

1 **Title:** Quality assessment of an UWB positioning system for indoor wheelchair court sports

2 **Submission type:** Original Article

3 **Authors:** Bertrand Perrat<sup>1</sup>, Martin J Smith<sup>1</sup>, Barry S Mason<sup>2</sup>, James M Rhodes<sup>2</sup>, Vicky L  
4 Goosey-Tolfrey<sup>2</sup>

5  
6 <sup>1</sup> The University of Nottingham (UK)

7 <sup>2</sup> Loughborough University (UK)

8

9 **Contact Details :** Bertrand Perrat

10 Email: isxpb1@nottingham.ac.uk

11 Address: Nottingham Geospatial Institute, Nottingham Geospatial Building, The University  
12 of Nottingham, Jubilee Campus, Triumph Road, Nottingham, NG7 2TU

13 Phone number : 00 44 790 883 2091

14

15 **Abstract word count:** 157

16 **Text word count:** 4910

17 **Number of figures:** 12

18 **Number of tables:** 4

19 **Abstract**

20 Ultra-Wide Band radio positioning systems are maturing very quickly and now represent a  
21 good candidate for indoor positioning. The aim of this study was to undertake a quality  
22 assessment on the use of a commercial Ultra-Wide Band positioning system for the tracking  
23 of athletes during indoor wheelchair court sports. Several aspects have been investigated  
24 including system setup, calibration, sensor positioning, determination of sport performance  
25 indicators and quality assessment of the output. With a simple setup procedure, it has been  
26 demonstrated that athletes tracking can be achieved with an average horizontal positioning  
27 error of 0.37 m ( $\sigma = \pm 0.24$  m). Distance covered can be computed after data processing with  
28 an error below 0.5% of the course length. It has also been demonstrated that the tag update  
29 rate and the number of wheelchairs on the court does not affect significantly the positioning  
30 quality; however, for highly dynamic movement tracking, higher rates are recommended for a  
31 finer dynamic recording.

32 **Keywords:** Player tracking, Ultra Wide Band, Training, Coaching, Error analysis

## 33 **1. Introduction**

34 There is an interest in the use of technology to enhance the performance of athletes in  
35 wheelchair sports whether this is for improving equipment within sports or monitoring  
36 athlete's performance. However, monitoring athletes during competition can be problematic.  
37 There are many challenges both in the collection and analysis of the data collected to fully  
38 understand where athletes can improve their performance. To add to the complexity of  
39 measuring athletes movements, team sports involve many players competing at one time on a  
40 pitch indoors or outdoors.

41 There are an increasing number of methods for tracking moving objects. Outdoors, Global  
42 Navigation Satellite System (GNSS) technology, normally the Global Positioning System  
43 (GPS), is often the simplest method to adopt [1] but indoors the situation becomes more  
44 complex as visibility of satellites for a GPS solution is not feasible or practical. There are  
45 however a number of systems available for indoor tracking [2]. These systems usually rely on  
46 different radio technologies such as WiFi [3–5], RFID [4,5] or Ultra-Wide Band [5–7].  
47 There are also a number of image based systems relying on infra-red [8] or traditional images  
48 [9]. This study focuses on the use of an Ultra-Wide Band (UWB) radio positioning system  
49 [10] which has been developed for the manufacturing, warehouse, and other industrial  
50 purposes. UWB systems are based on the use of fixed sensors around the region of movement  
51 where these sensors track the positions of tags which are fixed to the target object.

52 For every tracking application there is a requirement for a certain level of accuracy to ensure  
53 the location data collected is fit for purpose. As UWB radio positioning has not been  
54 designed for monitoring athletes, there is an even greater need to assess the capability of this  
55 technology to ensure the data collected and the derived information is valid.

## 56 **2. Aims and objectives**

57 The general aim of the current research was to determine the quality of position that can be  
58 achieved by the UWB system during a range of movements specific to the wheelchair court  
59 sports. The following objectives were investigated:

- 60 1. System setup
  - 61 a) Setup and configuration quality
  - 62 b) Sensors spatial configuration
- 63 2. Positioning quality analysis
  - 64 a) Stationnary positioning
  - 65 b) Dynamic positioning
  - 66 c) Impact of wheelchair environment
  - 67 d) Filtered positioning quality analysis for distance measurement
  - 68 e) Tag mounting location

### 69 3. Indoor Tracking System

#### 70 3.1. Background

71 UWB is a short-range, large bandwidth radio technology. Its signal properties offer a strong  
72 multipath resistance and good penetrability in materials, which makes it particularly suitable  
73 for indoor environments. Additionally, the use of UWB pulses enables a very good time-  
74 domain resolution which allows such radio systems to be used for precise location and  
75 tracking [10–14].

76 Impulse-radio-based UWB systems are composed of sensors and tags. Sensors are receivers  
77 distributed around the area of interest while tags are fixed on objects to be located and emit  
78 UWB pulses. The UWB pulse is received by a set of sensors which is used to compute a  
79 location based on Time Of Arrival (TOA), Time Difference Of Arrival (TDOA) or Angle Of  
80 Arrival (AOA) techniques [15,16].

81 The TOA technique uses the time of flight of the UWB pulse to determine the distance  
82 between the tag and the sensor. For each sensor, this leads to a sphere of possible solutions.  
83 The position is then estimated by intersecting spheres of several sensors, a technique  
84 commonly known as multi-lateration [17]. However, TOA requires a perfect time  
85 synchronisation of all sensors and tags alongside a time stamped UWB pulse so the time of  
86 flight can be determined. These requirements are critical and generally not practical for  
87 commercial systems.

88 The TDOA technique is based on the TOA principle but instead of computing the time of  
89 flight of an UWB pulse directly, it computes the difference in time of arrival between several  
90 sensors receiving the same UWB pulse. As this technique is time based, all sensors must be  
91 time synchronised. However, a benefit over TOA is that tags do not need to be synchronised.

92 If the orientation of the sensor is known, it is also possible to use the AOA technique to  
93 estimate the tag position. This is achieved for each sensor by determining the direction of  
94 arrival of the UWB pulse. Position can then be estimated by intersecting AOA of several  
95 sensors, a technique commonly known as multi-angulation [17].

96 All of these techniques are frequently combined together using non-linear regression or  
97 Kalman filtering to optimise the positioning quality [18].

#### 98 3.2. Ubisense real-time location system

99 This study focuses on the use of a commercial impulse-radio-based UWB system from  
100 Ubisense [6] (UWB system). This system is one implementation of UWB tracking among  
101 others. It is composed of sensors (Figure 1a) and tags (Figure 1b). Sensors are stationary and  
102 suitably distributed around the playing area to locate the positions of the tags "worn" by the

103 athletes. Typically, 4 or 6 sensors would be used to surround an indoor wheelchair rugby  
104 court (28 m x 15 m).

105 \*\*\*\*\* INSERT Figure 1: Ubisense Real-Time Location System

106

107 All sensors are linked by an Ethernet cable to a master sensor for time synchronisation. The  
108 tag location information is computed and displayed on a computer connected to the system.  
109 According to Ubisense Real-time location system (RTLS) specifications, an accuracy of 15-  
110 30 cm with an update rate over 10Hz can be expected.

