimal and time

Ppeak\_optim Optimal peak power estimated quantitatively

Ppeak\_rel\_optim Optimal relative peak power

Ppeak\_dist\_optim Distance at which optimal peak power is manifested

**Ppeak\_time\_optim** Time at which optimal peak power is manifested

optimal\_MSS\_MAC returns a data frame with the following columns

MSS Original MSS

MAC Original MAC

Pmax\_rel Relative maximal power estimated using MSS \* MAC / 4

slope Sprint profile slope

distance Distance

time Time to cover distance

MSS\_optim Optimal MSS

MSS\_coef Ratio between MSS\_optim an MSS

MAC\_optim Optimal MAC

MAC\_coef Ratio between MAC\_optim an MAC

Pmax\_rel\_optim Optimal relative maximal power estimated using MSS\_optim \* MAC\_optim / 4

**slope\_optim** Optimal sprint profile slope

profile\_imb Percent ratio between slope and optimal slope

time\_optim Time to cover distance when profile is optimal

time\_gain Difference in time to cover distance between time\_optimal and time

#### References

Samozino P, Peyrot N, Edouard P, Nagahara R, Jimenez-Reyes P, Vanwanseele B, Morin J. 2022. Optimal mechanical force-velocity profile for sprint acceleration performance. Scandinavian Journal of Medicine & Science in Sports 32:559–575. DOI: 10.1111/sms.14097.

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#### **Examples**

```
MSS <- 10
MAC <- 8
bodymass <- 75
fv <- create_FVP(MSS, MAC, bodymass)</pre>
dist <- seq(5, 40, by = 5)
opt_MSS_MAC_profile <- optimal_MSS_MAC(</pre>
  distance = dist,
  MSS,
  MAC
)[["profile_imb"]]
opt_FV_profile <- optimal_FV(</pre>
  distance = dist,
  fv$F0,
  fv$V0,
  fv$bodymass
)[["profile_imb"]]
opt_FV_profile_peak <- optimal_FV(</pre>
  distance = dist,
  fv$F0,
  fv$V0,
  fv$bodymass,
  method = "peak"
)[["profile_imb"]]
plot(x = dist, y = opt_MSS_MAC_profile, type = "1", ylab = "Profile imbalance")
lines(x = dist, y = opt_FV_profile, type = "l", col = "blue")
lines(x = dist, y = opt_FV_profile_peak, type = "1", col = "red")
abline(h = 100, col = "gray", lty = 2)
```

plot.shorts\_model

S3 method for plotting shorts\_model object

#### **Description**

S3 method for plotting shorts\_model object

```
## S3 method for class 'shorts_model'
plot(x, type = "model", ...)
```

predict.shorts\_model 29

#### **Arguments**

```
    x shorts_model object
    type Type of plot. Can be "model" (default), "kinematics-time", "kinematics-distance", or "residuals"
    ... Not used
```

#### Value

```
ggplot object
```

#### **Examples**

```
# Simple model with radar gun data
instant_velocity <- data.frame(
   time = c(0, 1, 2, 3, 4, 5, 6),
   velocity = c(0.00, 4.99, 6.43, 6.84, 6.95, 6.99, 7.00)
)

radar_model <- with(
   instant_velocity,
   model_radar_gun(time, velocity)
)

plot(radar_model)
plot(radar_model, "kinematics-time")
plot(radar_model, "kinematics-distance")
plot(radar_model, "residuals")</pre>
```

