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THE RELIABILITY OF POTENTIAL FATIGUE MONITORING MEASURES IN ELITE YOUTH SOCCER PLAYERS

Running Head: Reliability of Potential Fatigue Monitoring Measures

John F. Fitzpatrick^{1,2}, Kirsty M. Hicks¹, Mark Russell³, Philip R. Hayes¹

Author Affiliations:

¹Department Sport, Exercise and Rehabilitation, Northumbria University, Newcastle-upon-Tyne, UK.

²Sports Science and Medical Department, Newcastle United Football Club, Newcastle-upon-Tyne, UK.

³School of Social and Health Sciences, Leeds Trinity University, UK.

Corresponding Author:

Mr John F. Fitzpatrick, Sports Science and Medical Department, Newcastle United Football Club, Greenlee Drive, Little Benton, Newcastle upon Tyne, NE7 7SF; Tel: 0191 2153200; Email: john.fitzpatrick@nufc.co.uk

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1 ABSTRACT

2 Monitoring fatigue is of vital importance to practitioners, however, logistics and concerns
3 about reliability may impede the use of certain measures. This study aimed to quantify the
4 reliability of potential measures of fatigue; a subjective wellness questionnaire, jump
5 performance tests and tri-axial accelerometer variables derived during sub-maximal shuttle
6 running in elite youth soccer players. A secondary aim was to establish the minimum test
7 duration that could be used for the sub-maximal shuttle run while maintaining good reliability.
8 Seventeen male youth team players (age: 17.4 ± 0.5 years) were assessed on two occasions,
9 spaced seven days apart. Typical error (TE), coefficient of variation (CV%), interclass
10 correlation (ICC) and minimum detectable change (MDC) were calculated for a subjective
11 wellness questionnaire, countermovement jump (CMJ), squat jump (SJ) and drop jump contact
12 time (DJ-CT), jump height (DJ-JH), and reactive strength (DJ-RSI). A novel sub-maximal
13 shuttle running test was also used to assess tri-axial accelerometer data reliability. Results
14 suggest that CMJ, SJ, DJ-CT and DJ-RSI have good test re-test reliability (CV% = 4.5 – 7.7;
15 ICC = 0.80 – 0.88), however DJ-JH did not show acceptable reliability (CV% = 6.0; ICC =
16 0.76). Good reliability was found for all tri-axial accelerometer variables during a 3 min (2 min
17 analysis) sub-maximal shuttle run (CV% = 2.4 – 8.0; ICC = 0.81 – 0.95), except for %
18 PlayerLoadTM anterior–posterior (%PL_{AP}) (CV% = 7.2; ICC = 0.63). The subjective wellness
19 questionnaire demonstrated poor reliability for all items (CV% = 11.2 – 30.0; ICC = 0.00 –
20 0.78). The findings from this study provide practitioners with valuable information about the
21 reliability of a range of potential fatigue monitoring measures. This can be used to help make
22 accurate decisions about the magnitude of change in these assessments when used in practice.

23

24 **Key Words:** reliability; subjective wellness; jump performance; tri-axial accelerometer; sub-
25 maximal testing.

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1 INTRODUCTION

2 Team sport activity has been shown to elicit fatigue commensurate with performance
3 decrements and increased injury risk in youth and senior players (23, 25). Therefore, the ability
4 to monitor and manage training and fatigue is of vital importance to coaches and practitioners
5 (14). In an attempt to make informed decisions about readiness to train and training
6 prescription, practitioners seek methods that attempt to quantify the magnitude of fatigue
7 throughout the competitive week (1, 20). The broad use of the term fatigue within the literature
8 presents a challenge as this can encompass several different phenomena that are the
9 consequence of different physiological and perceptual processes (12). Practitioners have
10 therefore used a number of methods in an attempt to monitor fatigue in an applied setting; self-
11 report measures, autonomic nervous system function, physical performance tests and
12 biochemical markers to name a few (14, 30). However, logistical feasibility and concerns about
13 the reliability may impede the use of such methods on a regular basis (1).