111 The system is transportable so it can be easily deployed during competition or training at  
112 multiple venues. For this study, the system was installed and calibrated following Ubisense's  
113 recommended procedure. The first stage of the calibration is to determine sensor positions by  
114 measuring the distance between each sensor and two reference points. This is easily  
115 performed with a laser distance measurer. Each distance measured is given as an input to the  
116 Ubisense software which determines sensors' position. The next step is to perform the cable  
117 offset correction to take into account the different length of wires used to allow for time  
118 synchronisation. The last step is to determine the orientation of each sensor. Several methods  
119 are proposed by Ubisense. The method used in this study is a "dual calibration" which  
120 identifies both cable offset correction and sensors orientation at the same time using a  
121 surveyed point, approximately in the middle of the area of interest, so its coordinates are  
122 known. A tag is left stationary at this point whilst the system calibrates each sensor as the tag  
123 position is known [19]. Once this has been done, the system is set up and ready to use. A  
124 complete installation and calibration takes approximately 1 hour.

125 The UWB system only outputs tag positions so other information such as speed and distance  
126 travelled are derived from these coordinates. The different analysis and processing techniques  
127 presented in this study were integrated in software specifically developed at the University of  
128 Nottingham for indoor wheelchair court sports to assist sport scientists and coaches. [Some](#)  
129 [examples of applications can be found in \[20,21\].](#)

## 130 **4. Methodology**

### 131 *4.1. Trials and location facility*

132 All trials discussed in this paper are using the Ubisense RTLS system. Trials were undertaken  
133 in a large sports hall with a viewing gallery for setting up the surveying equipment  
134 (Section 4.2) providing a good view of the playing area (Figure 2).

135 \*\*\*\*\* INSERT Figure 2: Viewing gallery showing the Leica TS-30 surveying  
136 equipment and one UWB sensor on an elevated stand

137        *4.2. Surveying equipment*

138        The Leica TS-30 is a robotic total station that allows classic surveying tasks (angle and  
139        distance measurement) and also tracking of a moving prism shown in Figure 4.

140        According to the Leica specifications [22] the tracking mode gives a positioning accuracy of  
141        3 mm + 4 parts per million (ppm). With a 40 m maximum range for the trials presented, the  
142        ppm part can be neglected. Timing specifications give a maximum measurement rate of 5 Hz  
143        for the tracking mode.

144        When computing distance travelled from measured positions, the measurement frequency is a  
145        limitation. If the length of a trajectory is computed directly from the sum of distances  
146        between successive Leica TS-30 positions, it will be under-estimated as straight lines will  
147        join the points. In order to correct this under-estimation and knowing that Leica TS-30  
148        measurements are precise in position, an interpolation is applied to more closely follow the  
149        track (Figure 3). The interpolation used is a cubic interpolation so the interpolated trajectory  
150        effectively goes through all Leica TS-30 measured points as they are known to be precise in  
151        position.

152        \*\*\*\*\* INSERT Figure 3: Interpolation benefits to more closely follow the  
153        track

154        As a conclusion, we can say that the Leica TS-30 can also be used as a gold-standard for the  
155        distance assessment. However, this is true as long as the Leica TS-30 gives enough  
156        measurements for the interpolation to fill gaps. All Leica TS-30 traces used in this paper  
157        have been checked to ensure no unacceptable gaps (> 0.5 s) were present.

158        *4.3. Data collection*

159        As the aim was to assess the UWB positioning quality using the Leica TS-30 as a gold-  
160        standard, it was necessary to mount tags and prism as close as possible. Another critical  
161        requirement with the Leica TS-30 is to maintain a line-of-sight between the total station and  
162        the prism for the tracking. So mounted on a wheelchair was a pole with the prism on top and  
163        a plate attached to accommodate the tags, see Figure 4. With this setup, the actual horizontal  
164        position of the prism and tags is almost the same; the small offset being negligible compared  
165        with the expected system accuracy (15-30 cm). This setup was very convenient as the  
166        wheelchair can be pushed normally keeping the line-of-sight between the Leica TS-30 and  
167        the prism. Another advantage of this setup is that it was possible to reproduce wheelchair  
168        sports movements in an ecologically valid environment.

169        \*\*\*\*\* INSERT Figure 4: Tags and prism mounted on a wheelchair

170       4.4. *Data processing and smoothing*

171       The processing workflow can be seen on Figure 5 and time synchronisation is required after  
172       positions are determined as both systems are using their own internal clock for time  
173       stamping. The time synchronisation is achieved based on Fast-Fourier Transform (FFT).  
174       Once the time synchronisation is complete, the common time window is determined to make  
175       sure both systems are running during the analysis period. From here, several movement  
176       parameters and quality indicators are computed.

177       The first quality measure is obtained by computing the horizontal position error of each UWB  
178       measurement compared to the interpolated Leica trace considered as the gold standard. The  
179       second quality indicator is based on the distance covered as it is one of the key metrics used  
180       to monitor sports performance. This was computed by summing up the distance between  
181       consecutive points for both Ubisense tags and the Leica TS-30 interpolated data. However, as  
182       the UWB positions are subject to random noise, a filter was applied to mitigate this effect.  
183       The filtering used was a 3-pass sliding-average with a window size proportional to the  
184       acquisition frequency.

185       \*\*\*\*\* INSERT Figure 5: Workflow of the processing used in this study

186       **5. Trials, Results and Analysis**

187       5.1. *System setup*

188       5.1.1. *Setup and configuration quality*

189       The first objective was an assessment of the setup procedure recommended by Ubisense,  
190       which is a quick, low-cost and practical approach to the system installation. As mentioned in  
191       Section 3.2, the main step during this installation is to input sensor locations into the  
192       Ubisense software. Sensor locations were obtained by laser distance measurements relatively  
193       to two reference points, one at either end of the playing area.

194       In order to assess the precision of the laser measurement technique; sensors and reference  
195       points have been surveyed with the Leica TS-30. Comparing laser distances with Leica TS-30  
196       equivalent distances gives a root mean square error of 4 cm which is a typical result  
197       according to the measurement technology used.

198       The distance measurements were provided to the Ubisense Location Engine Configuration  
199       (LEC) tool [23] which computed estimates for sensors' locations. The comparison between  
200       these estimates and the ground truth provided by the Leica TS-30 has been made for one  
201       particular setup of 5 sensors (Table 1). This shows an expected result from the Ubisense setup  
202       procedure. There is some evidence of small systematic bias (for example all negative X  
203       differences) and random measurement errors (for example the variation in the negative X  
204       values) which are of a typical magnitude for the system.

205 \*\*\*\*\* INSERT Table 1: Sensors position differences using Ubisense LEC  
206 tool

### 207 5.1.2. Sensors spatial configuration

208 One advantage of a flexible UWB system is that the sensors' distribution can be adjusted to  
209 optimise the coverage of the area being tracked. Trials were undertaken in order to find the  
210 optimum sensors' spatial distribution to cover an indoor wheelchair rugby court.