# **Description**

S3 method for making predictions using shorts\_model

# Usage

```
## S3 method for class 'shorts_model'
predict(object, ...)
```

# **Arguments**

#### **Examples**

```
split_distances <- c(10, 20, 30, 40, 50)
split_times <- create_timing_gates_splits(
    gates = split_distances,
    MSS = 10,
    MAC = 9,
    FD = 0.25,
    TC = 0
)

# Simple model
simple_model <- model_timing_gates(split_distances, split_times)
predict(simple_model)</pre>
```

predict\_kinematics

Kinematics prediction functions

# **Description**

Predicts kinematic from known MSS and MAC parameters

```
predict_velocity_at_time(time, MSS, MAC)

predict_distance_at_time(time, MSS, MAC)

predict_acceleration_at_time(time, MSS, MAC)

predict_time_at_distance(distance, MSS, MAC)

predict_time_at_distance_FV(
    distance,
    F0,
    V0,
    bodymass = 75,
    inertia = 0,
    resistance = 0,
    ...
)

predict_velocity_at_distance(distance, MSS, MAC)

predict_acceleration_at_distance(distance, MSS, MAC)

predict_acceleration_at_velocity(velocity, MSS, MAC)
```

```
predict_air_resistance_at_time(time, MSS, MAC, ...)
predict_air_resistance_at_distance(distance, MSS, MAC, ...)
predict_force_at_velocity(
  velocity,
 MSS,
 MAC,
 bodymass = 75,
  inertia = 0,
  resistance = 0,
)
predict_force_at_time(
  time,
 MSS,
 MAC,
 bodymass = 75,
  inertia = 0,
 resistance = 0,
)
predict_force_at_distance(
 distance,
 MSS,
 MAC,
 bodymass = 75,
  inertia = 0,
  resistance = 0,
)
predict_power_at_distance(
  distance,
 MSS,
 MAC,
 bodymass = 75,
  inertia = 0,
  resistance = 0,
)
predict_power_at_time(
  time,
 MSS,
 MAC,
```

```
bodymass = 75,
     inertia = 0,
     resistance = 0,
   )
   predict_relative_power_at_distance(
     distance,
     MSS,
     MAC,
     bodymass = 75,
     inertia = 0,
     resistance = 0,
   )
   predict_relative_power_at_time(
     time,
     MSS,
     MAC,
     bodymass = 75,
     inertia = 0,
     resistance = 0,
   )
   predict_work_till_time(time, ...)
   predict_work_till_distance(distance, ...)
   predict_kinematics(
     object = NULL,
     MSS,
     MAC,
     max\_time = 6,
     frequency = 100,
     bodymass = 75,
     inertia = 0,
     resistance = 0,
     add_inertia_to_vertical = TRUE,
   )
Arguments
   time, distance, velocity
                   Numeric vectors
   MSS, MAC
                   Numeric vectors. Model parameters
```

F0, V0	Numeric vectors. FV profile parameters
bodymass	Body mass in kg. Used to calculate relative power and forwarded to $\texttt{get\_air\_resistance}$
inertia	External inertia in kg (for example a weight vest, or a sled). Not included in the air resistance calculation
resistance	External horizontal resistance in Newtons (for example tether device or a sled friction resistance)
• • •	Arguments passed on to get_air_resistance
	bodyheight In meters (m). Default is 1.75m
	barometric_pressure In Torrs. Default is 760Torrs
	air_temperature In Celzius (C). Default is 25C
	wind_velocity In meters per second (m/s). Use negative number as head wind, and positive number as back wind. Default is 0m/s (no wind)
object	If shorts_model object is provided, estimated parameters will be used. Otherwise provide MSS and MAC parameters
max_time	Predict from 0 to max_time. Default is 6seconds
frequency	Number of samples within one second. Default is 100Hz
add_inertia_to_	vertical
	Should inertia be added to bodymass when calculating vertical force? Use TRUE (Default) when using weight vest, and FALSE when dragging sled

#### Value

Numeric vector

Data frame with kinetic and kinematic variables

# References

Haugen TA, Tønnessen E, Seiler SK. 2012. The Difference Is in the Start: Impact of Timing and Start Procedure on Sprint Running Performance: Journal of Strength and Conditioning Research 26:473–479. DOI: 10.1519/JSC.0b013e318226030b.

