Comparing optical and electromagnetic tracking systems to facilitate compatibility in sports kinematics data

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ABSTRACT
Electromagnetic (EM) tracking has been used to quantify biomechanical parameters of the lower limb and lumbar spine during ergometer rowing to improve performance and reduce injury. Optical motion capture (OMC) is potentially better suited to measure comprehensive whole-body dynamics in rowing. This study compared accuracy and precision of EM and OMC displacements by simultaneously recording kinematics during rowing trials at low, middle, and high rates on an instrumented ergometer (n=12). Trajectories calculated from OMC and EM sensors attached to the pelvis, lumbar spine, and right leg were highly correlated, but EM tracking lagged behind ergometer and OMC tracking by approximately 6%, yielding large RMS errors. When this phase-lag was corrected by least squares minimization, agreement between systems improved. Both systems demonstrated an ability to adequately track large dynamic compound movements in the sagittal plane but struggled at times to precisely track small displacements and narrow angular ranges in medial/lateral and superior/inferior directions. An OMC based tracking methodology can obtain equivalence with a previously validated EM system, for spine and lower limb metrics. Improvements in speed and consistency of data acquisition with OMC are beneficial for dynamic motion studies. Compatibility ensures continuity by maintaining the ability to compare to prior work.

INTRODUCTION
Optical and electromagnetic motion capture systems are frequently employed to measure whole body kinematics in biomechanics research. These systems each have their own advantages and drawbacks. Optical motion capture (OMC) uses infrared cameras to record the position of reflective markers placed on the body, whereas electromagnetic tracking systems (EM) utilize receivers within an electromagnetic field to compute position and orientation of body segments in space. EM tracking provides numerous advantages in sports and clinical biomechanics, including relatively simple digitization, ability to record and display position and orientation with little data processing, and 6 degree of freedom sensing, without direct line-of-sight requirement between transmitter and receivers (Parent 2012; Franz et al. 2014).

Collaboration between Imperial College and British Rowing has used the ‘Flock of Birds’ (Ascension Technologies, USA) EM tracking system lower extremity and lumbar spine biomechanics research on elite rowers, offering a framework to describe proper sequencing (Bull et al. 1998; Bull and McGregor 2000; McGregor et al. 2004) and establishing links between biomechanics parameters and performance metrics, which continue to be used to analyze quality of elite rowing technique (Buckridge et al. 2016, 2015; Murphy 2009). However, current generation EM systems still present drawbacks, which relate to the functional range of accuracy of the magnetic field, a frame rate affected by the number of receivers (LaScalza et al. 2003), any restricted movement from tethered cable connections causing a hindrance to subjects (Sorrentino et al. 2020), a sensitivity to metal and other electronics (Meyer et al. 2008; Ng et al. 2009), and a smaller number of body segments that can be tracked due to limited number of physical sensors (Murphy et al. 2011). Most EM systems use a digitization and indirect tracking method wherein positions of digitized anatomical landmarks are defined relative to the EM sensors’ local coordinate system and used for building anatomical frames of body segments (Murphy 2009).

Optical motion technology presents advantages for whole-body tracking of dynamic activities in large capture volumes and has been successfully used to analyze posture and overall body movement during ergometer rowing (Pollock et al. 2009; Attenborough et al. 2012;
Skublewska-Paszkowska et al. (2016). Despite a line-of-sight requirement between markers and cameras, OMC overcomes EM system limitations with increased frame rate (up to 240 Hz), direct tracking of anatomical landmarks using reflective markers and offers greater flexibility in marker placement (both individual and clusters). However, because orientation cannot be directly measured, more markers are required than with EM systems, and an OMC approach typically requires longer post-processing time for complex data (Parent 2012; Sorriento et al. 2020).

Advances in research methods and technologies reduce bias, improve precision, and expand measurement options, but at the cost of limiting comparability to data derived using alternative systems (Pueo and Jimenez-olmedo 2017). While OMC systems have become the ‘gold standard’ method in motion capture (Corazza et al. 2010), there are studies using EM systems from a wide range of cohorts, and settings, whose records can be leveraged by developing a standard of interoperability between systems, including Imperial College’s wealth of historical biomechanics data on elite level rowers. This type of compatibility can be successfully used to preserve older data to operate with newer or different systems and facilitates interpretation of current and future OMC data by contextualization in relation to past EM measures (Rowlands et al. 2016).