211 In order to assess the impact of sensor locations on the output of the system, the wheelchair  
212 has been used with both tags and prism mounted. The trajectory pattern used to maximise  
213 coverage of the playing area is shown in Figure 6. Positioning quality has been evaluated  
214 with a statistical analysis on the horizontal positioning error of the UWB system obtained as  
215 described in Section 4.4.

216 \*\*\*\*\* INSERT Figure 6: Trajectory pattern used for sensors spatial  
217 configuration analysis

218 Different configurations have been tried to get an optimal coverage of the court using  
219 respectively four, five and six sensors. Using four sensors, one in each corner of the court,  
220 resulted in a mean error of 0.40 m ( $\sigma = \pm 0.28$  m). Similar results were obtained when adding  
221 one sensor on one of the middle side with a mean error of 0.39 m ( $\sigma = \pm 0.29$  m). Finally,  
222 slightly better results (Ubisense track closer to the TS-30 track) were obtained using six  
223 sensors, one in each corner at 4m height and two on the middle sides at 2 m providing a mean  
224 error of 0.35 m ( $\sigma = \pm 0.23$  m). This is the optimum setup tried and is the one used for all the  
225 trials presented next. A spatial analysis has been performed to identify possible areas with  
226 bad coverage. However, it appeared that the noise and random errors of the system did not  
227 allow for the identification of consistently weak areas.

## 228 5.2. Positioning quality analysis

### 229 5.2.1. Stationary Positioning

230 In order to assess the stationary positioning quality, tags were left stationary for 5 minutes in  
231 known court locations. As an example, Figure 7 shows the output of one tag left stationary on  
232 a corner of the playing area and is a typical pattern of measurements from various positions  
233 around the court. The plot shows that positions out of the UWB system are separated into two  
234 distinct clusters. This clustering can be explained by the noise due to sensors sets switching  
235 as described in Banerjee [24] which also propose a particle filter to mitigate this effects.

236 \*\*\*\*\* INSERT Figure 7: Stationary tag positioning

### 237 5.2.2. Dynamic Positioning

238 The first assessment evaluated the horizontal positioning error of each position output by the  
239 UWB system. In order to collect data relevant to indoor wheelchair court sports, trials were

240 conducted during a simulated wheelchair rugby match with 1 participant playing 4 quarters of  
241 8 minutes. The participant was asked to simulate a match play including turns, pivots, back  
242 and forth movements with rapid changes of speed. An example of a trajectory during a  
243 quarter is visible on Figure 8.

244 \*\*\*\*\* INSERT Figure 8: Example trajectory during a quarter of a simulated  
245 match

246 Two matches of this format have been conducted. Table 2 presents the statistical analysis of  
247 the error for both two matches.

248 \*\*\*\*\* INSERT Table 2: Mean (m)  $\pm$  standard deviation (m) of the UWB  
249 positioning error during 2 simulated matches with 9 tags operating at 3 different update rates

250 Results illustrated the accuracy of the system with a horizontal positioning mean error of  
251 0.37m and a standard deviation of  $\pm 0.24$ m. Detailed results are presented to highlight the  
252 consistency of the system. Note that numbers obtained are similar to those obtained in the  
253 sensor spatial configuration trial (Section 5.1.2) where a mean error of 0.35 m and a standard  
254 deviation of  $\pm 0.23$  m were found. Additionally, the tag update rate does not seem to affect the  
255 positioning quality according to the values grouped by update rate.

256 Finally, the cumulative distribution function (CDF) of the positioning error has been  
257 computed and a typical example is shown on Figure 9. The CDF has been represented for the  
258 3 different tag update rates used (4Hz, 8Hz, 16Hz) using the horizontal error on one  
259 simulated match (4 quarters of 8 min). The closeness of curves in Figure 9 show that tag  
260 update rate doesn't have a significant impact on the error distribution which confirms that the  
261 tag update rate does not affect the positioning quality as already seen on Table 2. The 90th-  
262 percentile positioning error is 0.63m regardless of tag rate. Similar results were obtained with  
263 an equivalent setup procedure in Muthukrishnan [18] where it is also compared to more  
264 complex system setups. The nominal update rates (4Hz, 8Hz, 16Hz) were checked against  
265 the time stamped measurement records and agreed within less than 1Hz.

266 \*\*\*\*\* INSERT Figure 9: Cumulative Distribution Function of the positioning  
267 error

### 268 5.2.3. *Impact of wheelchair environment*

269 Results presented in previous sections were obtained with only one wheelchair moving on the  
270 court which is not representative of a wheelchair rugby environment. In order to address this,  
271 a match (4 quarters of 8 minutes) has been simulated with another wheelchair interacting on  
272 the court. To simulate a game this involved close engagement between the wheelchairs and  
273 more distant separation between them. The positioning quality was assessed using the  
274 statistical analysis on the horizontal positioning error described in Section 4.4. These trials  
275 were done using the setup with 5 sensors described in Section 5.1.2 which reported a mean  
276 horizontal positioning error of 0.39 m ( $\sigma = \pm 0.29$  m).

277 During the first two quarters, one wheelchair was tracked by both the UWB system and the  
278 Leica TS-30, whilst the second wheelchair remained untracked. The first and second quarters  
279 reported a mean error of 0.38 m ( $\pm$  0.30 m) and 0.39 m ( $\pm$  0.35 m) respectively. Such results  
280 demonstrate that perturbations caused by a second wheelchair on court do not influence the  
281 positioning quality. To determine whether the tags were working entirely independent, the  
282 second wheelchair was also tracked by the UWB system during the third and fourth quarters.  
283 Subsequently, error values remained similar, with a mean error of 0.38 m ( $\pm$  0.26 m) and 0.36  
284 m ( $\pm$  0.27 m). As a result, tracking several independent objects in close and distant  
285 relationships did not affect the positioning quality.

286 Finally, some real match data has been investigated to evaluate the impact of having eight  
287 athletes and one referee on the court. Due to the need of a constant line of sight between the  
288 surveying equipment and the object tracked, no reference data was available. However, the  
289 data has been investigated regarding its availability (presence of data gaps) and visual  
290 correctness of the track.

291 As the real match data was obtained using a tag mounted on the foot strap, the presence of  
292 data gaps was compared to the results obtained in Section 5.2.5 using a similar tag mounting  
293 location. A gap is defined as being a data outage for more than 0.5s. The results from  
294 Section 5.2.5 are representative of a situation where only one wheelchair is on the court.  
295 Over the 488.1 seconds of the trial, 32 gaps were detected, giving an average occurrence of 1  
296 gap every 15.25 seconds of an average duration of 1.20 seconds. The gaps distribution is  
297 visible on Figure 10.

298 \*\*\*\*\* INSERT Figure 10: Distribution of the data gaps with one wheelchair  
299 on the court

300 Moving to the real match data, the whole match was considered with the 4 quarters of 8  
301 minutes including any stop in the game (ball out of play, fouls, ...) representing a total  
302 dataset of approximately 70 minutes. Over the 4196.8 seconds of the dataset, 299 gaps were  
303 detected, giving an average occurrence of 1 gap every 14.0 seconds of an average duration of  
304 1.28 seconds. Similarly, the gaps distribution is visible on Figure 11.

