

Exploring wearable sensors as an alternative to marker-based motion capture in the pitching delivery

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Background. Technological advancements in Inertial Measurement Unit (IMU) sensors have enabled high precision in the measurement of joint angles and acceleration on human subjects. This has resulted in new devices that reportedly measure joint angles, arm speed, and stresses of baseball players. This study seeks to validate one such sensor, the MotusBASEBALL unit, with a marker-based motion capture laboratory.

Hypothesis. We hypothesize that the joint angle measurements (“Arm Slot” and “Shoulder Rotation”) of the device will be accurate and reliable, but that the “Arm Speed” and “Stress” metrics will not be accurate due to limitations in IMU technology.

Methods. 10 healthy subjects threw 5-7 fastballs followed by 5-7 breaking pitches (slider or curveball) in the motion capture lab. Subjects will be wearing retroreflective markers and the MotusBASEBALL sensor simultaneously.

Results. It was found that the Arm Slot ($P < 0.001$), Shoulder Rotation ($P < 0.001$), and Stress ($P = 0.001$ when compared to elbow torque, $P = 0.002$ when compared to shoulder torque) measurements were all significantly correlated with the results from the motion capture lab. Arm Speed showed significant correlations to shoulder internal rotation speed ($P = 0.001$) and shoulder velocity magnitude ($P = 0.002$). For the entire test population, Arm Slot and Shoulder Rotation measurements were very close to measurements from the motion capture test, averaging 8 degrees less and 9 degrees less respectively. Arm Speed had a much larger difference, averaging 3745 deg/s lower than shoulder internal rotation velocity, and 3891 deg/s less than the shoulder velocity magnitude. The Stress metric was found to be 41 Nm less when compared to elbow torque, and 42 Nm less when compared to shoulder torque. Despite the differences in magnitude, the correlations were extremely strong, indicating that the MotusBASEBALL sensor could be valid for casual use.

Conclusions. This study validates the use of the MotusBASEBALL for future studies that look at the Arm Slot, Shoulder Rotation, Arm Speed, and Stress measurements from the MotusBASEBALL sensor. Excepting elbow extension velocity, all metrics from the MotusBASEBALL unit showed significant correlations to their corresponding metrics from motion capture. While magnitudes differ significantly in the Arm Speed and Stress metrics between MotusBASEBALL and the motion capture lab, the link between the metrics is strong enough to indicate valid casual use. Further research should be done to further investigate the validity and reliability of the Arm Speed metric.

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31 **Abstract**

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33 enabled high precision in the measurement of joint angles and acceleration on human subjects.
34 This has resulted in new devices that reportedly measure joint angles, arm speed, and stresses of
35 baseball players. This study seeks to validate one such sensor, the MotusBASEBALL unit, with
36 a marker-based motion capture laboratory.

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38 Rotation”) of the device will be accurate and reliable, but that the “Arm Speed” and “Stress”
39 metrics will not be accurate due to limitations in IMU technology.

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41 curveball) in the motion capture lab. Subjects will be wearing retroreflective markers and the
42 MotusBASEBALL sensor simultaneously.

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44 $= 0.001$ when compared to elbow torque, $P = 0.002$ when compared to shoulder torque)
45 measurements were all significantly correlated with the results from the motion capture lab. Arm
46 Speed showed significant correlations to shoulder internal rotation speed ($P = 0.001$) and
47 shoulder velocity magnitude ($P = 0.002$). For the entire test population, Arm Slot and Shoulder
48 Rotation measurements were very close to measurements from the motion capture test, averaging
49 8 degrees less and 9 degrees less respectively. Arm Speed had a much larger difference,
50 averaging 3745 deg/s lower than shoulder internal rotation velocity, and 3891 deg/s less than the
51 shoulder velocity magnitude. The Stress metric was found to be 41 Nm less when compared to
52 elbow torque, and 42 Nm less when compared to shoulder torque. Despite the differences in
53 magnitude, the correlations were extremely strong, indicating that the MotusBASEBALL sensor
54 could be valid for casual use.