14 An important factor to consider when selecting a potential monitoring tool is measurement
15 reliability. The reliability of a test refers to an acceptable level of consistency between repeated
16 tests within a practically relevant timeframe (2). A test with poor reliability will be unsuitable
17 for tracking changes in fatigue due to an inability to detect a true change in the measure (16).
18 Factors that influence reliability include the protocol, measurement device used to collect the
19 data and any systematic or random changes in the mental or physical state of the individual
20 between trials (2).

21 Self-report measures are widely used in team sports (29), however there has yet to be a
22 consensus on the most appropriate questionnaire to be used. Profile of Mood States (22),
23 Recovery-Stress Questionnaire (18) and Daily Analyses of Life Demands (27) are just some of
24 the assessment tools which have been used within the literature. However, their length, narrow
25 focus or lack of specificity to the sporting context has led many sports programs to develop
26 their own questionnaires (28). Subsequently, practitioners and researchers have incorporated
27 customised, shortened questionnaires into their monitoring practices and research (15, 21),
28 although, reliability and sensitivity of these shortened wellness questionnaires has yet to be

1 established. By contrast, tests of jump performance are well established and demonstrate good
2 reliability, with reported coefficients of variations (CV%) of 5% for the countermovement
3 jump (CMJ) (10), 3% for the squat jump (SJ) (13) and 5-8% for variables derived from a drop
4 jump (DJ) (3).

5 A survey showed that 61% of elite European soccer teams regularly use a sub-maximal, non-
6 exhaustive performance test to assess autonomic function (1). However, research suggests that
7 due to the variability of heart rate measures this approach offers limited meaningful information
8 (19). Notwithstanding issues with reliability, validity and sensitivity, a possible solution maybe
9 to utilise the sub-maximal performance tests that are already widely used in practice, by
10 analysing other data streams that can be collected during this assessment. The use of tri-axial
11 accelerometers, such as those integrated into micro-electro-mechanical systems (MEMS), have
12 demonstrated an ability to detect fatigue post exercise (24), with vertical acceleration showing
13 changes under fatigue during both match-play (9) and training (26). More recently, Buchheit
14 et al. (7) assessed the reliability of stride variables derived from MEMS devices during
15 treadmill running. They found that measures of contact time, flight time and vertical stiffness
16 have good to moderate reliability (4-16% CV). These data give preliminary insights into the
17 ability of accelerometer data to monitor fatigue, however the reliability of measures derived
18 from MEMS devices during sub-maximal field tests has yet to be established.

19 Establishing the reliability of measures used to monitor athletes is an imperative aspect of
20 applied research and practice. Further, field-based, in situ reliability assessments are required
21 in order to quantify “real” changes in potential monitoring tools in athletes within their normal
22 training environment and across time periods that are typically used to quantify the effects of
23 any intervention (2). Therefore, the aim of this study was to quantify the test re-test reliability
24 of a subjective wellness questionnaire, assessments of jump performance and tri-axial
25 accelerometer variables derived during sub-maximal shuttle running in elite youth soccer
26 players. A secondary aim is to establish the minimum test duration that can be used for the sub-
27 maximal shuttle run in order to maintain good reliability.

1 **METHODS**

2 **Experimental Approach to the Problem**

3 This study was completed at the beginning of the 2015-16 season (October) and consisted of
4 two testing sessions spaced seven days apart. All players were in full training during the study,
5 completing around 10.5 h per week of pitch (8.5 h) and gym (2 h) based activity. All players
6 were familiarized with the experimental procedures prior to commencing the study. Each
7 testing session consisted of morning ratings of subjective wellness (n = 17), three different
8 assessments of jump performance; CMJ, SJ and DJ (n = 17) and a sub-maximal shuttle run test
9 used to assess accelerometer variables (n = 15), in respective order. Training load was
10 monitored carefully throughout the study period ensure limited differences between training
11 weeks (sRPE; Week 1 = 2247, Week 2 = 2280, CV% = 9.9%). Both testing sessions were
12 preceded by 48 h of rest and were conducted at the same time of day to limit the influence of
13 possible circadian variation.

