

¹ JumpMetrics: A Python package computing countermovement and squat jump events and metrics²

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⁶ Summary

⁷ Researchers and practitioners (e.g., sports team scientists, analysts, or coaches) commonly assess countermovement jump and squat jump performance on force plates. A countermovement jump involves a vertical jump where the jumper first dips downwards before immediately jumping upwards as high as possible. In contrast, a squat jump is a vertical jump whereby the jumper pauses briefly in the bottom “squat” position after dipping downwards, and then jumps upwards to minimize contributions from the stretch-shortening cycle. A scientist, analyst, or coach can use data from these vertical jump variations for various reasons, such as to evaluate people’s neuromuscular capacity, injury risk, or readiness/preparedness for high-intensity training. For evaluating capacity, some researchers have examined variables such as maximum jump height, peak force, rate of force development, and impulse [@mcmahon:2017]. When assessing injury risk, researchers have examined landing forces (e.g., [@pedley2020]) or have leveraged statistical techniques on various countermovement jump variables simultaneously [@bird:2022]. For training readiness, researchers have measured changes in vertical jump height between consecutive training sessions (e.g., [@watkins:2017]). Examining the difference in performance between the two jump variations may also provide insights into the strengths and weaknesses of the athlete to direct future training [@vanHooren2017]. For example, some researchers may compute an “eccentric utilization ratio” by comparing metrics from the countermovement jump relative to the squat jump and use that to inform whether someone should focus their training on improving their ability to leverage the stretch shortening cycle to maximize jumping performance [@vanHooren2017]. Although it is common to collect this kinetic data from a force plate for these jump variations for several applications, there are currently no free, open-source resources to detect events and compute metrics for reproducible and accessible data processing. JumpMetrics fills this gap in both applied practice and in the sports science literature.

³¹ Statement of need

³² JumpMetrics is a free, open-source Python package for computing countermovement jump and squat jump events and metrics. Currently, there are no free alternatives for processing force plate data for vertical jumps relative to the numerous proprietary (i.e., closed-source) algorithms sold by commercial force plates companies (e.g., Vald, Hawkin Dynamics). This makes both the analyses, and ensuring reproducibility of results from these analyses, more challenging for the various professionals conducting jump analyses from force plate data. The API for JumpMetrics was designed to be modular and easy to use for those familiar with Python. Researchers, data scientists, analysts, and even coaches can examine each jump’s takeoff and landing phases individually or together, depending on their needs and research questions. Furthermore, JumpMetrics provides various helper functions to prepare the data for

42 detecting events and computing metrics. These helper functions involve cropping longer force
43 traces and digitally low-pass filtering time series data (if these processing steps are required for
44 a researcher or practitioner's particular analysis). In addition to its applications in sports science
45 research and in professional practice, JumpMetrics can also be leveraged by undergraduate or
46 graduate students to assist with sport or exercise science-related projects. Finally, the code,
47 docstrings, and documentation provided in this package may help sports science students learn
48 Python programming if they aren't already familiar.

49 State of the Field

50 Force plate analysis for vertical jumps has traditionally been dominated by proprietary software
51 solutions from commercial manufacturers such as Vald (ForceDecks) and Hawkin Dynamics,
52 which provide closed-source algorithms for event detection and metrics computation. While
53 these commercial platforms offer comprehensive solutions, they limit reproducibility and
54 accessibility for researchers and practitioners with budget constraints. Furthermore, although
55 these solutions may provide APIs for practitioners to build their own data pipelines, they offer
56 limited flexibility for tuning the actual algorithms to compute jump-related metrics.

57 Several open-source tools exist for biomechanical analysis, though none specifically address
58 force plate-based countermovement and squat jump analysis in Python. biomechZoo ([Dixon et al., 2017](#))
59 is a MATLAB-based toolbox providing general biomechanical data processing
60 capabilities, including ground reaction force processing. However, it requires MATLAB licensing
61 and is not jump-specific. The R-based shiny-vertical-jump application (GPL-3.0 license
62 on GitHub at [mattsams89/shiny-vertical-jump](#)) provides vertical jump analysis functionality
63 but operates within the R ecosystem. Other open-source alternatives like Chronojump and
64 video-based tools (e.g., Kinovea) utilize different measurement modalities (jump mats or
65 computer vision) rather than force plates.