Jovanović, M., Vescovi, J.D. (2020). shorts: An R Package for Modeling Short Sprints. Preprint available at SportRxiv. https://doi.org/10.31236/osf.io/4jw62

Samozino P. 2018. A Simple Method for Measuring Force, Velocity and Power Capabilities and Mechanical Effectiveness During Sprint Running. In: Morin J-B, Samozino P eds. Biomechanics of Training and Testing. Cham: Springer International Publishing, 237–267. DOI: 10.1007/978-3-319-05633-3\_11.

```
MSS <- 8
MAC <- 9

time_seq <- seq(0, 6, length.out = 10)

df <- data.frame(
   time = time_seq,</pre>
```

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```
distance_at_time = predict_distance_at_time(time_seq, MSS, MAC),
 velocity_at_time = predict_velocity_at_time(time_seq, MSS, MAC),
 acceleration_at_time = predict_acceleration_at_time(time_seq, MSS, MAC)
)
df$time_at_distance <- predict_time_at_distance(df$distance_at_time, MSS, MAC)</pre>
df$velocity_at_distance <- predict_velocity_at_distance(df$distance_at_time, MSS, MAC)</pre>
df$acceleration_at_distance <- predict_acceleration_at_distance(df$distance_at_time, MSS, MAC)
df$acceleration_at_velocity <- predict_acceleration_at_velocity(df$velocity_at_time, MSS, MAC)
# Power calculation uses shorts::get_air_resistance function and its defaults
# values to calculate power. Use the ... to setup your own parameters for power
# calculations
df$power_at_time <- predict_power_at_time(</pre>
 time = df$time, MSS = MSS, MAC = MAC,
 # Check shorts::get_air_resistance for available params
 bodymass = 100, bodyheight = 1.85
)
df
# Example for predict_kinematics
split_times <- data.frame(</pre>
 distance = c(5, 10, 20, 30, 35),
 time = c(1.20, 1.96, 3.36, 4.71, 5.35)
)
# Simple model
simple_model <- with(</pre>
 split_times,
 model_timing_gates(distance, time)
)
predict_kinematics(simple_model)
```

print.shorts\_model

S3 method for printing shorts\_model object

#### **Description**

S3 method for printing shorts\_model object

```
## S3 method for class 'shorts_model'
print(x, ...)
```

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#### **Arguments**

```
x shorts_model object... Not used
```

# **Examples**

```
split_distances <- c(10, 20, 30, 40, 50)
split_times <- create_timing_gates_splits(
    gates = split_distances,
    MSS = 10,
    MAC = 9,
    FD = 0.25,
    TC = 0
)

# Simple model
simple_model <- model_timing_gates(split_distances, split_times)
simple_model</pre>
```

probe\_functions

Probe profile functions

# **Description**

Family of functions that serve a purpose of probing sprint or force-velocity profile. This is done by increasing individual sprint parameter for a percentage and calculating which parameter improvement yield biggest deduction in sprint tim

probe\_FV "probes" F0 and V0 and calculates which one improves sprint time for a defined distance probe\_MSS\_MAC "probes" MSS and MAC and calculates which one improves sprint time for a defined distance