14

15 **Subjects**

16 Seventeen youth soccer players (Age: 17.4 ± 0.5 years [range: 16-18 years], Height: $176.7 \pm$
17 5.2 cm, Body Mass: 72.1 ± 9.2 kg), competing in the English Under-18 Premier League agreed
18 to participate in the present study. Participants were given full details of the study procedures
19 and informed of the risks and benefits of the study prior to any data collection. Participants
20 provided personal, and when under 18, parental or guardian, written informed consent before
21 participation. Institutional ethical approval was gained prior to any study involvement. Prior
22 to inclusion in the study, all participants were deemed fit and free of illness or injury by the
23 soccer club's medical staff.

24

25 **Procedures**

26 *Subjective wellness*

1 A psychometric questionnaire based on previous recommendations was collected each day to
2 assess general indicators of player wellness (15). Participants recorded their scores each
3 morning, in private, on an electronic device using a custom made application, as soon as they
4 entered the training ground. The questionnaire was composed of 5 questions relating to fatigue,
5 sleep quality, muscle soreness, stress and mood. Each question was scored on a 5-point Likert
6 scale with 1-point increments (scores of 1–5, with 1 and 5 representing very poor and very
7 good, respectively) (21). Additionally, the summation of all 5 scores provided a total wellness
8 score between 5 - 25.

9 *Jump performance*

10 A standardised warm up consisting of three minutes light aerobic activity on a cycle ergometer
11 at a self-selected pace (Keiser, Fresno, CA, USA), followed by dynamic mobility exercises and
12 three submaximal practice jumps was conducted prior to each testing session. Players then
13 performed three different tests to assess jump performance; CMJ, SJ and DJ. The CMJ was
14 executed to a self-selected depth with the hands placed on the hips. Players were instructed to
15 jump as high as possible with no knee or hip flexion during the flight phase. The same
16 instructions were given for the SJ however, players were instructed to hold their self-selected
17 depth for a four second count. The DJ was performed from a 30 cm box with hands on their
18 hips. Players were instructed to step off the box, rebound off the floor as quickly as possible
19 and jump as high as possible. All participants were well drilled and familiarised with each test.
20 Each assessment consisted of four attempts, separated by one minute of rest. Jumps were
21 performed in a randomized counterbalanced manner to reduce order effects. All jumps were
22 completed using an optical timing system (Optojump, Microgate, Italy), the validity of which
23 has been previously established (13). Jump height was recorded for CMJ and SJ, while for DJ,
24 contact time (DJ-CT), jump height (DJ-JH) and reactive strength index (DJ-RSI) were
25 recorded.

26 *Sub-maximal shuttle running*

27 A sub-maximal shuttle running test was used to assess players' mechanical loading. All players
28 were fitted with a MEMS device (MinimaxX S4, Catapult Sports, Melbourne, Australia) worn

1 between the scapular in a tight-fitting vest to reduce movement artefact. Devices contained a
2 tri-axial piezoelectric linear accelerometer (Kionix: KXP94) sampling at a frequency of 100
3 Hz. Following this, a continuous 20 m shuttle run was performed for a 5 min period, at an
4 average speed of $12 \text{ km}\cdot\text{h}^{-1}$, on an artificial 3G surface. Pacing was controlled using a custom
5 audio track played over a loudspeaker. Data were downloaded using the manufacturer's
6 software (Catapult Sprint, Version 5.1.7) and raw data were exported to Microsoft Excel. The
7 first minute of data was discarded as a stabilization period, the subsequent 2, 3 and 4 minutes
8 of the collection period were used for statistical analysis. Combined tri-axial accelerometer
9 data were presented as PlayerLoad (PL), which is a modified vector magnitude expressed as
10 the square root of the sum of the squared instantaneous rates of change in acceleration in each
11 of the three planes divided by 100 (5). Individual component planes of PL, anterior-posterior
12 [PL_{AP}], mediolateral [PL_{ML}], and vertical [PL_V] were recorded and expressed in arbitrary units
13 (au). The percentage contribution of each component plane to overall PL was also calculated.