66 JumpMetrics fills a unique gap as the first Python-based, open-source toolkit specifically
67 designed for force plate analysis of countermovement and squat jumps. By integrating with
68 the scientific Python ecosystem (NumPy, SciPy, pandas, matplotlib), it provides accessible,
69 reproducible jump analysis for researchers, practitioners, and students without licensing costs
70 or proprietary constraints. Furthermore, it provides flexibility to tune analyses based on the
71 unique constraints and needs of researchers and practitioners.

72 Software Design

73 The design of JumpMetrics prioritizes modularity, extensibility, and integration with the
74 scientific Python ecosystem. The architecture follows three core principles. First, object-
75 oriented processor classes provide a consistent API across jump types. All processor classes
76 inherit from a common ForceTimeCurveTakeoffProcessor base class that handles kinematic
77 calculations (acceleration, velocity, displacement) via numerical integration. Specialized
78 classes (ForceTimeCurveCMJTakeoffProcessor, ForceTimeCurveSQJTakeoffProcessor,
79 ForceTimeCurveJumpLandingProcessor) implement jump-specific event detection algorithms
80 while maintaining a uniform interface for metric computation. Second, the dual-level API
81 design supports both rapid batch processing and fine-grained analysis. The high-level
82 process_jump_trial() wrapper function handles complete pipelines (cropping, filtering,
83 event detection, metrics computation) for batch processing scenarios. Alternatively, users
84 can directly instantiate processor classes for granular control over individual processing steps,
85 custom visualizations, or algorithm debugging. Finally, tunable parameters with sensible
86 defaults balance ease-of-use with adaptability. Event detection thresholds, filtering parameters,
87 and phase duration criteria all use default values validated against real force plate data, but
88 remain configurable to accommodate different experimental protocols and data collection
89 methodologies.

90 This design reflects trade-offs between simplicity and flexibility: default parameters enable
 91 immediate use for standard protocols, while exposed parameters allow researchers to adapt
 92 algorithms to novel jump variations, research designs, or equipment constraints. Integration
 93 with SciPy's signal processing, pandas DataFrames, and matplotlib visualization ensures
 94 compatibility with existing scientific Python workflows.

95 How the Library Works

96 Event Detections and Metrics

97 JumpMetrics computes various events and metrics for the countermovement and squat jump
 98 leveraging the vertical axis data from a force plate. The events (and thus metrics) are slightly
 99 different between the jump variations, given the differences in their movement executions.
 100 There are classes for processing various phases of the countermovement jump, a helper function
 101 to process the entire jump and landing, as well as individual functions for even more granularity
 102 for analyses. Furthermore, JumpMetrics computes the vertical axis acceleration, velocity, and
 103 displacement of the estimated center of mass trajectory for each frame of data, irrespective
 104 of whether one is using a triaxial or uniaxial force plate. These data are computed by first
 105 dividing the force trace by the individual's computed bodymass to obtain the acceleration data.
 106 Then, the acceleration signal is integrated to compute the instantaneous velocity. Finally, the
 107 signal is integrated one more time to compute the instantaneous displacement. Some examples
 108 of these computed waveforms are shown in Figures 1, 2, and 3.

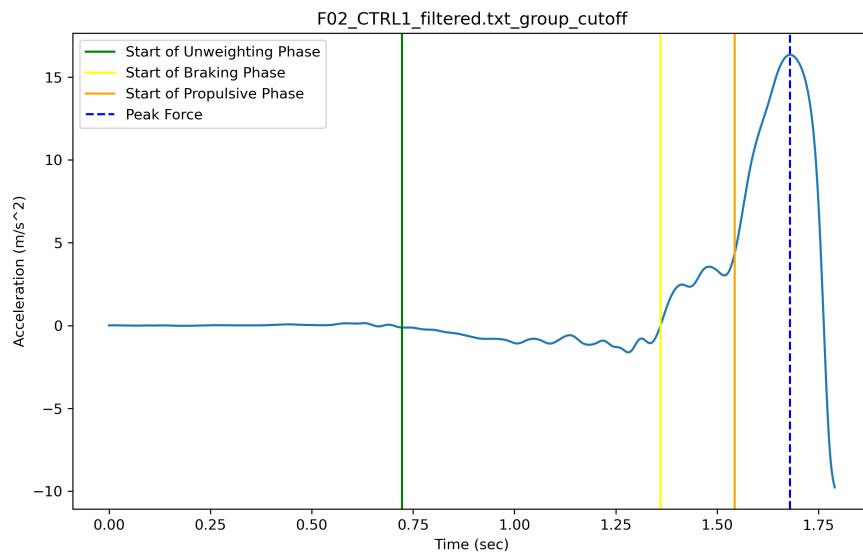


Figure 1: Example acceleration waveform (i.e., the raw force trace divided by the body mass, which was estimated from bodyweight).