```
probe_FV(
   distance,
   F0,
   V0,
   bodymass = 75,
   inertia = 0,
   resistance = 0,
   perc = 2.5,
   ...
)
probe_MSS_MAC(distance, MSS, MAC, perc = 2.5)
```

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#### **Arguments**

distance Numeric vector

F0, V0 Numeric vectors. FV profile parameters

bodymass Body mass in kg

inertia External inertia in kg (for example a weight vest, or a sled). Not included in the

air resistance calculation

resistance External horizontal resistance in Newtons (for example tether device or a sled

friction resistance)

perc Numeric vector. Probing percentage. Default is 2.5 percent

... Arguments passed on to get\_air\_resistance

velocity Instantaneous running velocity in meters per second (m/s)

bodyheight In meters (m). Default is 1.75m

barometric\_pressure In Torrs. Default is 760Torrs air\_temperature In Celzius (C). Default is 25C

wind\_velocity In meters per second (m/s). Use negative number as head

wind, and positive number as back wind. Default is 0m/s (no wind)

MSS, MAC Numeric vectors. Model parameters

#### Value

probe\_FV returns a data frame with the following columns

F0 Original F0

V0 Original F0

bodymass Bodymass

inertia Inertia

resistance Resistance

Pmax Maximal power estimated using F0 \* V0 / 4

Pmax\_rel Relative maximal power

slope FV profile slope

distance Distance

time Time to cover distance

probe\_perc Probe percentage

**F0\_probe** Probing F0

**F0\_probe\_time** Predicted time for distance when F0 is probed

F0\_probe\_time\_gain Difference in time to cover distance between time\_optimal and time

V0\_probe Probing V0

**V0\_probe\_time** Predicted time for distance when V0 is probed

V0\_probe\_time\_gain Difference in time to cover distance between time\_optimal and time

profile\_imb Percent ratio between V0\_probe\_time\_gain and F0\_probe\_time\_gain

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```
probe_MSS_MAC returns a data frame with the following columns
```

MSS Original MSS

MAC Original MAC

Pmax\_rel Relative maximal power estimated using MSS \* MAC / 4

slope Sprint profile slope

distance Distance

time Time to cover distance

probe\_perc Probe percentage

MSS\_probe Probing MSS

MSS\_probe\_time Predicted time for distance when MSS is probed

MSS\_probe\_time\_gain Difference in time to cover distance between probe time and time

MAC\_probe Probing MAC

MAC\_probe\_time Predicted time for distance when MAC is probed

MAC\_probe\_time\_gain Difference in time to cover distance between probing time and time

profile\_imb Percent ratio between MSS\_probe\_time\_gain and MAC\_probe\_time\_gain

```
MSS <- 10
MAC <- 8
bodymass <- 75
fv <- create_FVP(MSS, MAC, bodymass)</pre>
dist < - seq(5, 40, by = 5)
probe_MSS_MAC_profile <- probe_MSS_MAC(</pre>
  distance = dist,
  MSS,
  MAC
)[["profile_imb"]]
probe_FV_profile <- probe_FV(</pre>
  distance = dist,
  fv$F0,
  fv$V0,
  fv$bodymass
)[["profile_imb"]]
plot(x = dist, y = probe_MSS_MAC_profile, type = "1", ylab = "Profile imbalance")
lines(x = dist, y = probe_FV_profile, type = "l", col = "blue")
abline(h = 100, col = "gray", lty = 2)
```

radar\_gun\_data

Radar Gun Data

# Description

Data generated from known MSS and TAU and measurement error for N=5 athletes using radar gun with sampling frequency of 100Hz over 6 seconds.

# Usage

```
data(radar_gun_data)
```

#### **Format**

Data frame with 4 variables and 3000 observations:

athlete Character string

bodyweight Bodyweight in kilograms

time Time reported by the radar gun in seconds

velocity Velocity reported by the radar gun in m/s

residuals.shorts\_model

S3 method for returning residuals of shorts\_model

# Description

S3 method for returning residuals of shorts\_model

# Usage

```
## S3 method for class 'shorts_model'
residuals(object, ...)
```

# Arguments

object shorts\_model object
... Extra arguments. Not used

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#### **Examples**

```
split_distances <- c(10, 20, 30, 40, 50)
split_times <- create_timing_gates_splits(
    gates = split_distances,
    MSS = 10,
    MAC = 9,
    FD = 0.25,
    TC = 0
)

# Simple model
simple_model <- model_timing_gates(split_distances, split_times)
residuals(simple_model)</pre>
```

split\_times

Split Testing Data

# **Description**

Data generated from known MSS and TAU and measurement error for N=5 athletes using 6 timing gates: 5m, 10m, 15m, 20m, 30m, 40m