14

15 **Statistical Analysis**

16 Test re-test reliability was calculated for each variable and expressed as a typical error (TE),
17 coefficient of variation (CV%) and interclass correlation (ICC), and calculated using a custom
18 spreadsheet (17). To assess reliability, thresholds of $\leq 10\%$ for CV% and ≥ 0.80 for ICC were
19 set (16). To assess the ability of each variable to assess "real" change the minimum detectable
20 change (MDC; 75% confidence level) was also calculated (31). To evaluate the internal
21 consistency of the subjective wellness questionnaire Cronbach's Alpha (α) was calculated (11),
22 with a threshold of >0.7 being set for an acceptable α (4) and inter-item correlations also
23 considered.

24

25 **RESULTS**

26 **Subjective Wellness**

1 Reliability statistics for each subjective wellness measure are shown in Table 1. The TE for
2 individual subjective wellness questions ranged from 0.30 to 0.60 and was 1.59 for total
3 wellness. When expressed as a CV% values ranged from 11.2% to 30.0%. Therefore no
4 subjective wellness measures met the threshold for acceptable reliability of $\leq 10\%$. Similarly,
5 ICCs for subjective wellness measures ranged from -0.01 to 0.78 meaning no measures met the
6 threshold for an acceptable ICC of ≥ 0.80 . The MDC for each subjective wellness measure is
7 displayed in Table 1. As an additional measure of reliability for the subjective wellness
8 questionnaire, Cronbach's Alpha was assessed, this analysis resulted in $\alpha = 0.45$ meaning the
9 internal consistency of the 5 items within this subjective wellness questionnaire is poor. Further
10 analysis into the inter-item correlations is shown in Table 2.

11 INSERT TABLES 1 AND 2 AROUND HERE

12 **Jump Performance**

13 Reliability statistics for each jump test are shown in Table 3. The TE for CMJ, SJ and DJ-JH
14 was 1.6 cm, 1.5 cm and 1.8 cm respectively. When expressed as a CV% these values were
15 4.8%, 4.5% and 6.0% respectively. The TE and CV% for DJ-CT was 0.01 s and 4.9% and for
16 DJ-RSI was 0.11 and 7.7%. Therefore all measures derived from jump assessments met the
17 threshold for acceptable reliability of $\leq 10\%$. Similarly, ICCs for the jump assessments ranged
18 from 0.76 to 0.88 meaning all measures expect for DJ-H (0.76) met the threshold for an
19 acceptable ICC of ≥ 0.80 . The MDC for each jump test is displayed in Table 3.

20 INSERT TABLE 3 AROUND HERE

21 **Sub-Maximal Shuttle Running**

22 A summary of reliability for the accelerometer data from a sub-maximal shuttle run are shown
23 in Table 4. Reliability was consistent across all analysis time frames with all measures meeting
24 the threshold of $\leq 10\%$ for an acceptable CV%. Similarly, ICCs for each accelerometer variable
25 across all time frames ranged from 0.63 to 0.96 meaning that all variables expect for % PL_{AP}
26 (0.63) met the threshold for an acceptable ICC of ≥ 0.80 . The MDC for each accelerometer
27 variable is displayed in Table 4.

1 The assessment of fatigue via jump testing represents a popular method by which to monitor
2 neuromuscular function in the field and has been shown to be associated with objective
3 measures of peripheral fatigue (6). Indeed, a survey indicated that 39% of a sample of 41 elite
4 European soccer teams utilize a form of jump testing on a weekly basis (1). Despite this
5 widespread use, there are only a limited number of studies that have evaluated the test re-test
6 reliability of various jump tests. Findings from the present study indicate good reliability for
7 all jump measures apart from DJ-JH ($ICC = 0.76$), which did not meet the criteria from an
8 acceptable $ICC (\geq 0.80)$. These results are in agreement with previous observations that have
9 reported $CV\%$ of 5% for the countermovement jump (CMJ) (10), 3% for the squat jump (SJ)
10 (13) and 5-8% for variables derived from a drop jump (13). An important consideration when
11 reviewing the literature on reliability of jump assessments is the range of testing modalities
12 used (contact mats, force platforms and photoelectric technology). The present study used the
13 OptoJump system which has previously shown good reliability and validity when assessing
14 CMJ and SJ (13).