55 **Conclusions.** This study validates the use of the MotusBASEBALL for future studies that look
56 at the Arm Slot, Shoulder Rotation, Arm Speed, and Stress measurements from the
57 MotusBASEBALL sensor. Excepting elbow extension velocity, all metrics from the
58 MotusBASEBALL unit showed significant correlations to their corresponding metrics from
59 motion capture. While magnitudes differ significantly in the Arm Speed and Stress metrics
60 between MotusBASEBALL and the motion capture lab, the link between the metrics is strong
61 enough to indicate valid casual use. Further research should be done to further investigate the
62 validity and reliability of the Arm Speed metric.

63

64 **Introduction**

65 Marker-based motion capture has been shown in research to be suitable in measuring the
66 kinematics and kinetics of a baseball throw (Richards, 1999). The OptiTrack camera system
67 (Natural Motion / OptiTrack; Corvallis, Oregon) used in this study has also been shown in
68 research to be comparable to other high-end motion capture systems (Thewlis et al., 2013).

69 Inertial Measurement Unit (IMU) sensors have been validated in research for joint angle
70 measurements (Leardini et al., 2014), gait analysis (Kavanagh and Menz, 2008), kinematics of
71 runners (Provot et al., 2017), as well as swimming biomechanics (de Magalhaes et al., 2014).
72 IMU sensors have started to gain popularity in measuring the kinematics of throwers, but
73 validation of such sensors has been limited.

74 One study used IMU sensors to measure kinematics of youth throwers, but the study focused
75 primarily on pelvis and torso rotation; the sensor attached to the wrist was only used to identify
76 the timing of the acceleration phase of the throwing motion (Grimpampi et al., 2016).

77 The MotusBASEBALL unit (Motus; New York, NY) is a popular IMU sensor that purports to
78 measure the biomechanics of a thrower's elbow. The only existing validation of the unit comes
79 from Camp et al., 2017, which states that the MotusBASEBALL sensor was evaluated
80 simultaneously with an 8-camera motion capture system. Correlation coefficients (r-values)
81 between measurements with the 2 systems were found to be "good to excellent" for all
82 measurements, though no supplemental data was provided. Following studies have used the
83 MotusBASEBALL unit to look at elbow torque and other parameters in pitchers throwing
84 fastballs and offspeed pitches as well (Makhni et al., 2018)

85 A case study performed by Motus showed the MotusONE, a multi-unit IMU device, to be
86 comparable to multi-camera motion capture setups, but the study only had one subject, and to
87 our knowledge was not published in any peer reviewed journal.

88 The purpose of this study is to compare the outputs of the MotusBASEBALL sensor, which are
89 Arm Speed, Arm Slot, Shoulder Rotation, and Stress, against the OptiTrack motion capture
90 system.

91 **Methods**

92 Ten healthy pitchers were selected to participate in the study, nine threw overhead and one threw
93 sidearm and all were right-handed. Participants were provided a verbal explanation of the study
94 and its risks and were asked to read and sign an Informed Consent document before testing.
95 Testing only proceeded once investigators received verbal confirmation and obtained a witnessed
96 legal signature from the athlete.

97 Hummingbird IRB approved the study and granted ethical approval to carry out the data
98 collection at the author's facilities (Hummingbird IRB #: **2018-10**).

99 Heights, weights, and ages of the participants was recorded before the beginning of testing.
100 (Table 1)

101

102 [Table 1]

103

104 *Testing Procedure*

105 Athletes were given as much time as necessary to prepare and warm-up to throw off of the
106 pitching mound. Once ready, pitchers were fitted with reflective markers in preparation for the
107 motion capture test. Forty-seven reflective markers were attached bilaterally on the third distal
108 phalanx, lateral and medial malleolus, calcaneus, tibia, lateral and medial femoral epicondyle,
109 femur, anterior and posterior iliac spine, iliac crest, acromial joint, midpoint of the humerus,
110 lateral and medial humeral epicondyle, midpoint of the ulna, radial styloid, ulnar styloid, distal
111 end of index metacarpal, parietal bone, and frontal bone, as well as on the inferior angle of
112 scapula, C7 and T10 vertebrae, the sternal end of the clavicle, and the xiphoid process.