109 Figure 1. Example countermovement jump acceleration trace with events detected during
 110 the takeoff phase. Positive accelerations represent the center of mass accelerating upwards,
 111 whereas negative accelerations represent the center of mass accelerating downwards towards
 112 the floor/force plate.

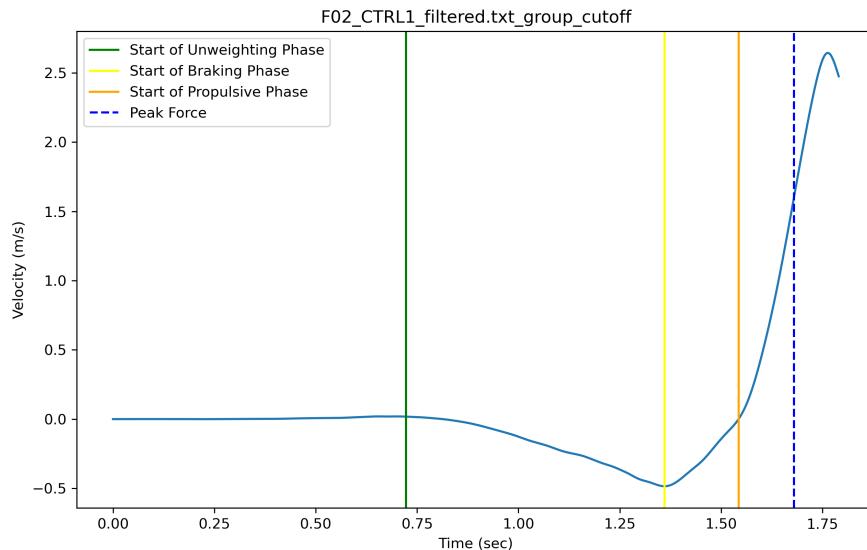


Figure 2: Example velocity waveform (i.e., the integrated acceleration waveform, assuming 0 velocity during quiet standing).

113 Figure 2. Example countermovement jump velocity trace with events detected during the
 114 takeoff phase. Positive velocities represent the center of mass is moving upwards, whereas
 115 negative velocities represent the center of mass moving downwards.

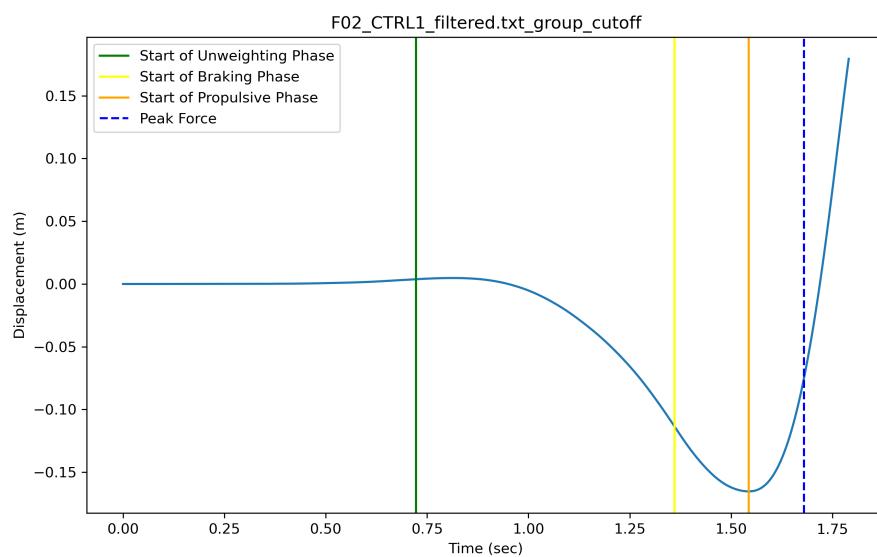


Figure 3: Example displacement waveform (i.e., the integrated velocity waveform, assuming 0 displacement during quiet standing).