15 In the present study all individual subjective wellness items and overall total wellness did not
16 meet the criteria for acceptable reliability. These findings suggest that the current subjective
17 wellness questionnaire is not reliable enough to track changes in fatigue from a week to week
18 basis. To our knowledge, this is the first study to examine the reliability of this type of short
19 psychometric questionnaire that is regularly used in the applied environment. Other, larger
20 questionnaires such as the Recovery-Stress questionnaire (76-questions) have shown large test
21 re-test correlations ($r = 0.79$) (18). The factor/s mediating the poor reliability found in the
22 present study may be due to the simplicity of the questionnaire used. The categorical nature of
23 a 5-point Likert style question means a 1 point change, e.g. from 5 to 4, is the equivalent to a
24 20% decrease. This makes the suggested criteria of a $CV\% \leq 10\%$ difficult to meet.

25 Another aspect of the reliability of a psychometric questionnaire is the internal consistency of
26 each question, this was assessed via Cronbach's α , with results indicating that the internal
27 consistency of this questionnaire is poor ($\alpha = 0.45$). This has implications for the composite
28 total wellness score which is the summation of all five questions. As each item has relatively

1 low inter-item correlations (Table 2) it could be suggested that a composite score for total
2 wellness should be used with caution. Future research should look to amend which items are
3 included in the composite score in order to improve the internal consistency. In conclusion,
4 given the high CV% and low ICC of each variable, and the poor internal consistency of the
5 composite total wellness score, in order to make this subjective wellness questionnaire more
6 reliable and robust in an applied environment the low categorical nature of the Likert scale
7 should be addressed, perhaps by increasing the number of point within the scale to > 5 .
8 Additionally the internal consistency of the 5 items needs to be improved in order to make total
9 wellness a viable measure of fatigue.

10 In conclusion, this study has established good reliability for CMJ, SJ and DJ variables. Tri-
11 axial accelerometer data; PL, PL_{AP}, PL_{ML}, PL_V, %PL_{ML} and %PL_V, gained during sub-maximal
12 shuttle running also displayed good reliability. However, results suggest that subjective
13 wellness assessed via a short 5 item psychometric questionnaire has poor test re-test reliability
14 and internal consistency, therefore, caution must be taken when assessing changes in subjective
15 wellness as an indicator of a players fatigue status. These findings suggest that measures with
16 good week-to-week test re-test reliability may provide the greatest potential as markers of
17 fatigue in an applied environment, with future research looking to establish the sensitivity of
18 these measure to fatigue.

19

20 **PRACTICAL APPLICATIONS**

21 If looking to assess levels of fatigue in youth soccer players, all accelerometer variables, expect
22 for % PL_{AP}, display good test re-test reliability from 3 minutes (2-minutes analysis) of sub-
23 maximal shuttle running, suggesting this may be a novel way of assessing the fatigue levels of
24 a large group of players in a short amount of time. Assessments of CMJ, SJ, DJ-CT and DJ-
25 RSI have displayed good reliability on a week-to-week basis and therefore may provide sound
26 estimates of a player's physical performance level. Finally, the MDC calculated for each
27 variable provides researchers and practitioners with thresholds for what may be considered a

1 “real” change, allowing practitioners to make accurate decisions about the magnitude of
2 fatigue.

3

4 **ACKNOWLEDGEMENTS**

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6 help and expertise during the testing procedures.

Table 1.

Reliability statistics for subjective wellness measures.

Data are presented as group means (\pm SD) for each trial, typical error (TE), interclass correlation (ICC), coefficient of variation (CV%), and minimum detectable change (MDC) (75% confidence level).