Hassan et al. (2007) reported direct comparisons between OMC and EM tracking systems when receivers were affixed to a robotic articulating arm, providing an assessment of each system for slow movements with small angular deviations. They found that mean difference between each system and the robotic arm position did not exceed 2° but stressed the need for smoothing and rigid body correction in data processing. That study did not address the inaccuracies native to human subject testing, such as skin motion artifact, for which such rigid body corrections may be inappropriate. Few in-vivo studies have compared the performance of these systems for compound motions under dynamic conditions. Lugade et al. (2015) examined intra- and inter-day repeatability of EM and OMC measurements during a sit-to-stand task, finding high coefficients of multiple correlation (CMC) for sagittal plane hip, knee, and ankle joint angles and no differences between systems in joint range of motion (Lugade et al. 2015). There are no publications that assess these technologies in more complex, higher speed, whole-body motions, such as rowing.

The aims of this study were to analyze the accuracy and precision of EM and OMC displacements in three dimensions during ergometer rowing, quantifying the relative error in reported position of directly tracked receivers, digitized anatomical landmarks, and calculated joint centers.

Methods

Twelve healthy subjects with at least 2 years of rowing experience participated in the study (11 female/1 male; age: 25.6 ± 2.2 years; height: 178.6 ± 7.8 cm; mass: 74.8 ± 6.0 kg). All subjects were club or national team athletes rowing regularly at the time of the study. Imperial College London research ethics committee granted approval, written informed consent was obtained from each participant prior to testing, and all athlete data was anonymized.

Apparatus

External kinetics were recorded on a bespoke instrumented ergometer with load cells at the handle, seat, and footplates and a rotary encoder on the flywheel (Murphy 2009; Buckeridge et al. 2012). Kinematic data was simultaneously recorded with a ten-camera optical motion tracking system, operating at 100 Hz (‘MX T-series’, Vicon, Oxford, UK), and a four receiver ‘extended range’ EM tracking system operating at 75 Hz (‘Flock of Birds’, Ascension Technologies, VT, USA). The lab was arranged for a large capture volume (36 m³). The EM system transmitter was located 1 m to the right of the ergometer slide rail and 1.25 m above the floor on a wooden tower (Figure 1). OMC data was streamed to Vicon Nexus 1.8.5 software, while ergometer outputs and EM data were streamed to a custom data acquisition program (LabView 2016, National Instruments, TX, USA). Real-time feedback was displayed to athletes and researchers during testing (McGregor et al. 2016).

Protocol

Athlete tracking

OMC markers and EM receivers were attached to the pelvis, lumbar spine, and right leg of each subject (Figure 1). Four EM receivers were attached to the skin at the thoracolumbar (T12/L1) and lumbosacral junction (L5/S1) using adhesive pads and midway along the thigh and shank using fabric straps. EM system calibration followed the procedure previously described (Murphy et al. 2011), which involves digitizing relevant bony landmarks by palpation with an EM receiver attached to a stylus. Nine OMC markers were placed on the digitized bony landmarks as well as atop all four EM receivers and the ergometer handle (Figure 1). Digitization and marker placement was performed by a single operator, an experienced sports biomechanist with training in landmark palpation, skin marker placement and extensive experience with methods used in the current protocol.
Each rower performed three rowing trials at low, medium, and high stroke rates (18, 24, 28 strokes per minute, spm) on an instrumented ergometer (Urbanczyk et al. 2019, 2020). Real-time feedback from the ergometer monitor to the athlete facilitated maintaining a fixed stroke rate. Participants could rest as needed between each trial. Knee and ankle joint centers were defined as the midpoint between lateral and medial epicondyles and malleoli, respectively. Right hip joint center was determined from functional calibration (Camomilla et al. 2006).