116 Figure 3. Example countermovement jump displacement trace with events detected during the
 117 takeoff phase. Positive displacements represent the center of mass being higher relative to
 118 quiet standing, whereas negative displacements represent the center of mass being closer to
 119 the floor relative to quiet standing.

120 To compute the relevant events and metrics, there are specific methods that a user should
121 adhere to that are outlined in @mcmahon:2018. These methods underpin the assumptions
122 required to collect and process data with the code provided in this package. First, the jumper
123 must stand still (i.e., minimizing swaying or any other body movements) at the start of the data
124 collection and for at least 1 second before starting the initiation of the jump. This quiet standing
125 is used to calculate one's bodyweight, and bodyweight is used for subsequent acceleration,
126 velocity, and displacement calculations used for event detections (as well as for computing net
127 vertical impulse). The default setting in this package is currently to use the first 0.4 seconds
128 of the trial to compute bodyweight (as this was found to work well for previous analyses),
129 but users can tune this parameter themselves for their own data collections depending on the
130 length of the quiet standing at the start of the data collection. For a countermovement jump,
131 the functions in this package also require the person to perform one continuous downwards
132 and upwards motion during the jump; any pausing may negatively impact the event detection
133 algorithms provided. In contrast, for the squat jump the default parameter for identifying the
134 start of the propulsive phase expects at least a 1 second pause. In practice, previous research
135 has outlined a pause should be approximately 3 seconds [@vanHooren2017]. The functions
136 provided in `JumpMetrics` permit the user to select a different minimum pause to assume if the
137 default of 1 second is not appropriate for their research.

138 Phases Until Takeoff

139 Countermovement Jumps There are two main phases preceding the moment of takeoff
140 during countermovement jumps. The two phases are the “lowering” (sometimes referred to as
141 the eccentric) and “ascending” (sometimes referred to as the concentric) phases. The specific
142 events within these two phases detected in this package are based on @mcmahon:2018. Each
143 event corresponds the start of each subphase within the jump. These involve: 1) the start of
144 the unweighting phase, 2) the start of the braking phase, 3) the start of the propulsive phase
145 (the event which separates the lowering and ascending phases), 4) the frame corresponding to
146 the peak force, and 5) the frame corresponding to takeoff.

147 `JumpMetrics` computes the start of the unweighting phase in the same manner as outlined in
148 @owen:2014 whereby the first frame of force data that exceeds five times the standard deviation
149 of the force data (default value; this is a tuneable parameter depending on the data collection
150 parameters) during quiet standing (sometimes referred to as the weighing phase) defines the
151 start of the unweighting phase. The braking phase starts at the frame corresponding to the
152 maximum downward movement velocity (of the individual's estimated center of mass; see
153 Figure 2). The propulsive phase starts at the frame corresponding to the minimum downward
154 displacement (of the individual's estimated center of mass; see Figure 3). This event is used
155 because it is reliably detected and avoids any potential awkwardness of using a minimum
156 positive velocity to define the start of this event whereby the person is moving upwards, but
157 is still considered to be in the “braking” phase (e.g., if the minimum threshold to determine
158 the start of the propulsive phase is 0.10m/s, but the person is currently moving 0.05m/s
159 upwards they would still be in the “braking” phase). The peak force event is captured using
160 `find_peaks` from the `scipy` package and looks for a “peak” in the force series. The takeoff
161 event is detected by looking for the first frame of data whereby the force series is below a
162 certain threshold (default is 10 Newtons) for a specific period of time (default is 0.25 seconds).
163 `JumpMetrics` makes these various parameters for detecting events tunable in cases where the
164 defaults may not accurately detect events due to unexpected noise in the data or if there are
165 any changes in the methodology proposed with future research.