305 \*\*\*\*\* INSERT Figure 11: Distribution of the data gaps during a real match  
306 with 8 athletes + 1 referee on the court

307 The second aspect of the data that has been investigated is the visual correctness of the track.  
308 This visual investigation also included a real-time replay of the track to detect any unnatural  
309 or irregular movement of the athlete. After investigation, no major anomalies or irregularities  
310 could be detected. An example of the track produced for a quarter during a real match is  
311 visible on Figure 12.

312 \*\*\*\*\* INSERT Figure 12: Example of an athlete's trajectory during one  
313 quarter of a real match

314           5.2.4. *Filtered positioning quality analysis for distance measurement*

315 Before being able to compute the distance covered, UWB measurements were processed  
316 using the smoothing described in Section 4.4. Then distances are computed for both UWB  
317 smoothed measurements and ‘interpolated’ Leica tracks and compared.

318 Detailed results are presented in Table 3 which summarises the errors obtained for 9 tags,  
319 operating at 3 different update rates, during the 2 simulated matches.

320 \*\*\*\*\* INSERT Table 3: Distance error of 9 tags operating at 3 different  
321 update rates during 2 simulated matches

322 The mean error on the distance travelled for all tags of the UWB system is 0.45% of course  
323 length. Each quarter trajectory was approximately 1000m in length therefore a 0.45% error is  
324 equivalent to 4.5m. Additionally, results showed that the higher the update rate, the better the  
325 distance estimate. This can be explained as the trajectory will be composed of more points  
326 which give a better recording of the dynamics with the same positioning quality  
327 (Section 5.2.2).

328           5.2.5. *Tag mounting location*

329 Section 5.2.4 presented the results for distance estimation considering different update rates  
330 with tags attached as shown in Figure 4. In order to find mounting locations more appropriate  
331 to wheelchair court sports, a specific trial was conducted with tags located in different places  
332 on the wheelchair or worn by the athlete. Below are the detailed tag mounting locations  
333 considered:

- 334           • Vest: Tag positioned between the scapula using a GPS vest worn by the participant
- 335           • Frame: Tag attached to the wheelchair frame located at the front of the chair.
- 336           • Foot strap: Tag positioned onto the foot strap of the wheelchair
- 337           • Camber bar: Tag secured to the camber bar of the chair located beneath the seat.

338 Additionally, one tag was left attached to the prism pole to allow for a direct comparison with  
339 the results presented in Section 5.2.4. A quarter of 8 minutes has been simulated and distance  
340 estimates have been computed for each tag. The Leica TS-30 has measured a distance of  
341 752.81 m for this quarter. Results are summarised in Table 4 which also includes the number  
342 of data gaps (no measurement for more than 0.5s) for each tag.

343 \*\*\*\*\* INSERT Table 4: Distance estimates and data gaps for different  
344 mounting locations

345 **6. Discussion**

346 The setup procedure that has been used and assessed in the current study was recommended  
347 by Ubisense which was chosen for its simplicity, convenience and speed of set up. Results of  
348 Section 5.1.1 showed that the root mean square error using a laser distance measurer is  
349 around 4 cm when measuring the distance between each sensor and reference points. This

350 was an acceptable result for the technology used. More expensive, time-consuming and  
351 complex setup procedure are possible [18,19] but wouldn't be as easy and quick to deploy for  
352 a mobile system used in a sports environment.

353 Several sensor configurations were investigated to optimise coverage of the court. The best  
354 configuration was obtained using 6 sensors, with one in each corner of the court 4m high and  
355 2 on the middle-sides of the court 2m high. With such a configuration the mean horizontal  
356 positioning error was found to be 0.35 m ( $\sigma = \pm 0.23$  m). The stationary positioning analysis  
357 in the study showed that the distribution of computed positions is typical of an UWB radio-  
358 positioning systems with evidence of sensors set switching noise [24].

359 An important aspect of this study was the dynamic positioning quality assessment using a  
360 robotic total station with tracking capabilities. Previous studies were limited on dynamic  
361 assessment by performing only basic linear drills [25], differential comparison (tags  
362 comparison instead of a gold-standard comparison) [7] or by asking participants to follow a  
363 predefined path marked on the ground [25]. These have shortcomings for the validation of an  
364 UWB system in a sports performance context where athletes perform multi-directional  
365 movements at varying intensities. A more recent study [26] addressed most of these  
366 shortcomings by using a trundle wheel to obtain a distance reference. While this allows more  
367 freedom for the players, it still has some limitations in the dynamics that can be tracked and  
368 the measure may not reflect exactly the actual distance covered by the athlete. Finally, as  
369 mentioned by the authors [26], the trundle wheel provides a way to assess distance estimation  
370 but does not provide any information about the positioning quality. The protocol using a  
371 robotic total station addresses these shortcomings. Results during 2 simulated matches  
372 showed a mean horizontal positioning error of 0.37 m ( $\sigma = \pm 0.24$  m) which correlates with  
373 the sensors spatial configuration analysis results ( $\mu = 0.35$  m,  $\sigma = \pm 0.23$  m). This validated  
374 the capacity of an UWB system to track highly dynamic movements. Additionally, having a  
375 second wheelchair moving and tracked on the court did not affect the positioning quality.  
376 While it was not possible to use the surveying equipment during a real match due to the need  
377 for a constant line of sight, some real match data has been investigated. Two aspects were  
378 considered, the availability and regularity or smoothness of the track. The availability study  
379 was based on a direct comparison to the results obtained in Section 5.2.5. With only one  
380 athlete on the court, on average, a data gap occurred every 15.2 seconds for an average  
381 outage of 1.20 seconds. During a real match, with eight athletes and one referee on the court,  
382 the occurrence rate slightly increased to one gap every 14.0 seconds for an average outage of  
383 1.28 seconds. The comparison of the respective gaps distribution (Figure 10 and Figure 11)  
384 shows that larger gaps ( $> 2.5$  seconds) are more likely to appear during a real match situation.  
385 However, this difference is not significant since 94% of the gaps are below 2.5 seconds in the  
386 real match dataset compared to 97% in the single wheelchair situation. Additionally, the  
387 visual correctness of the track has been checked using a real-time replay and did not show  
388 any evidence of major anomalies or irregularities. As a conclusion, going from a single  
389 athlete on the court to a real match situation with eight athletes and one referee only slightly  
390 degrades the performance of the tracking system. However, according to the results of our

391 analysis, the degradation is not significant and does not affect the suitability of the system to  
392 be used for indoor wheelchair court sports.

393 The next focus of this study was on the distance covered, an important metric in analysing  
394 athlete's performance. A previous study [26] assessed that the same UWB system can  
395 provide a distance estimate with an error of  $3.45 \pm 1.99$  % of the course length in a basketball  
396 context. These results were obtained using a combination of Kalman filter and low-pass filter.  
397 The approach adopted in this study uses a 3-pass sliding average (Section 4.4). Using this  
398 filtering technique the distance can be known with a mean error below 0.5% of course length.  
399 This difference in results may also be partly explained by the protocol used in [26] as the  
400 trundle wheel does not follow exactly the same path as the tag. Also, the tags update rate in  
401 [26] (4Hz) may contribute to a reduced quality of distance measurement. The present work  
402 found that higher tag update rates ( $\geq 8$  Hz) are more suitable for distance estimation as outlier  
403 effects can be more easily mitigated by the processing and are also giving the finest recording  
404 of the dynamics.