**Apparatus tracking**

Three subjects participated in a secondary protocol, which involved tracking the ergometer apparatus under static and dynamic conditions but did not include marker placement on the athletes themselves. OMC and EM systems simultaneously recorded displacement of the ergometer handle and the ergometer seat – (left and right sides; Figure 1). OMC and EM measurements were compared to ergometer positions measured by the rotary encoder and a precision measuring tape during dynamic and static tracking, respectively.

**Data analysis and statistics**

Synchronized motion data and external force data were processed in MATLAB (2017, MathWorks, MA, USA). Continuous rowing trials were smoothed using a 4th order low-pass Butterworth filter with a 6-Hz cutoff frequency (Pollock et al. 2009) and divided into individual strokes where the start was identified as the minimum anterior/posterior (A/P) handle position. The catch was defined as the onset of handle force exceeding 75 N, with a steep, increasing slope. The release was defined as maximum A/P handle displacement. Each stroke was time normalized from 0–100% of completion using a cubic spline interpolation, such that the drive time was from start (0%) to release, and the recovery time was from release to a subsequent next start (100%) (Urbanczyk et al. 2019, 2020). All measurements shared a global coordinate frame, and each trial was referenced to initial marker/receiver position, permitting comparison of system drift between and within subjects. Statistics were run in RStudio (RStudio Team 2016) with a significance level of $\alpha = 0.05$. Initial ANOVA analysis compared reported sensor position between stroke rates for each system; however, no significant difference was found as stroke rate increased. Therefore, data was pooled to reduce statistical complexity, with mean
tracked position and 95% confidence intervals calculated across all speeds. Bland-Altman analysis was used to compare the bias and limits of agreement of each motion tracking system. Root mean square error (RMSE) was used to quantify the difference in each system’s tracked position to ground truth position during static testing. Geers metric (GM) was used to discriminate phase and magnitude differences in tracked position between systems during dynamic testing (Schwer 2007), where metric values of zero indicate perfect agreement, less than 0.30 difference is considered good and greater than 0.50 difference is considered poor (Schwer 2007; Klement et al. 2019). Analysis of EM system lag and its correction used a method of least squares fitting, which retrieved the wavefront phase shift of the EM receivers relative to the OMC and ergometer systems. For every participant, and for each continuous 3-minute rowing trial, a single time-shift translational offset was calculated and applied to the EM system kinematic data as determined by the residuals of the minimization.

Results

Apparatus tracking: data consistency and capture latency

Static tracking of the handle and seat show that differences for both tracking systems, with respect to ground truth position, were similar (Figure 2). Average RMS error for the OMC system was 0.08 ± 0.04 mm, while EM system error was 1.44 ± 2.17 mm. Comparing variance of static measurements, the OMC system was much less prone to noise ($p < 0.001$), with 95% confidence intervals for the OMC system lying well within the limits of the EM system (Figures 2, 3). Under dynamic motion tracking, comparison of stroke trajectories captured by both tracking systems and the ergometer indicates that the lower frame rate and burst transmission of the EM system contribute to jagged tracking of smooth movements (Figure 3).

Seat width, calculated from the distance between tracking markers/receivers on left and right sides of the ergometer seat, could be directly compared to the true measured seat width (315.0 mm), providing an estimate of system accuracy in the medial/lateral (M/L) direction (Figure 4). The OMC system slightly overestimates seat width in both static (315.6 ± 0.5 mm; RMS = 0.62 mm) and dynamic trials (316.1 ± 0.3 mm; RMS = 1.16 mm), while the EM system was less consistent during dynamic trials (314.1 ± 2.8 mm; RMS = 2.69 mm), underestimating seat width near the catch and release positions, where the receivers were radially farther from the transmitter and overestimating seat width when the receivers were closer to the transmitter during mid-slide (Figure 4). In static tracking, receiver position reported by the EM system has large errors and between-trial variance (290.3 ± 23.4 mm). At slide positions farther from the transmitter, seat width was underestimated (catch: 285.2 ± 0.3 mm, release: 260.4 ± 6.9 mm), but for mid-slide position closer to the transmitter, estimates were more accurate (mid-slide: 314.1 ± 1.8 mm).