# Usage

```
data(split_times)
```

#### **Format**

Data frame with 4 variables and 30 observations:

athlete Character string
bodyweight Bodyweight in kilograms

distance Distance of the timing gates from the sprint start in meters

time Time reported by the timing gate

summary.shorts\_model S3 method for providing summary for the shorts\_model object

#### **Description**

S3 method for providing summary for the shorts\_model object

```
## S3 method for class 'shorts_model'
summary(object, ...)
```

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# Arguments

```
object shorts_model object ... Not used
```

#### **Examples**

```
split_distances <- c(10, 20, 30, 40, 50)
split_times <- create_timing_gates_splits(
    gates = split_distances,
    MSS = 10,
    MAC = 9,
    FD = 0.25,
    TC = 0
)

# Simple model
simple_model <- model_timing_gates(split_distances, split_times)
summary(simple_model)</pre>
```

vescovi

Vescovi Timing Gates Sprint Times

#### **Description**

Timing gates sprint times involving 52 female athletes. Timing gates were located at 5m, 10m, 20m, 30m, and 35m. See **Details** for more information.

#### Usage

```
data(vescovi)
```

# Format

Data frame with 17 variables and 52 observations:

**Team** Team or sport. Contains the following levels: 'W Soccer' (Women Soccer), 'FH Sr' (Field Hockey Seniors), 'FH U21' (Field Hockey Under 21), and 'FH U17' (Field Hockey Under 17)

Surface Type of testing surface. Contains the following levels: 'Hard Cours' and 'Natural Grass'

Athlete Athlete ID

**Age** Athlete age in years

Height Body height in cm

Bodyweight Body weight in kg

BMI Body Mass Index

BSA Body Surface Area. Calculated using Mosteller equation sqrt((height/weight)/3600)

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5m Time in seconds at 5m gate

10m Time in seconds at 10m gate

20m Time in seconds at 20m gate

30m Time in seconds at 30m gate

35m Time in seconds at 35m gate

10m-5m split Split time in seconds between 10m and 5m gate

20m-10m split Split time in seconds between 20m and 10m gate

30m-20m split Split time in seconds between 30m and 20m gate

35m-30m split Split time in seconds between 35m and 30m gate

#### **Details**

This data-set represents sub-set of data from a total of 220 high-level female athletes (151 soccer players and 69 field hockey players). Using a random number generator, a total of 52 players (35 soccer and 17 field hockey) were selected for this data-set. Soccer players were older (24.6 $\pm$ 3.6 vs. 18.9 $\pm$ 2.7 yr, p < 0.001), however there were no differences for height (167.3 $\pm$ 5.9 vs. 167.0 $\pm$ 5.7 cm, p = 0.886), body mass (62.5 $\pm$ 5.9 vs. 64.0 $\pm$ 9.4 kg, p = 0.500) or any sprint interval time (p > 0.650).

The protocol for assessing linear sprint speed has been described previously (Vescovi 2014, 2016, 2012) and was identical for each cohort. Briefly, all athletes performed a standardized warm-up that included general exercises such as jogging, shuffling, multi-directional movements, and dynamic stretching exercises. Infrared timing gates (Brower Timing, Utah) were positioned at the start line and at 5, 10, 20, and 35 meters at a height of approximately 1.0 meter. Participants stood with their lead foot positioned approximately 5 cm behind the initial infrared beam (i.e., start line). Only forward movement was permitted (no leaning or rocking backwards) and timing started when the laser of the starting gate was triggered. The best 35 m time, and all associated split times were kept for analysis. The assessment of linear sprints using infrared timing gates does not require familiarization (Moir, Button, Glaister, and Stone 2004).

#### Author(s)

Jason D. Vescovi
University of Toronto
Faculty of Kinesiology and Physical Education
Graduate School of Exercise Science
Toronto, ON Canada
<vescovij@gmail.com>

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