113 The motion capture system was calibrated using Motive:Body software (Natural Motion /
114 OptiTrack; Corvallis, Oregon) and the ground plane was set; the system typically showed 1mm
115 or less of mean three-dimensional error, and never exceeded 2mm.

116 The MotusBASEBALL sensor, which is typically applied using a sleeve, was fixed to the athlete
117 along the medial throwing elbow using double sided skin-tape to avoid the sleeve causing
118 interference with any of the markers. (Figure 1)

119

120 [Figure 1]

121

122 Pitchers then threw 5-7 fastballs, followed by 5-7 offspeed pitches, either a curveball or a slider,
123 with approximately 30-60 seconds of rest in between throws. All pitches were made at a medium
124 effort level. Research has shown that offspeed pitches may result in significant changes to
125 kinetics and kinematics (Escamilla et al., 2017; Fleisig et al., 2006). For this reason athletes were
126 asked to throw their preferred offspeed pitch. Fatigue was assumed to be negligible with such a
127 low pitch count.

128 Throws were made using a 5-oz. (142g) regulation baseball off the mound to a strike zone target
129 (Oates Specialties, LLC, Huntsville, TX) located above home plate, which was 60' 6'' (18.4 m)
130 away. Testing concluded when the investigators were satisfied they had at least five valid motion
131 capture takes of each pitch type for analysis.

132 For each trial, ball velocity was measured by a Doppler radar gun (Applied Concepts; Stalker
133 Radar, Richardson, Texas). Additionally, for all trials, the three-dimensional motions of the
134 reflective markers were tracked with a multi-camera motion-capture system, sampling at 240 Hz
135 (Natural Motion / Optitrack, Corvallis, Oregon). This motion-capture system contained a mixture
136 of Prime 13 and Prime 13W cameras, totaling 15 cameras. These cameras were placed
137 symmetrically around the capture volume, approximately 8-12 feet from the center of the

138 pitching mound at varying heights. A total of 6 cameras were mounted on a truss system in front
139 of the pitcher to avoid collisions. (Figure 2)

140

141 [Figure 2]

142

143 In total, 9 kinematic and kinetic values (3 position, 4 velocity, and 2 kinetic) were calculated
144 using personal code based on Fleisig methods (Fleisig et al., 2017) in Visual3D (C-Motion Inc.,
145 Germantown MD). Marker position data was filtered using a 20-Hz Butterworth low-pass filter.
146 The mean values for all variables were calculated for each participant based upon their 5 best
147 throws (Escamilla et al., 1998).

148 Three position values were found at ball release (BR): trunk lateral tilt, shoulder abduction, and
149 maximum shoulder external rotation. Measurements were taken as their local joint angles
150 measured in degrees.

151 The four velocity parameters were taken as the maximum speeds of shoulder internal rotation,
152 shoulder abduction, shoulder horizontal abduction, and elbow angular extension. All velocities
153 were calculated as the rate of change in the joint angle, measured in degrees/second.

154 The two kinetic values calculated were elbow varus torque and shoulder internal rotation torque,
155 which were measured in Newton meters (Nm).

156 All MotusBASEBALL data was collected with an iPhone (Apple Inc., Cupertino CA) and the
157 supplied app, which was then manually transferred into labeled spreadsheets for storage and later
158 analysis.

159 *Statistical Analysis*

160 The differences in metrics were analyzed as both a total population of twenty (20) pitches and
161 two separate equal-sized groups controlled for the type of pitch: Fastballs and off-speed pitches.
162 To account for any differences in the scale of magnitude between the motion capture
163 measurements and MotusBASEBALL measurements, the statistical analyses centered on a
164 correlation test based around Pearson's product moment of correlation coefficient and an n-2
165 number of degrees of freedom. The correlation test was used to test the hypothesis of a linear
166 relationship between the two metrics, as opposed to its common statistical compatriot of a T-test,
167 which would rather test for differences in the means of the two metrics.