Differences in A/P handle motion during dynamic apparatus tracking from each system were compared to handle position output measured using the rotary encoder directly on the ergometer. The EM showed significant differences in A/P handle position from both other systems (RMS = 136.2 mm; $p < 0.05$). The EM system lags both ergometer and OMC reported position by ~6% per stroke (Figure 3). There was no significant difference in measured A/P handle position between OMC and ergometer systems (RMS = 34.2 mm; $p = 0.93$). This apparent capture latency in the EM system tracking impacts time-related kinetic and kinematic relationships and any correlations or conclusions drawn therefrom. This capture latency could be corrected for by least squares minimization, compared

![Figure 2](image-url)  
**Figure 2.** Difference from ground truth position of static marker/receivers in A/P plane for OMC and EM systems. Similar patterns were found for M/L and S/I directions.
to the known ergometer output positions. After latency correction, quality of fit between the EM system, the OMC system, and the ergometer improved substantially, with RMS error reduced from 136.4 mm to 34.4 mm (Figure 3).

**Athlete tracking: co-localized markers and joint centre estimates**

GM values (Table 1) across co-localized marker/sensor displacements (Figure 5) and predicted joint centre trajectories (Figure 6) were highly related in A/P direction; however, agreement overall was poorest in M/L direction, with large differences attributed to magnitude. Deviations attributable to phase differences can be substantially reduced by applying phase-lag correction to the EM data, and metric values were good for most trajectories (Table 1 – Phase; GM: moderate-to-good), expect for S/I hip joint
center, for which GM attributes all deviation to phase effects, despite lag correction. Deviations attributable to magnitude differences show some improvement with phase-lag correction, with good agreement overall in A/P and S/I directions (Table 1 – Magnitude; GM: moderate-to-good). However, in the M/L direction, hip and ankle magnitudes improve but were still poor overall, while M/L femur and knee worsen after correction (GM: moderate-to-poor). Some of these deviations may be confounded by large confidence intervals present in M/L knee joint center (Figure 6), femur receiver, and tibia receiver (Figure 5).

Rowing is predominately defined by motion in the A/P direction, with large displacements and smaller relative variance yielding a high signal-to-noise ratio. Displacement ranges in M/L and S/I directions were smaller with larger relative standard deviations (Figures 5, 6). Caution is warranted when analyzing small displacements and angular rotations. OMC is more sensitive to small perturbations in marker position, and lower resolution makes EM less sensitive to true shifts.

**Discussion**

This study assessed the ability of OMC and EM systems to represent ground truth position in static and dynamic apparatus tracking and the precision and accuracy of those systems when tracking volunteers performing
whole-body ergometer rowing. While previous work has compared simple, well-defined movements, tracking systems are increasingly utilized in highly sensitive and complicated performance analyses (Koivukangas et al. 2013; Pueo and Jimenez-Olmedo 2017) where accuracy and precision are paramount.

Accuracy in OMC systems is dependent on appropriate positioning of cameras to observe marker movement and on the quality of calibration of the capture volume. Accuracy in EM systems is dependent on the distance between transmitter and receivers, number of receivers in use, and ferromagnetic interference in the capture volume. Similar studies comparing the accuracy of EM and OMC systems have been conducted utilizing markers/receivers applied to a robotic, articulated arm where the true range of motion was known (Hassan et al. 2007; Lugade et al. 2015). Hassan et al. (2007) concluded that accuracy of EM and OMC systems were comparable for measuring simulated upper extremity kinematics when appropriate post-hoc filtering and corrections were applied.

The average bias, or discrepancy between tracking methods, was sufficient to be important for assessing motion in multiple planes. If the limits of agreement had been narrow and the bias small, the two methods could be considered essentially equivalent; however, this was not the case between the OMC and EM systems tested here. Examining A/P handle position by Bland-Altman analysis (Figure 7), before phase-lag correction (OMC-EM), there was a clear cyclic trend, where the absolute
value of bias increases as mean stroke length increases (mean bias (limits) = 28.1 (−220.3, 276.6) mm). After correcting for the EM system capture latency (OMC-CorEM), this cyclic trend disappears, the limits of agreement contract (mean bias (limits) = 2.2 (−27.3, 31.7) mm) and the variance around the bias line become consistently flat (Figure 7).