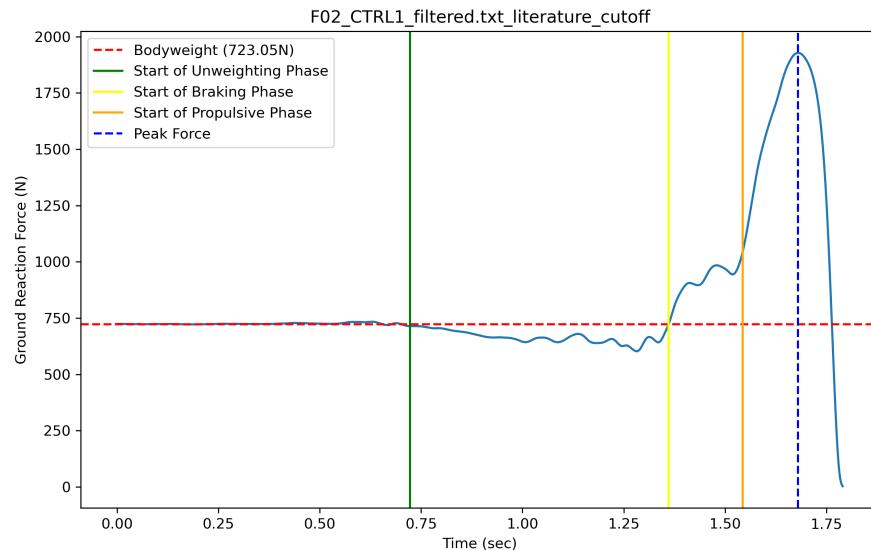


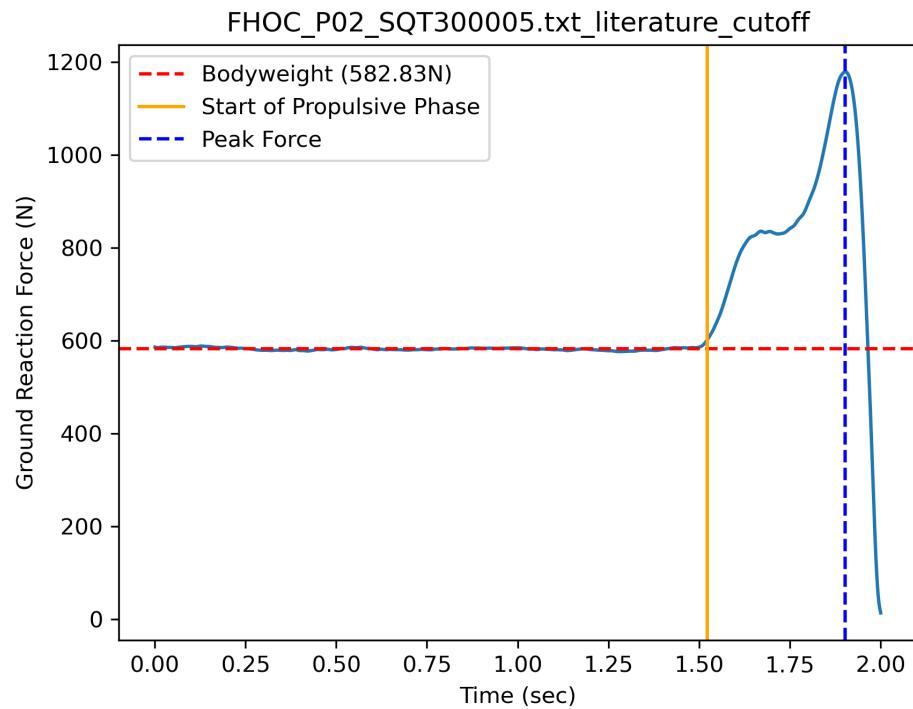
Figure 4: Example countermovement jump force-time trace with events detected during the takeoff phase.

166 Figure 4. Example countermovement jump force-time trace with events detected during the
167 takeoff phase.

168 JumpMetrics computes the rate of force development, net vertical impulse, and average force
169 between all events detected during the countermovement jump. Additionally, metrics such as
170 the jump height (based on the net vertical impulse using the impulse-momentum relationship
171 as well as the center of mass velocity at the final frame of data before takeoff), takeoff velocity,
172 movement time, unweighting time, braking time, propulsive time, and lowering displacement
173 are also all computed. Table 1 contains a complete list of the metrics JumpMetrics exports
174 for countermovement jumps. Note that in order for the events and metrics to be computed
175 accurately, a brief weighing phase (default is the first 0.25 seconds of data, but this can be
176 modified depending on how long the individual was standing) must be present at the start
177 of the waveform in order to determine the individual's bodyweight and to detect subsequent
178 jump events. Table 2 contains a small sample of how, more generally, some of the metrics
179 computed in JumpMetrics can be leveraged for various applications (please note that this table
180 only covers potential applications and the final decision of when to use a particular metric
181 is dependent on how the data is collected and the particular research question the analyst
182 wishes to answer). Critically, the helper functions in jump metrics, such as `compute_rfd` (i.e.,
183 compute rate of force development) allow the user to easily compute any necessary metric
184 beyond the defaults computed in the current version of this package.