405 The tag attached to the prism pole provided the best distance estimate with results in  
406 agreement with those found in Section 5.2.4. Regarding mounting locations relevant to  
407 wheelchair sports, the vest appeared to show the smallest error. It also appeared that lower  
408 mounting places are more subject to data gaps which could affect the distance estimate.  
409 Nevertheless, the magnitude of error was still minimal regardless of location, and lower  
410 places may offer the most practically relevant tag locations.

## 411 **7. Conclusion**

412 This study has assessed the quality of an UWB system for tracking wheelchair athletes  
413 indoors. With a quick and easy deployment procedure, dynamic tracking can be achieved  
414 with a mean horizontal positioning error of 0.37m ( $\sigma = \pm 0.24$  m). Additionally, distance  
415 covered can be determined with an error below 0.5% of course length with adequate data  
416 processing. Tag update rates did not have a significant impact on the positioning quality;  
417 however, higher rates ( $\geq 8$  Hz) provided a greater number of points to more closely record the  
418 high dynamic movements. It was also found that having many wheelchairs on the court did  
419 not have a significant effect on the positioning. Finally, several tag mounting places have  
420 been tried with the smallest error obtained for the tag worn by the athlete in a GPS vest.  
421 Although the results presented are sport specific the method has wider potential application to  
422 other indoor and possibly outdoor sports.

423 **Acknowledgements**

424 This work was supported by the Engineering and Physical Sciences Research Council  
425 (EPSRC) through an Industrial CASE studentship in collaboration with the English Institute  
426 of Sport (formerly UK Sport); the Peter Harrison Centre for Disability Sport at  
427 Loughborough University along with UK Sport (for funding the equipment). The authors  
428 would like to acknowledge McLaren Applied Technologies for their initial input into the  
429 project.

430 **References**

- 431 1. Barbero-Alvarez JC, Coutts A, Granda J, Barbero-Alvarez V, Castagna C. The validity  
432 and reliability of a global positioning satellite system device to assess speed and  
433 repeated sprint ability (RSA) in athletes. *J Sci Med Sport [Internet]*. 2010  
434 Mar;13(2):232–5. Available from: <http://www.ncbi.nlm.nih.gov/pubmed/19446495>.  
435 Accessed November 2012, 4
- 436 2. Mautz R. Indoor Positioning Technologies [Internet]. 2012 Accessed December 2013,  
437 22. Available from: <http://e-collection.library.ethz.ch/eserv/eth:5659/eth-5659-01.pdf>.  
438 Accessed December 2013, 22
- 439 3. ZXY [Internet]. Available from: <http://www.zxy.no>. Accessed September 2014, 10
- 440 4. Aeroscout [Internet]. Available from: <http://aeroscout.com/asset-tracking>. Accessed  
441 September 2014, 10
- 442 5. Zebra [Internet]. Available from: <http://www.zebra.com/gb/en/solutions/location-solutions/location-solutions-overview.html>. Accessed September 2014, 10
- 444 6. Ubisense [Internet]. Available from: <http://www.ubisense.net/en/>. Accessed December  
445 2013, 22
- 446 7. Hedley M, Zhang J. Accurate Wireless Localization in Sports. 2012;:64–70. Available  
447 from: [http://ieeexplore.ieee.org/xpls/abs\\_all.jsp?arnumber=6178191](http://ieeexplore.ieee.org/xpls/abs_all.jsp?arnumber=6178191). Accessed April  
448 2013, 10
- 449 8. Codamotion [Internet]. Available from: <http://www.codamotion.com/>. Accessed  
450 September 2014, 10
- 451 9. Sportvu [Internet]. Available from: <http://www.stats.com/sportvu/sportvu.asp>.  
452 Accessed September 2014, 10
- 453 10. Stelzer a. Concept and application of LPM-a novel 3-D local position measurement  
454 system. *IEEE Trans Microw Theory Tech [Internet]*. 2004 Dec;52(12):2664–2669.  
455 Available from:  
456 <http://ieeexplore.ieee.org/lpdocs/epic03/wrapper.htm?arnumber=1366537>. Accessed  
457 November 2013, 11
- 458 11. Ramirez-Mireles F. On the performance of ultra-wide-band signals in Gaussian noise  
459 and dense multipath. *IEEE Trans Veh Technol [Internet]*. 2001;50(1):244–249.  
460 Available from: [http://ieeexplore.ieee.org/xpls/abs\\_all.jsp?arnumber=917932](http://ieeexplore.ieee.org/xpls/abs_all.jsp?arnumber=917932).  
461 Accessed November 2013, 11
- 462 12. Siwiak K. Ultra-wide band radio: introducing a new technology. *Veh Technol Conf*  
463 *[Internet]*. 2001; Available from:  
464 [http://ieeexplore.ieee.org/xpls/abs\\_all.jsp?arnumber=944546](http://ieeexplore.ieee.org/xpls/abs_all.jsp?arnumber=944546). Accessed November  
465 2013, 11