After phase-lag correction of the EM system data, the sensor positions reported by both systems tended to fall within 95% confidence interval of each other, suggesting that either system is sufficiently accurate when the signal-to-noise ratio is high, as was the case for large A/P movements (Figures 5, 6). Both systems tended to have difficulty accurately tracking smaller M/L movements (Figure 6). A similar magnitude effect was reported by Hassan et al. (2007) where deviations from the ‘ground truth’ articulated robotic arm position occurred mostly at small angles and displacements. The authors did not assess repeatability of the tracking systems using an in vivo model. Lugade et al. (2015) reported sagittal plane lower limb joint angles during a sit-to-stand activity, finding that OMC had lower RMS errors and higher intra- and inter-day CMC values (0.99 ± 0.001; 0.97 ± .025) than EM (0.97 ± 0.05; 0.94 ± 0.038) but concluded that both systems were adequate for tracking dynamic motion (Lugade et al. 2015).

Assessment of accuracy and precision of EM systems in the context of rowing kinematics have been reported (Bull and McGregor 2000; Ng et al. 2009). Ng et al. (2009) limited their study to the assessment of a static multi-hinge model being moved with constant rate (22 spm) along an ergometer slide rail but suggested that different velocities may elicit variations in error. While sequential data transmission of EM receivers could be detrimental to data acquisition of higher rate movements, this does not appear to have affected the current study. Increasing from 18 spm to 28 spm did not translate into changes in mean position differences, nor in the data variance. Ng et al. (2009) also noted that of the four receivers employed during their experiments, the receiver located farthest from the transmitter produced the largest angular error, while the receiver closest to the transmitter produced the smallest angular error. This is consistent with results from the current study’s variability in ergometer seat width as receivers were moved along the slide rail past the transmitter, with estimates deviating more from the true value as the radial distance of each receiver to the transmitter increased (Figure 4).

Bull and McGregor (2000) demonstrated that an EM device was sufficiently accurate to discern differences in rowing technique and movement patterns in the sagittal plane, with reported spinal parameters used to discriminate between qualitatively ‘good’ and ‘bad’ rowing technique. Results from the current study appear to support the conclusion that the EM system is capable of discerning salient changes in A/P trajectories (Figure 5), due to displacements in the rowing stroke being very large in the A/P direction compared to the resolution of the receivers. M/L and S/I displacements during rowing were relatively small, and many of the Geers metric magnitude values were moderate-to-poor in the M/L direction. Given reported difficulties in analyzing small displacements and angular rotations using an EM system (Hassan et al. 2007), discriminating differences in spinal torsion and lateral bending may not be feasible, and re-examining published spinal flexion/extension values may be future work.

![Figure 7. Bland-Altman comparison of bias and limits of agreement (mean ± 95% CI) during dynamic apparatus tracking for A/P handle displacements between OMC and EM systems before (OMC-EM) and after (OMC-CorEM) phase-lag correction. After phase-lag correction the cyclic trend collapses and the limits of agreement contract.](image-url)
Limitations

The left seat EM receiver was thus farther from the EM transmitter than the right seat EM receiver. Despite a published functional radius of up to 1.5 m for the EM system, tracking of movement near the catch and the release positions may have been affected by the greater radial distance from the transmitter, contributing to the discrepancy in A/P seat position between left and right receivers and in greater error estimating seat width. Ng et al. (2009) noted that differences in ferrous material content in different parts of the ergometer may contribute to variability of the data.

Anatomical variation among participants, biomechanical differences in performing the rowing task, and skin motion artifact could also have affected results for both systems. Digitization differences at the start of testing were minimized by using one well-trained operator for marker placement, who was an experienced sports biomechanist and physiotherapist with substantial training in landmark palpation, skin marker placement and specific experience with methods used in the current protocol. However, OMC markers and EM receivers can shift during testing, introducing error in anatomical landmark position. Intrinsic kinematics of rowing regularly cause occlusion of OMC markers fixed to the anterior pelvis, and relying on digital reconstruction may be a source of error.