185 **Squat Jumps** Given that the squat jump is intentionally performed with a pause
186 to minimize the influence of the stretch shortening cycle of the lower body muscles
187 from a continuous countermovement (i.e., there is no lowering phase), the only events
188 JumpMetrics detects are the start of the propulsive phase, the peak force event,
189 and the takeoff event. The start of the propulsive phase is the first frame of data
190 that exceeds five times the standard deviation of the force data (default value; this is
191 a tuneable parameter depending on the data collection parameters) during the squat
192 phase. The peak force event is computed similarly to the countermovement jumps whereby
193 `find_peaks` from the `scipy` package and looks for a "peak" in the force series. The takeoff
194 event is detected by looking for the first frame of data whereby the force series is below

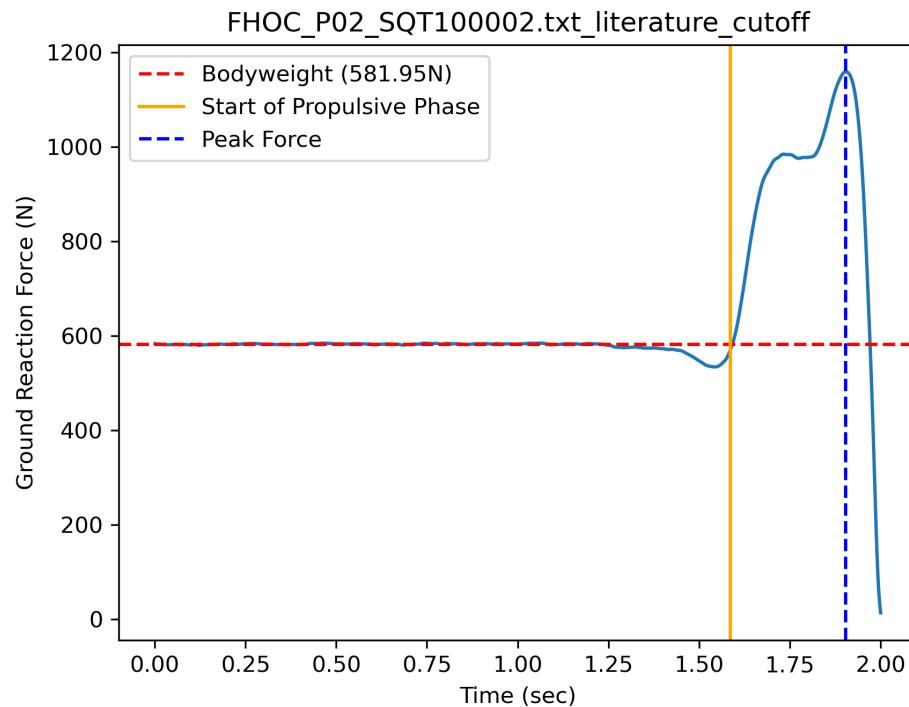
¹⁹⁵ a certain threshold (default is 10 Newtons) for a specific period of time (default is 0.25 seconds).



¹⁹⁶

¹⁹⁷ Figure 5. Example squat jump force-time trace during the takeoff phase.

¹⁹⁸ Although there is not supposed to be any countermovement/lowering phase during a
¹⁹⁹ squat jump, depending on the instructions and guidance provided to the participant,
²⁰⁰ as well as their general movement behaviours, there may be a minor countermovement
²⁰¹ that would negate the trial from being a true squat jump. JumpMetrics detects
²⁰² and flags this motion (with a warning and an estimated frame in the metrics
²⁰³ output) to make the user aware of this potential flaw in the squat jump trial.



204
205 Figure 6. Example squat jump force-time trace with an inappropriate countermovement
206 detected during the takeoff phase.

207 JumpMetrics computes the rate of force development, net vertical impulse, and average force
208 between all events detected during the squat jump. Additionally, metrics such as the jump
209 height (based on the net vertical impulse and using the impulse-momentum relationship
210 and the velocity at the final frame of data before takeoff), takeoff velocity, movement time,
211 and propulsive time are also all computed. Table 1 contains a complete list of the metrics
212 JumpMetrics exports for squat jumps. Note that in order for the events and metrics to be
213 computed accurately, a brief weighing phase (default is the first 0.25 seconds of data, but this
214 can be modified depending on how long the individual was standing) must be present in the
215 waveform in order to determine the individual's bodyweight.