- 466 13. Zhang J, Orlik P V., Sahinoglu Z, Molisch AF, Kinney P. UWB Systems for Wireless  
467 Sensor Networks. *Proc IEEE [Internet]*. 2009 Feb;97(2):313–331. Available from:  
468 <http://ieeexplore.ieee.org/lpdocs/epic03/wrapper.htm?arnumber=4802196>. Accessed  
469 January 2014, 13
- 470 14. Sathyan T, Humphrey D. WASP: A system and algorithms for accurate radio  
471 localization using low-cost hardware. *IEEE Trans Syst Man Cybern [Internet]*.  
472 2011;41(2):211–222. Available from:  
473 [http://ieeexplore.ieee.org/xpls/abs\\_all.jsp?arnumber=5499111](http://ieeexplore.ieee.org/xpls/abs_all.jsp?arnumber=5499111). Accessed May 2013, 2
- 474 15. Liu H, Darabi H. Survey of wireless indoor positioning techniques and systems. *IEEE*  
475 *Trans Syst Man Cybern [Internet]*. 2007;37(6):1067–1080. Available from:  
476 [http://ieeexplore.ieee.org/xpls/abs\\_all.jsp?arnumber=4343996](http://ieeexplore.ieee.org/xpls/abs_all.jsp?arnumber=4343996). Accessed November  
477 2013, 11
- 478 16. Rappaport T, Reed J, Woerner B. Position location using wireless communications on  
479 highways of the future. *IEEE Commun Mag [Internet]*. 1996;(October):33–41.  
480 Available from: [http://ieeexplore.ieee.org/xpls/abs\\_all.jsp?arnumber=544321](http://ieeexplore.ieee.org/xpls/abs_all.jsp?arnumber=544321).  
481 Accessed November 2013, 11
- 482 17. Uren J, Price W. Control Networks. In: Surveying for engineers. Palgrave Macmillan;  
483 2006. p. 241–245.
- 484 18. Muthukrishnan K, Hazas M. Position estimation from UWB pseudorange and angle-  
485 of-arrival: A comparison of non-linear regression and Kalman filtering. *Locat Context*  
486 *Aware [Internet]*. 2009; Available from: [http://link.springer.com/chapter/10.1007/978-](http://link.springer.com/chapter/10.1007/978-3-642-01721-6_14)  
487 [3-642-01721-6\\_14](http://link.springer.com/chapter/10.1007/978-3-642-01721-6_14). Accessed May 2013, 2
- 488 19. Mandeljc R, Perč J, Kristan M, Kovač S. An Alternative Way to Calibrate Ubisense  
489 Real-Time Location System via Multi-Camera Calibration Methods [Internet]. In: 19th  
490 International Electrotechnical and Computer Science Conference. 2010 Accessed  
491 December 2013, 22. Available from: <http://vision.fe.uni-lj.si/docs/rokm/erk.pdf>.  
492 Accessed December 2013, 22
- 493 20. Rhodes J, Mason B, Perrat B, Smith M, Goosey-Tolfrey V. The validity and reliability  
494 of a novel indoor player tracking system for use within wheelchair court sports. *J*  
495 *Sports Sci [Internet]*. 2014 Apr 23;(June 2014):1–9. Available from:  
496 <http://www.ncbi.nlm.nih.gov/pubmed/24758599>. Accessed June 2014, 4
- 497 21. Rhodes JM, Mason BS, Perrat B, Smith MJ, Malone LA, Goosey-Tolfrey VL. Activity  
498 Profiles of Elite Wheelchair Rugby Players During Competition. *Int J Sports Physiol*  
499 *Perform [Internet]*. 2014 Sep 5; Available from:  
500 <http://europepmc.org/abstract/med/25202822>. Accessed December 2014, 19
- 501 22. Leica. Leica TS30 - Technical data [Internet]. Accessed December 2013, 22. Available  
502 from: [http://www.leica-geosystems.com/en/Engineering-Monitoring-TPS-Leica-](http://www.leica-geosystems.com/en/Engineering-Monitoring-TPS-Leica-TS30_77093.htm)  
503 [TS30\\_77093.htm](http://www.leica-geosystems.com/en/Engineering-Monitoring-TPS-Leica-TS30_77093.htm). Accessed December 2013, 22

- 504 23. Ubisense. How to set up a Ubisense system [Internet]. Available from:  
505 [http://eval.ubisense.net/howto/SystemSetup1\\_article/SystemSetup1.html](http://eval.ubisense.net/howto/SystemSetup1_article/SystemSetup1.html). Accessed  
506 September 2014, 9
- 507 24. Banerjee S, Suski W, Hoover A. Sensor set switching noise in UWB indoor position  
508 tracking. *2012 IEEE Int Conf Ultra-Wideband*. 2012 Sep;:297–301.
- 509 25. Sathyan T, Shuttleworth R, Hedley M, Davids K. Validity and reliability of a radio  
510 positioning system for tracking athletes in indoor and outdoor team sports. *Behav Res*  
511 *Methods [Internet]*. 2012 Apr 5;:1108–1114. Available from:  
512 <http://www.ncbi.nlm.nih.gov/pubmed/22477436>. Accessed October 2012, 30
- 513 26. Leser R, Schleindlhuber A, Lyons K, Baca A. Accuracy of an UWB-based position  
514 tracking system used for time-motion analyses in game sports. *Eur J Sport Sci*  
515 *[Internet]*. 2014 Feb 10;(September):37–41. Available from:  
516 <http://www.ncbi.nlm.nih.gov/pubmed/24512176>. Accessed September 2014, 9

517

518 **Tables :**

519

520 **Table 1: Sensors position differences using Ubisense LEC tool**

	X [along court] difference (m)	Y [cross court] difference (m)	Z difference (m)	Total difference
Sensor 1	-0.224	-0.031	0.094	0.245
Sensor 2	-0.128	-0.058	0.001	0.141
Sensor 3	-0.069	-0.134	0.017	0.152
Sensor 4	-0.102	+0.050	0.037	0.119
Sensor 5	-0.165	-0.252	0.002	0.301

521

522

523

**Table 2: Mean (m) ± standard deviation (m) of the UWB positioning error during 2 simulated matches with 9 tags operating at 3 different update rates**

	Match 1				Match 2			
	Q1	Q2	Q3	Q4	Q1	Q2	Q3	Q4
Tag 1 (16Hz)	0.33 ± 0.21	0.31 ± 0.21	0.36 ± 0.26	0.34 ± 0.22	0.39 ± 0.24	0.34 ± 0.22	0.41 ± 0.25	0.39 ± 0.27
Tag 2 (16Hz)	0.36 ± 0.23	0.31 ± 0.22	0.36 ± 0.25	0.35 ± 0.23	0.40 ± 0.22	0.34 ± 0.23	0.40 ± 0.24	0.39 ± 0.26
Tag 3 (16Hz)	0.36 ± 0.22	0.34 ± 0.21	0.39 ± 0.24	0.37 ± 0.22	0.40 ± 0.26	0.37 ± 0.21	0.41 ± 0.25	0.41 ± 0.25
Tag 4 (8Hz)	0.34 ± 0.21	0.33 ± 0.21	0.36 ± 0.25	0.36 ± 0.22	0.39 ± 0.23	0.33 ± 0.23	0.40 ± 0.24	0.39 ± 0.27
Tag 5 (8Hz)	0.36 ± 0.22	0.34 ± 0.21	0.38 ± 0.26	0.37 ± 0.22	0.41 ± 0.24	0.37 ± 0.24	0.41 ± 0.24	0.42 ± 0.25
Tag 6 (8Hz)	0.35 ± 0.21	0.34 ± 0.26	0.37 ± 0.25	0.36 ± 0.23	0.40 ± 0.24	0.37 ± 0.22	0.41 ± 0.25	0.41 ± 0.28
Tag 7 (4Hz)	0.35 ± 0.27	0.32 ± 0.22	0.38 ± 0.25	0.36 ± 0.24	0.39 ± 0.26	0.33 ± 0.23	0.40 ± 0.25	0.40 ± 0.26
Tag 8 (4Hz)	0.35 ± 0.23	0.32 ± 0.21	0.39 ± 0.25	0.36 ± 0.24	0.39 ± 0.26	0.34 ± 0.22	0.41 ± 0.26	0.40 ± 0.26
Tag 9 (4Hz)	0.36 ± 0.23	0.34 ± 0.22	0.41 ± 0.28	0.36 ± 0.24	0.38 ± 0.30	0.38 ± 0.25	0.42 ± 0.26	0.42 ± 0.26
Mean	(16 Hz) 0.37 ± 0.23 m		(8 Hz) 0.37 ± 0.24 m		(4 Hz) 0.37 ± 0.25 m		(Total) 0.37 ± 0.24 m	

524

525

526 **Table 3: Distance error of 9 tags operating at 3 different update rates during 2 simulated matches**