168 In order to compare measurements from the motion capture trial to the MotusBASEBALL
169 metrics, additional calculations were done. Corrections to the metrics were done following
170 Motus's guidelines which were sent to us via email by representatives from Motus; those
171 corrections follow below.

172 Arm Slot was taken as the sum of the lateral trunk tilt and shoulder abduction at BR. Shoulder
173 Rotation was measured as the maximum amount of shoulder external rotation measured in the
174 global coordinate system. MotusBASEBALL's Arm Speed metric, which was taken from the

175 MotusTHROW app, was compared to elbow extension velocity and shoulder internal rotation
176 velocity, which are the most common standards for measuring arm speed. Per Motus's
177 recommendation, Arm Speed was also compared to the magnitude of the shoulder angular
178 velocities, that is, the square of of the sum of the squares of shoulder abduction velocity, ω_{Sa} ,
179 shoulder horizontal abduction velocity, ω_{Sha} , and shoulder internal rotation velocity, ω_{Sir} .
180 $\sqrt{\omega_{Sa} + \omega_{Sha} + \omega_{Sir}}$.

181 MotusBASEBALL stress was compared to elbow varus torque and shoulder internal rotation
182 torque, which are the two most commonly addressed kinetic markers in pitching research. All
183 torque metrics were in Nm.

184 First, the descriptive metrics (means and standard errors of means) for the holistic group and
185 subgroups for all the marker-based biomechanics measurements and MotusBASEBALL
186 measurements were outlined and recorded.

187 Then these metrics were matched together across paired results (each subject having been
188 recorded on the two separate systems), and had both their Pearson correlation coefficient ρ taken,
189 as well as their p-value, following a Student's T test distribution. The correlation test posits the
190 hypothesis of there being a significant linear association versus the null hypothesis of there being
191 no correlation, or $\rho = 0$.

192 **Results**

193 The results for the three separate groups are displayed in Tables 2 and 3:

194

195 [Table 2]

196

197 [Table 3]

198

199 As is somewhat intuitive given the nature of the more similar sub-populations, the correlation
200 coefficient covers a higher proportion of variability (R^2) within said smaller groups, while also
201 simultaneously recording higher p-values, due to the smaller sample sizes and subsequent
202 degrees of freedom. Interestingly, the fastball group had more significant p-values for the Arm
203 Speed metrics comparisons and the off-speed group had more significant p-values for the Stress
204 metrics.

205 Following a default alpha value of 0.05, the fastball group found significant associations between
206 four of the metrics (Arm Slot, Shoulder Rotation, and the second and third Arm Speed metrics),
207 while the off-speed group found significant associations between six metrics (Arm Slot,
208 Shoulder Rotation, the second and third Arm Speed metrics, and both Stress metrics).

209 Given the small sample size circumstance for these tests, a prudent power analysis on the
210 strength of the correlation was also performed. While a typical low p-value cut-off limits the

211 chances of making a Type I error, or a false positive, a statistical power analysis is performed to
212 limit the chances of making a Type II error, or a false negative (also referred to as beta). The
213 magnitude of the power is $(1 - P(\text{Type II error}))$, and a usual cutoff of 0.80 and greater is paired
214 with the alpha level of 0.05 and lower of a p-values significance test. Table 4 illustrates the
215 computed power values for each subgroup of pitches.

216

217 [Table 4]

218

219 Reducing the risk of Type I error will lead to increasing the risk of a Type II error, so both our
220 significance and power testing has to be carefully calibrated. Looking at the typical 80% power
221 cut-off with a 5% significance level, the fastball population records robust power for the Arm
222 Slot metric while the off-speed population registers robust power for the Arm Slot and Shoulder
223 Rotation metrics.

224 Discussion

225 Arm slot was found to be strongly correlated across all groups, though MotusBASEBALL's arm
226 slot was roughly 6 degrees lower than the results from our motion capture system.

227 Shoulder rotation was also strongly correlated between the two systems. On average the shoulder
228 rotation measured by MotusBASEBALL was 9 degrees lower than what the motion capture
229 system detected for the total group.