216 **Landing Phase**

217 The landing phase is defined by the methodology outlined in @mcmahon:2018. The initial
218 landing event is detected by looking for the first frame of data whereby the force series is
219 above a certain threshold (default is 20 Newtons) for a specific period of time (default is
220 0.015 seconds). The landing phase's end is the point where the estimated center of mass
221 velocity becomes greater than, or equal to, 0 meters per second (given that a negative velocity
222 represents a downward movement).

223 During the jump's landing phase, JumpMetrics computes the maximum landing force, average
224 landing force, landing time, landing displacement, and various landing rate of force development
225 metrics.

226 **Wrapper Function**

227 There is also a wrapper function named `process_jump_trial()` that combines the classes for
228 both the phases up to takeoff and following the landing of the jump to compute all relevant
229 events and metrics. Additionally, because this function uses the entire force trace as its input,

230 this function also computes two additional metrics. These metrics are the jump height based
231 on the flight time (i.e., the between takeoff and landing) as well as the flight time itself.
232 Although jump height based on flight time is less robust relative to jump height based on
233 the impulse-momentum theorem [xu:2023], the flight time jump height is still included for
234 researchers or practitioners who may have historical reference data using this method.

235 **Tables**

Table 1. Specific metrics computed and exported by JumpMetrics at Takeoff for Countermovement Jumps (CMJ) and Squat Jumps (SJ). | Metric | CMJ | SJ |

238 Description | |
239 | propulsive_peakforce_rfd_slope_between_events | yes | yes | Rate of force development
240 computed as the slope between the start of the propulsive phase to the frame of
241 peak force (N/s) | | propulsive_peakforce_rfd_instantaneous_average_between_events
242 | yes | yes | Rate of force development computed as the average instantaneous value
243 between each frame the start of the propulsive phase to the frame of peak force
244 (N/s) | | propulsive_peakforce_rfd_instantaneous_peak_between_events | yes | yes
245 | Rate of force development computed as the peak instantaneous value between
246 each frame the start of the propulsive phase to the frame of peak force (N/s) | |
247 braking_peakforce_rfd_slope_between_events | yes | | Rate of force development
248 computed as the slope between the start of the braking phase to the frame of peak
249 force (N/s) | | braking_peakforce_rfd_instantaneous_average_between_events | yes
250 | | Rate of force development computed as the average instantaneous value between
251 each frame during the start of the braking phase to the frame of peak force (N/s) | |
252 braking_peakforce_rfd_instantaneous_peak_between_events | yes | | Rate of force development
253 computed as the peak instantaneous value between each frame during the start of the braking
254 phase to the frame of peak force (N/s) | | braking_propulsive_rfd_slope_between_events |
255 yes | | Rate of force development slope from the braking phase to the start of the propulsive
256 phase (N/s) | | braking_propulsive_rfd_instantaneous_average_between_events | yes | | Rate
257 of force development computed as the average instantaneous value from the braking to the
258 propulsive phase (N/s) | | braking_propulsive_rfd_instantaneous_peak_between_events |
259 yes | | Rate of force development computed as the peak instantaneous value from braking
260 to the propulsive phase (N/s) | | braking_net_vertical_impulse | yes | | Net vertical impulse
261 during the braking phase (N s) | | propulsive_net_vertical_impulse | yes | | Net vertical
262 impulse during the propulsive phase (N s) | | braking_to_propulsive_net_vertical_impulse
263 | yes | | Net vertical impulse between the start of the braking phase to the propulsive
264 phase (N s) | | total_net_vertical_impulse | yes | yes | Net vertical impulse during the entire
265 jump (N s) | | peak_force | yes | yes | Peak force defined using find_peaks() in scipy (N) | |
266 maximum_force | yes | yes | Global maximum value of force recorded during the jump (N) | |
267 average_force_of_braking_phase | yes | | Average force applied during the braking phase (N) | |
268 average_force_of_propulsive_phase | yes | yes | Average force applied during the propulsive
269 phase (N) | | takeoff_velocity | yes | yes | Estimated takeoff velocity of the center of mass
270 (m/s) | | jump_height_takeoff_velocity | yes | yes | Jump height calculated using the estimated
271 takeoff velocity of the center of mass (m) | | jump_height_net_vertical_impulse | yes | yes
272 | Jump height calculated using the impulse-momentum theorem. Usually identical to the
273 jump height based on takeoff velocity except for rounding errors due to slightly different
274 computations (m) | | jump_height_flight_time** | yes | yes | Jump height calculated using
275 flight time (m) | | flight_time** | yes | yes | Flight time of the jump (s) | | movement_time | yes
276 | yes | Time between initiation of jump and takeoff (s) | | unweighting_time | yes | | Time during
277 the unweighting phase of the jump (s) | | braking_time | yes | | Time during the braking phase
278 of the jump (s) | | propulsive_time | yes | yes | Time during the propulsive phase of the jump
279 (s) | | lowering_displacement | yes | | Estimated displacement of the center of mass during the
280 lowering phase (m) | | frame_start_of_unweighting_phase | yes | | Frame corresponding to the
281 start of the unweighting phase | | frame_start_of_breaking_phase | yes | | Frame corresponding