Conclusions

This study specifically compared the dynamic accuracy and precision of EM and OMC displacements in three dimensions during ergometer rowing by simultaneous motion tracking. While both EM and OMC systems demonstrated an ability to adequately track large dynamic compound movements in the sagittal plane during ergometer rowing, both systems struggled to precisely track small displacements in M/L and S/I directions. OMC outperformed EM tracking for salient measurements of precision, speed, and consistency in data acquisition. Expanded capture capabilities with OMC are beneficial for whole-body motion studies and justify further use of OMC without losing the ability to compare to prior work, as system compatibility facilitates interpretation of future OMC data in relation to EM measures.

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No potential conflict of interest was reported by the author(s).

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References

Attenborough AS, Smith RM, Sinclair PJ. 2012. Effect of gender and stroke rate on joint power characteristics of the upper extremity during simulated rowing. J Sports Sci. 30:449–458. doi:10.1080/02640414.2011.616949.
Buckeridge E, Hislop S, Bull A, McGregor A. 2012. Kinematic asymmetries of the lower limbs during ergometer rowing. Med Sci Sports Exerc. 44:2147–2153. doi:10.1249/MSS.0b013e3182625231.
Buckeridge EM, Bull AMJ, Mcgrregor AH. 2015. Biomechanical determinants of elite rowing technique and performance. Scand J Med Sci Sport. 25:e176–e183. doi:10.1111/sms.12264.
Buckeridge EM, Bull AMJ, McGregor AH. 2016. Incremental training intensities increases loads on the lower back of elite female rowers. J Sports Sci. 34:369–378. doi:10.1080/02640414.2015.1056821.
Bull AM, Berkshire FH, Amis AA. 1998. Accuracy of an electromagnetic measurement device and application to the measurement and description of knee joint motion. Proc Inst Mech Eng 212:347–355. doi:10.1243/0954411981534123.
Bull AMJ, McGregor AH. 2000. Measuring spinal motion in rowers: the use of an electromagnetic device. Clin Biomech. 15:772–776. doi:10.1016/S0268-0033(00)00043-7.
Camomilla V, Cereatti A, Vannozzi G, Cappozzo A. 2006. An optimized protocol for hip joint centre determination using the functional method. J Biomech. 39:1096–1106. doi:10.1016/j.jbiomech.2005.02.008.
Corazza S, Mündermann L, Gambaretto E, Ferrigno G, Andriacchi TP. 2010. Markerless motion capture through visual hull, articulated ICP and subject specific model generation. Int J Comput Vis. 87:156–169. doi:10.1007/s11263-009-0284-3.
Franz AM, Haidegger T, Birkfellner W, Cleary K, Peters TM, Maier-Hein L. 2014. Electromagnetic tracking in medicine - A review of technology, validation, and applications. IEEE Trans Med Imaging. 33:1702–1725. doi:10.1109/TMI.2014.2321777.
Hassan EA, Jenkyn TR, Dunning CE. 2007. Direct comparison of kinematic data collected using an electromagnetic tracking system versus a digital optical system. J Biomech. 40:930–935. doi:10.1016/j.jbiomech.2006.03.019.
Klement C, Nolte D, Ding Z, Rane L, Quest RA, Finnegan ME, Walker M, Reilly P, Bull AMJ. 2019. Anthropometric scaling of anatomical datasets for subject-specific musculoskeletal modelling of the shoulder. Ann Biomed Eng. 47:924–936. doi:10.1007/s10439-019-02207-2.
Koivukangas T, Katisko JP, & Koivukangas JP. 2013. Technical accuracy of optical and the electromagnetic tracking systems. SpringerPlus. 2:90. doi:10.1186/2193–1801–2–90.
LaScalza S, Arico J, Hughes R. 2003. Effect of metal and sampling rate on accuracy of Flock of Birds electromagnetic tracking system. J Biomech. 36:141–144. doi:10.1016/S0021-9290(02)00322-6.

Lugade V, Chen T, Erickson C, Fujimoto M, Juan JGS, Karduna A, Chou L-S. 2015. Comparison of an electromagnetic and optical system during dynamic motion. Biomed Eng Appl Basis Commun. 27:1550041. doi:10.10017/s0106237215500416.