	Match 1				Match 2			
	Q1	Q2	Q3	Q4	Q1	Q2	Q3	Q4
Tag 1 (16Hz)	-0.05%	-0.40%	0.00%	0.04%	-0.55%	0.22%	-0.07%	-0.36%
Tag 2 (16Hz)	-0.62%	-1.05%	-0.25%	-0.56%	-0.90%	-0.35%	-0.14%	-0.36%
Tag 3 (16Hz)	-0.44%	-0.76%	-0.06%	-0.43%	-0.72%	0.13%	-0.08%	-0.02%
Tag 4 (8Hz)	-0.63%	-1.09%	-0.38%	-0.50%	-0.80%	-0.56%	-0.46%	-1.04%
Tag 5 (8Hz)	-0.40%	-0.97%	0.01%	-0.35%	-0.72%	0.00%	-0.04%	-0.17%
Tag 6 (8Hz)	0.27%	0.20%	0.49%	0.33%	-0.12%	0.78%	0.42%	0.63%
Tag 7 (4Hz)	-0.92%	-1.68%	-0.80%	-0.82%	-1.22%	-1.07%	-0.46%	-1.19%
Tag 8 (4Hz)	-0.05%	-0.56%	0.07%	0.31%	-0.54%	0.00%	0.22%	-0.12%
Tag 9 (4Hz)	-0.21%	-0.72%	-0.09%	-0.27%	-0.51%	0.49%	-0.01%	-0.11%
Mean	(16 Hz) 0.36 %		(8 Hz) 0.47 %		(4 Hz) 0.52 %		(Total) 0.45 %	

527

528 **Table 4: Distance estimates and data gaps for different mounting locations**

	Distance estimate	Error (%)	Data gaps
Prism pole	751.8 m	-0.13 %	1
Vest	755.4 m	0.35 %	6
Frame	750.1 m	-0.36 %	31
Foot strap	745.9 m	-0.91 %	32
Camber bar	736.9 m	-1.85 %	34

529

530

531 **Figures :**

532



a) Sensor (20cm x 14cm x 9.5cm)

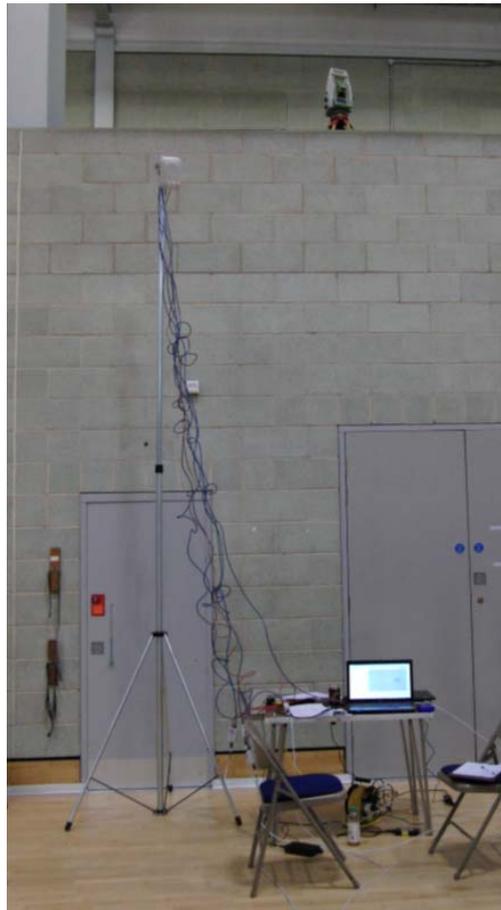
b) Tag (40mm x 40mm x 10mm)