282 to the start of the braking phase| | frame_start_of_propulsive_phase | yes | yes | Frame
 283 corresponding to the start of the propulsive phase| | frame_peak_force | yes | yes | Frame
 284 corresponding to the peak force during the jump| | frame_of_potential_unweighting_start | |
 285 yes | Frame corresponding to the potential start of the unweighting phase|

286 Table 2. Table of general metrics along with potential typical applications. Note that the
 287 appropriateness of a potential metric for the listed purpose (or beyond the listed purpose) is
 288 dependent on how the data was collected and the particular question the analyst is attempting
 289 to answer. This table is a guide only based on a limited selection of papers.

| General Metric | Capacity | Injury Risk | Readiness | Example Paper(s) |
|---------------------------------|----------|----------------------|-----------|-------------------------------------|
| RFD (Rate of Force Development) | yes | yes | | [@mcmahon:2017], [@bird:2022] |
| Net Vertical Impulse | yes | yes | | [@mcmahon:2017], [@bird:2022] |
| Peak Force | yes | yes (during landing) | | [@mcmahon:2017], [@pedley2020] |
| Maximum Force | yes | yes (during landing) | | [@mcmahon:2017], [@pedley2020] |
| Jump Height | yes | | yes | [@watkins:2017], [@mcmahon:2017] |
| Braking Time | | yes | | [@bird:2022] |
| Propulsive Time | | yes | | [@bird:2022] |

290 Note that jump_height_flight_time and flight_time are only available when using the
 291 process_jump_trial() function as both takeoff and landing must be detected to determine
 292 the flight time.

293 Research Impact

294 Since its initial release in June 2024, JumpMetrics has begun establishing itself within the
 295 biomechanics research and sports science communities. The software is currently being used by
 296 external researchers and practitioners for force plate jump analysis, with adoption in educational
 297 settings for teaching biomechanics concepts in sports science programs.

298 As the only open-source, Python-based tool specifically designed for force plate
 299 counter movement and squat jump analysis, JumpMetrics addresses a critical gap in
 300 accessible biomechanical analysis tools. The project demonstrates active development and
 301 maintains comprehensive test coverage (84% overall, 100% on critical algorithms) across
 302 203 unit and integration tests at the time of publication, ensuring reliability for scientific
 303 applications.

304 The GitHub repository has garnered interest from the biomechanics community with 12 stars
 305 and community engagement at the time of writing, indicating recognition of its value for
 306 reproducible jump analysis. As the software matures, it will provide a foundation for reproducible
 307 research in jump biomechanics, enabling researchers worldwide to apply consistent, validated
 308 algorithms without proprietary software constraints.

309 Beyond immediate research applications, JumpMetrics serves as an educational resource
 310 for students learning both biomechanical analysis and scientific Python programming, with
 311 comprehensive documentation and example workflows demonstrating best practices in sports
 312 science data analysis.

313 AI Usage Disclosure

314 No generative AI tools were used in the development of this software, the writing of this
315 manuscript, or the preparation of supporting materials.

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317 I acknowledge contributions from Malinda Hapuarachchi for providing the data required to
318 develop, test, and verify the functions in this package. I also wish to acknowledge Jacob Rauch
319 for his feedback on the paper.

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