	Trial 1 (SD)	Trial 2 (SD)	TE (90% CI)	ICC (90% CI)	CV% (90% CI)	MDC
Fatigue (au)	3.2 (0.5)	3.1 (0.7)	0.3 (0.2, 0.4)	0.78 (0.56, 0.90)	14.9 (11.4, 21.8)	0.5
Sleep Quality (au)	3.8 (0.5)	3.6 (0.6)	0.6 (0.5, 0.8)	-0.01 (-0.41, 0.39)	21.0 (16.1, 31.1)	0.9
Muscle Soreness (au)	3.0 (0.9)	2.7 (0.8)	0.6 (0.5, 0.9)	0.54 (0.17, 0.77)	30.0 (22.7, 45.1)	0.9
Stress (au)	3.4 (0.7)	3.1 (0.9)	0.6 (0.5, 0.8)	0.46 (0.08, 0.73)	22.7 (17.3, 33.6)	0.9
Mood (au)	3.9 (0.6)	3.9 (0.7)	0.6 (0.5, 0.8)	0.15 (-0.27, 0.52)	19.2 (14.7, 28.2)	0.9
Total Wellness (au)	17.3 (1.9)	16.5 (2.0)	1.6 (1.2, 2.3)	0.35 (-0.06, 0.66)	11.2 (8.6, 16.2)	2.5

An acceptable threshold for reliability was set at 0.80 for ICC and 10% for CV%

* ICC \geq 0.80

** CV% \leq 10%

Table 2.
Subjective wellness inter-item correlation matrix

	Fatigue	Sleep	Soreness	Stress	Mood
Fatigue	-	0.12	0.25	0.16	0.04
Sleep	-	-	0.13	0.35	0.18
Soreness	-	-	-	-0.10	-0.12
Stress	-	-	-	-	0.70
Mood	-	-	-	-	-

Table 3.

Reliability statistics for jump tests; countermovement jump height (CMJ), squat jump height (SJ), drop jump contact time (DJ-CT) drop jump height (DJ-JH) and drop jump reactive strength index (DJ-RSI).

Data are presented as group means (\pm SD) for each trial, a typical error (TE), interclass correlation (ICC), coefficient of variation (CV%), and minimum detectable change (MDC) (75% confidence level).

	Trial 1 (SD)	Trial 2 (SD)	TE (90% CI)	ICC (90% CI)	CV% (90% CI)	MDC
CMJ (cm)	35.3 (4.5)	35.0 (4.0)	1.6 (1.3, 2.3)	0.88 * (0.73, 0.94)	4.8 ** (3.7, 6.9)	2.5
SJ (cm)	34.7 (4.3)	34.4 (3.8)	1.5 (1.2, 2.1)	0.88 * (0.75, 0.95)	4.5 ** (3.5, 6.4)	2.3
DJ-CT (s)	0.20 (0.02)	0.21 (0.02)	0.01 (0.01, 0.01)	0.85 * (0.69, 0.93)	4.9 ** (3.8, 7.0)	0.01
DJ-JH (cm)	29.8 (3.5)	29.9 (3.4)	1.8 (1.4, 2.5)	0.76 (0.51, 0.89)	6.0 ** (4.6, 8.6)	2.8
DJ-RSI (m.s ⁻¹)	1.49 (0.23)	1.43 (0.23)	0.11 (0.08, 0.15)	0.80 * (0.59, 0.91)	7.7 ** (6.0, 11.1)	0.16

An acceptable threshold for reliability was set at 0.80 for ICC and 10% for CV%

* ICC \geq 0.80

** CV% \leq 10%

Table 4.

Summary of reliability statistics for 2, 3 and 4 minutes for PlayerLoad (PL), individual component planes; anterior-posterior (PL_{AP}), mediolateral (PL_{ML}), and vertical (PL_V), and the % contribution of each plane. Data are presented as group means (\pm SD) for each trial, typical error (TE), interclass correlation (ICC), coefficient of variation (CV%), and minimum detectable change (MDC) (75% confidence level).