230 Arm speed from MotusBASEBALL showed strong correlations to both shoulder rotation speed
231 metrics, but no correlation to elbow extension speed. This could be due to the fact that the
232 MotusBASEBALL sensor is placed very close to the elbow joint, so movement of the forearm
233 caused by elbow extension is much less detectable due to the shorter lever arm that it detects
234 rotation from.

235 The numerical difference between the two systems is fairly substantial. Average
236 MotusBASEBALL arm speed, which was 925 deg/s, was dramatically lower than the measured
237 shoulder internal rotation speeds and magnitude of shoulder rotational velocities, which were
238 4670 deg/s and 4816 deg/s respectively. Motus defines their arm speed metric as the "resultant
239 angular velocity of the forearm segment," which was not directly calculated for this comparison.

240 It is also worth noting that the arm speed metric that MotusBASEBALL outputs in the app is
241 different than the metric that is in their web-based portal. Because MotusBASEBALL's arm
242 speed metric in the app would scale linearly to the metric in the portal, it follows that the
243 comparison of motion capture arm speed metrics to the arm speed in the app would still be valid.

244 Both comparisons to MotusBASEBALL's Stress metric were significant. Both stress
245 measurements (from MotusBASEBALL and from motion capture) were shown to be consistent
246 across the athlete population. Kinetics calculations are heavily dependent on the athlete's height
247 and weight, along with the weight of the ball (Feltner and Dapena, 1986). Motus has stated that
248 their calculation also takes these things into consideration and are part of the inputs required to

249 use the MotusBASEBALL sensor. The fact that those inputs are taken into account for both
250 stress calculations could be part of the reason there is a solid correlation between them.

251 Because the numerical outputs from the MotusBASEBALL unit are noticeably different from the
252 outputs from marker-based motion capture outputs, which is the gold standard of biomechanical
253 analysis, MotusBASEBALL may best be used for relative comparisons of an athlete.

254 MotusBASEBALL has shown to be consistent when used by the same athlete, which makes it a
255 good tool for noting significant changes to an athlete's mechanics. While the MotusBASEBALL
256 unit cannot replace motion capture, the gold-standard of biomechanical analysis, it has a
257 significant advantage in that it can be used in games and practice situations without serious
258 preparation, while clearly a marker-based biomechanics system cannot for various reasons.

259 The MotusBASEBALL unit is likely best applied by laypeople, coaches, and those who do not
260 have regular access to a sophisticated motion capture system.

261 As mentioned previously, we did not use a sleeve to place the sensor unlike what athletes would
262 do if they bought the sensor commercially. It therefore is important for athletes and coaches to
263 maintain the position of the sensor as they throw to maintain accurate readings. As movement of
264 the sleeve from the intended sensor location will likely change the readings.

265 Further research should be done with a larger sample size to further investigate the Arm Speed
266 metric to find if there is a more significant correlation to a measurement of arm speed from a
267 motion capture test.

268 **Conclusion**

269 This results from this study show that MotusBASEBALL could be a suitable low-cost and partial
270 alternative to performing a full biomechanics capture, particularly for the arm slot, shoulder
271 rotation, and stress metrics. Arm speed was shown to have a weak correlation to the results that
272 were found in the motion capture test. It should be noted that while all metrics from
273 MotusBASEBALL had significant variance in values when compared to the motion capture
274 metrics, the numbers were consistent for each subject and across all groups; Arm Slot averaged 8
275 degrees less than motion capture, Shoulder Rotation averaged 9 degrees less than motion
276 capture, and Stress averaged 41 and 42 Nm less than motion capture for elbow torque and
277 shoulder torque respectively. These three metrics could reasonably be used in future studies, and
278 for use in monitoring an individual athlete's mechanics from session to session.

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Figure 1

Placement of the motusBASEBALL sensor on the elbow

How we affixed the motusBASEBALL sensor to the arm using adhesive instead of the provided sleeve.



Figure 2

The Motion Capture System

The multi-camera OptiTrack camera system consisting of Prime 13 and Prime 13W cameras, used to evaluate pitcher kinematics and kinetics.