McGregor AH, Buckeridge E, Murphy AJ, Bull AM. 2016. Communicating and using biomechanical measures through visual cues to optimise safe and effective rowing. Proc Inst Mech Eng Part P J Sport Eng Technol. 230:246–252. doi:10.1177/1754337115618552.

McGregor AH, Bull AMJ, Byng-Maddick R. 2004. A comparison of rowing technique at different stroke rates: a description of sequencing, force production and kinematics. Int J Sports Med. 25:465–470. doi:10.1055/s-2004-820936.

Meyer KE, Saether EE, Soiney EK, Shebeck MS, Paddock KL, Ludewig PM. 2008. Three-dimensional scapular kinematics during the throwing motion. J Appl Biomech. 24:24–34. doi:10.1123/jab.24.1.24.

Murphy AJ. 2009. Elite rowing: technique and performance. [PhD dissertation] London: Imperial College London, p. 289.

Murphy AJ, Bull AMJ, McGregor AH. 2011. Optimizing and validating an electromagnetic tracker in a human performance laboratory. Proc Inst Mech Eng Part H J Eng Med. 225:343–351. doi:10.1177/2041303310393231.

Ng L, Burnett A, Campbell A, O'Sullivan P. 2009. Caution: the use of an electromagnetic device to measure trunk kinematics on rowing ergometers. Sport Biomech. 8:255–259. doi:10.1080/14763140903229492.

Parent R. 2012. Motion capture. In: Parent RBT-CA, editor. Computer animation. 3rd ed. Boston: Morgan Kaufmann; p. 187–198. doi:10.1016/B978-0-12-415842-9.00006-X.

Pollock CL, Jenkyn TR, Jones IC, Ivanova TD, Garland SJ. 2009. Electromyography and kinematics of the trunk during rowing in elite female rowers. Med Sci Sports Exerc. 41:628–636. doi:10.1249/MSS.0b013e31818c1300.

Pueo B, Jimenez-Olmedo JM. 2017. Application of motion capture technology for sport performance analysis (El uso de la tecnología de captura de movimiento para el análisis del rendimiento deportivo). Retos. 2041:241–247. doi:10.47197/retos.v0i32.56072.

Rowlands AV, Cliff DP, Fairclough SJ, Boddy LM, Olds TS, Parfitt G, Noonan RJ, Downs SJ, Knowles ZR, Beets MW. 2016. Moving forward with backward compatibility: translating wrist accelerometer data. Med Sci Sports Exerc. 48:2142–2149. doi:10.1249/MSS.000000000001015.

RStudio Team, 2016. RStudio: integrated Development for R, 1. 1.456.ed. RStudio, Inc., Boston.

Schwer LE. 2007. Validation metrics for response histories: perspectives and case studies. Eng Comput. 23:295–309. doi:10.1007/s00366-007-0070-1.

Skublewska-Paszkowska M, Montusiewicz J, Łukasik E, Pszczola-Pasierbiewicz I, Baran K, Smolka J, Pueo B. 2016. Motion capture as a modern technology for analysing ergometer rowing. Adv Sci Technol Res J. 10:132–140. doi:10.12913/22998624/61941.

Sorriento A, Porfido MB, Mazzoleni S, Calvosa G, Tenucci M, Ciuti G, Dario P. 2020. Optical and electromagnetic tracking systems for biomedical applications: a critical review on potentialities and limitations. IEEE Rev Biomed Eng. 13:212–232. doi:10.1109/RBME.2019.2939091.

Urbanczyk CA, Mcgregor AH, Bull AMJ. 2019. Modelling scapular biomechanics to enhance interpretation of kinematics and performance data in rowing. ISBS Conf Proc Arch. 37(1):133–136. https://commons.nmu.edu/isbs/vol37/iss1/31

Urbanczyk CA, Miller E, Mcgregor AH, Bull AMJ. 2020. Fatigue leads to altered spinal kinematics during high performance ergometer rowing. ISBS Conf Proc Arch. 38 (1):256–259. https://commons.nmu.edu/isbs/vol38/iss1/66