		Trial 1 (SD)	Trial 2 (SD)	TE (90% CI)	ICC (90% CI)	CV% (90% CI)	MDC
2 min	PL	39.6 (3.8)	39.5 (3.6)	0.9 (0.7, 1.4)	0.95 * (0.87, 0.98)	2.4 ** (1.8, 3.5)	1.4
	PL _{AP}	14.8 (2.3)	14.5 (2.5)	1.1 (0.9, 1.6)	0.81 * (0.59, 0.92)	8.0 ** (6.1, 11.8)	1.7
	PL _{ML}	14.4 (2.0)	13.9 (1.9)	0.7 (0.5, 1.0)	0.89 * (0.74, 0.95)	5.7 ** (4.4, 8.5)	1.1
	PL _V	28.1 (3.1)	28.1 (3.0)	0.9 (0.7, 1.3)	0.93 * (0.84, 0.97)	3.2 ** (2.5, 4.8)	1.3
	% PL _{AP}	25.8% (3.0%)	25.6% (2.7%)	1.8% (1.4%, 2.6%)	0.63 (0.28, 0.83)	7.2 ** (5.5, 10.6)	2.8%
	% PL _{ML}	25.2% (2.8%)	24.6% (3.1%)	1.3% (1.0%, 1.9%)	0.83 * (0.63, 0.93)	5.9 ** (4.5, 8.7)	2.0%
	% PL _V	49.0% (2.8%)	49.8% (2.9%)	1.3% (1.0%, 1.9%)	0.83 * (0.62, 0.93)	2.7 ** (2.0, 3.9)	2.0%
3 min	PL	59.3 (5.8)	59.1 (5.4)	1.5 (1.1, 2.1)	0.94 * (0.86, 0.98)	2.5 ** (1.9, 3.7)	2.2
	PL _{AP}	22.3 (3.2)	21.9 (3.4)	1.4 (1.1, 2.0)	0.85 * (0.66, 0.94)	6.3 ** (4.8, 9.3)	2.1
	PL _{ML}	21.7 (3.1)	21.0 (2.8)	0.9 (0.7, 1.3)	0.93 * (0.82, 0.97)	4.3 ** (3.3, 6.4)	1.3
	PL _V	42.3 (4.8)	42.5 (4.6)	1.5 (1.2, 2.3)	0.91 * (0.79, 0.96)	3.9 ** (2.9, 5.7)	2.3
	% PL _{AP}	25.8% (2.8%)	25.5% (2.3%)	1.4% (1.1%, 2.0%)	0.73 (0.45, 0.88)	5.4 ** (4.2, 8.0)	2.1%
	% PL _{ML}	25.2% (2.8%)	24.7% (3.1%)	1.0% (0.8%, 1.4%)	0.91 * (0.78, 0.96)	4.3 ** (3.3, 6.3)	1.5%
	% PL _V	49.0% (2.7%)	49.8% (2.9%)	1.1% (0.9%, 1.7%)	0.85 * (0.67, 0.94)	2.4 ** (1.8, 3.5)	1.7%
4 min	PL	79.4 (7.9)	78.9 (7.2)	1.7 (1.3, 2.4)	0.96 * (0.90, 0.98)	2.1 ** (1.6, 3.1)	2.5
	PL _{AP}	30.1 (4.3)	29.4 (4.6)	1.7 (1.3, 2.5)	0.87 * (0.71, 0.95)	5.9 ** (4.5, 8.7)	2.6
	PL _{ML}	29.0 (4.0)	28.1 (3.8)	1.2 (0.9, 1.8)	0.92 * (0.80, 0.96)	4.8 ** (3.6, 7.0)	1.8
	PL _V	56.5 (6.4)	57.0 (6.2)	1.9 (1.4, 2.7)	0.93 * (0.83, 0.97)	3.4 ** (2.6, 5.0)	2.8
	% PL _{AP}	26.0% (2.7%)	25.6% (2.4%)	1.3% (1.0%, 2.0%)	0.75 (0.48, 0.89)	5.3 ** (4.1, 7.9)	2.1%
	% PL _{ML}	25.2% (2.8%)	24.6% (3.1%)	1.1% (0.8%, 1.6%)	0.88 * (0.73, 0.95)	4.9 ** (3.7, 7.2)	1.7%
	% PL _V	48.8% (2.8%)	49.8% (3.0%)	1.2% (0.9%, 1.8%)	0.85 * (0.67, 0.94)	2.5 ** (1.9, 3.7)	1.8%

An acceptable threshold for reliability was set at 0.80 for ICC and 10% for CV%

* ICC \geq 0.80

** CV% \leq 10%

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