Table 1 (on next page)

Age, height, weight, fastball (FB) velocity, and offspeed (OS) velocity of the participants in the study

Biological and performance data on the subjects in the study.

1 TABLE 1: Age, height, weight, fastball (FB) velocity, and offspeed (OS) velocity of the
2 participants in the study
3

10 Subjects	Height (in)	Weight (lbs)	FB Velocity (mph)	OS Velocity (mph)
Age: 23.8 ± 4.0	73.3 ± 0.8	206.1 ± 5.5	83.8 ± 3.5	71.0 ± 3.6

4

5

Table 2 (on next page)

Averages of the Metrics Taken from Motion Capture Analysis Compared with the Corresponding Metrics from MotusBASEBALL

A comparison of the Motion Capture System using high-precision OptiTrack cameras compared with the metrics the motusBASEBALL unit provides.

1 TABLE 2: Averages of the Metrics Taken from Motion Capture Analysis Compared with the
 2 Corresponding Metrics from MotusBASEBALL
 3

Group	All		Fastball		Off-Speed	
Sample Size	20		10		10	
Metric	Motion Capture	MotusBASEBALL	Motion Capture	MotusBASEBALL	Motion Capture	MotusBASEBALL
Arm slot (deg)	62 ± 3	54 ± 8	63 ± 5	53 ± 8	61 ± 5	54 ± 5
Shoulder rotation (deg)	167 ± 2	158 ± 5	167 ± 3	156 ± 5	168 ± 3	157 ± 3
Arm speed - elbow extension speed (deg/s)	2404 ± 38	925 ± 24	2398 ± 49	945 ± 33	2410 ± 61	935 ± 20
Arm speed - shoulder internal rotation speed (deg/s)	4670 ± 130	925 ± 24	4648 ± 178	945 ± 33	4692 ± 199	935 ± 20
Arm speed - shoulder velocity magnitude (deg/s)	4816 ± 120	925 ± 24	4795 ± 167	945 ± 33	4838 ± 181	935 ± 20
Stress - Varus torque (Nm)	106 ± 4	65 ± 3	103 ± 5	62 ± 2	110 ± 6	64 ± 2
Stress - shoulder IR torque (Nm)	107 ± 4	65 ± 3	104 ± 5	62 ± 2	111 ± 6	64 ± 2

4

Table 3 (on next page)

P-Values and Correlation Coefficients (R^2)

Statistical analysis of the comparisons between the Motion Capture System and the motusBASEBALL unit, indicating high correlation.

TABLE 3: P-Values and Correlation Coefficients (R²)

Group	All		Fastball		Off-Speed	
Sample Size	20		10		10	
Metric	P-Value	R ²	P-Value	R ²	P-Value	R ²
Arm Slot	<0.001*	0.975	<0.001*	0.978	<0.001*	0.974
Shoulder rotation	<0.001*	0.749	0.022*	0.710	0.007*	0.784
Arm speed	0.207	0.295	0.341	0.337	0.413	0.292
Arm speed	0.001*	0.668	0.010*	0.762	0.045*	0.643
Arm speed	0.002*	0.659	0.017*	0.727	0.041*	0.651
Stress	0.001*	0.667	0.077	0.583	0.011*	0.759
Stress	0.002*	0.653	0.094	0.557	0.010*	0.763

* indicates that the metric was found to be statistically significant

Table 4(on next page)

Power Analysis of the Correlations Between Motus Biomechanics Measurements and Marker-Based Biomechanics Measurements

Statistical power of the correlations found between the motusBASEBALL unit and the gold-standard Motion Capture Analysis system.

1 TABLE 4: Power Analysis of the Correlations Between Motus Biomechanics Measurements and
2 Marker-Based Biomechanics Measurements
3

Group	Fastball	Off-Speed
Sample Size	10	10
Metric	1-Beta	1-Beta
Arm Slot	0.9999	0.9999
Shoulder rotation	0.6848	0.8260
Arm speed	0.1619	0.1315
Arm speed	0.7853	0.5574
Arm speed	0.7173	0.5728
Stress	0.4518	0.7799
Stress	0.4109	0.7863

4