533

534

**Figure 1: Ubisense Real-Time Location System [6]**

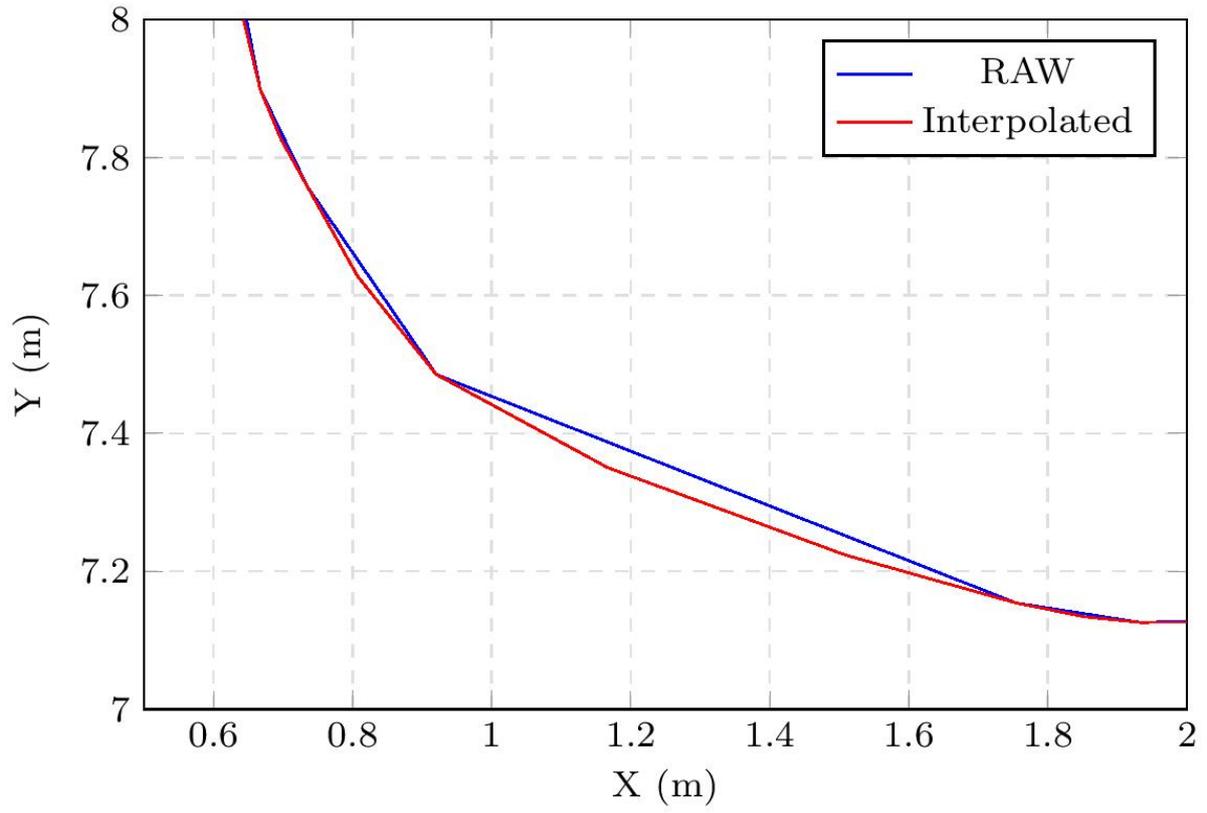
535



536

537

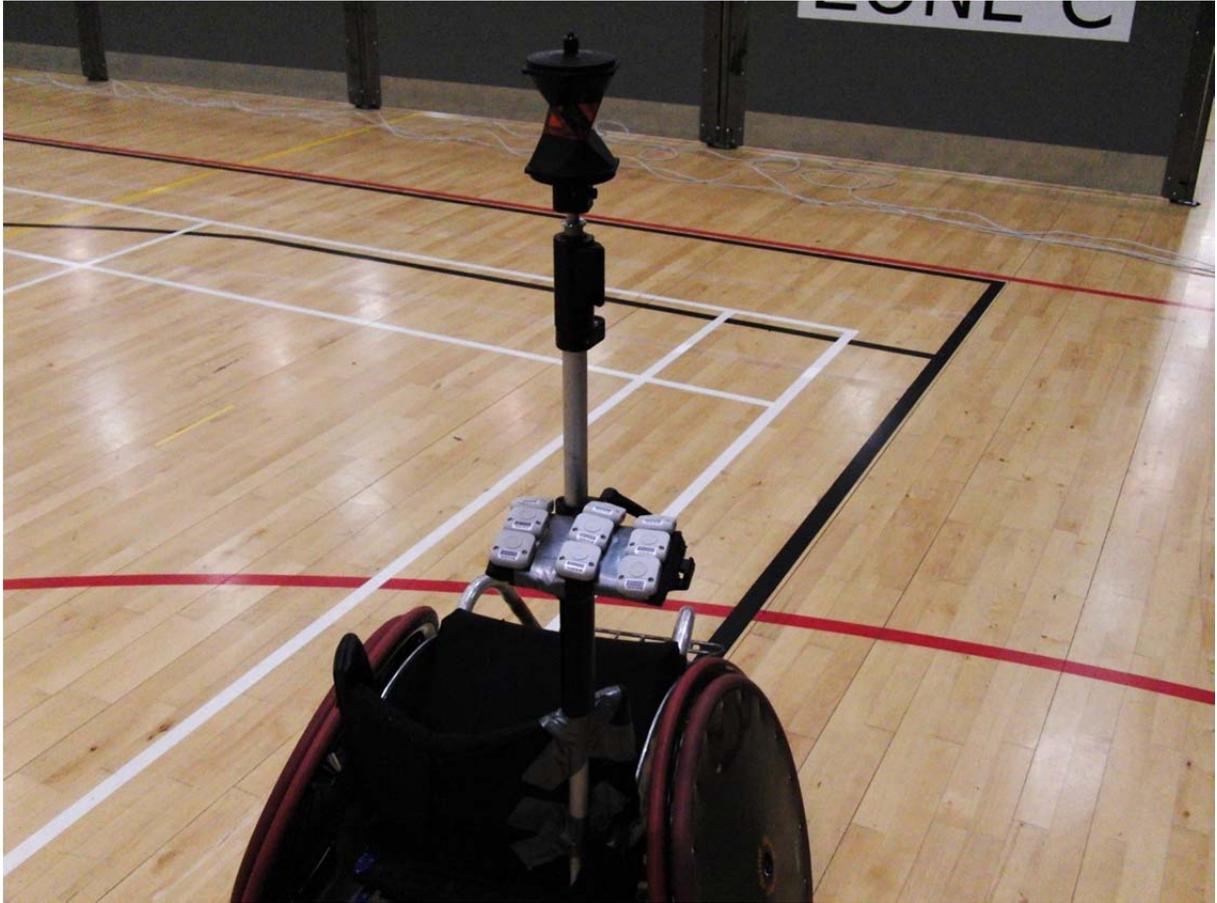
**Figure 2: Viewing gallery showing the Leica TS-30 surveying equipment and one UWB sensor on an elevated stand**



538

539

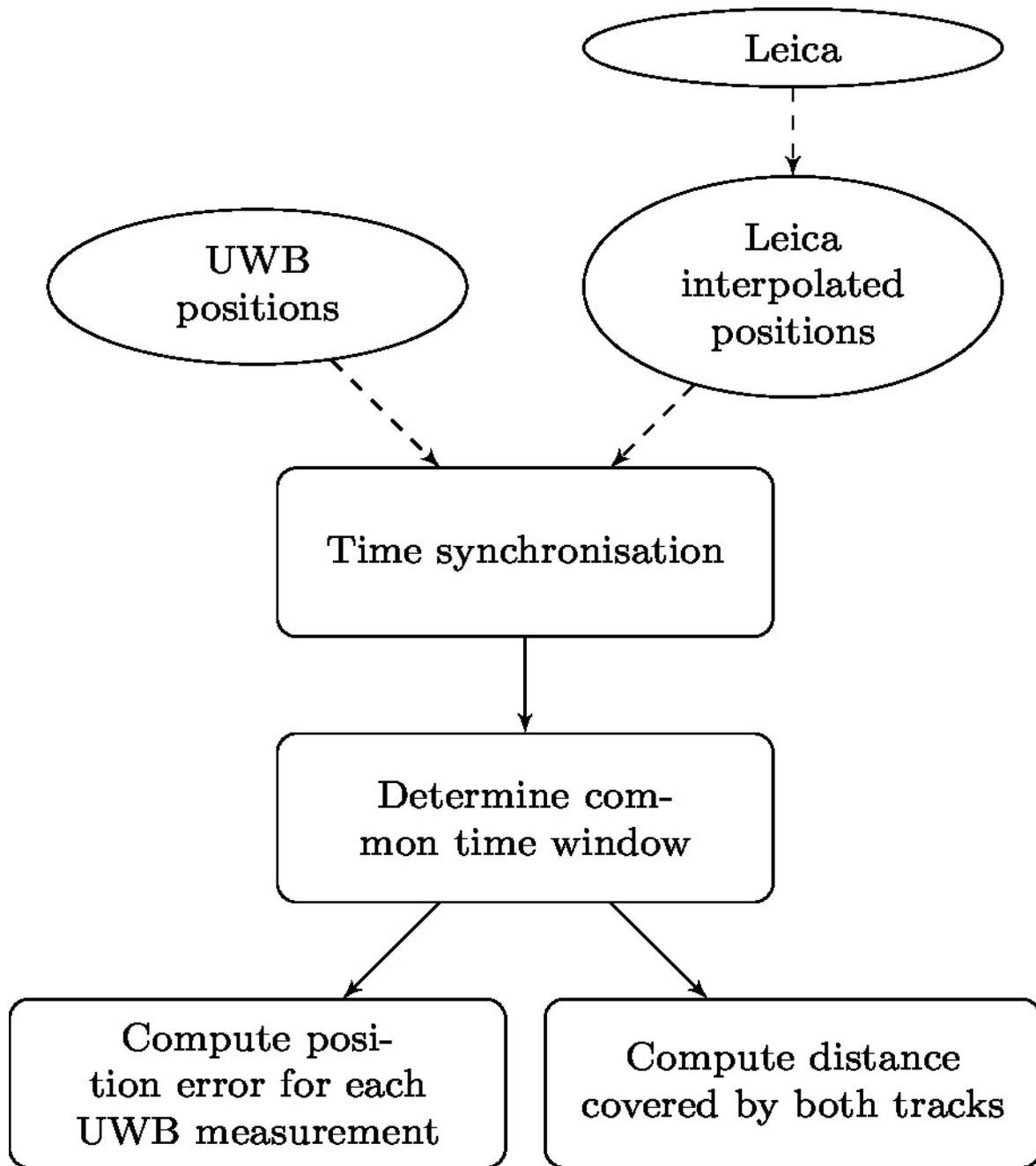
**Figure 3: Interpolation benefits to more closely follow the track**



540

541

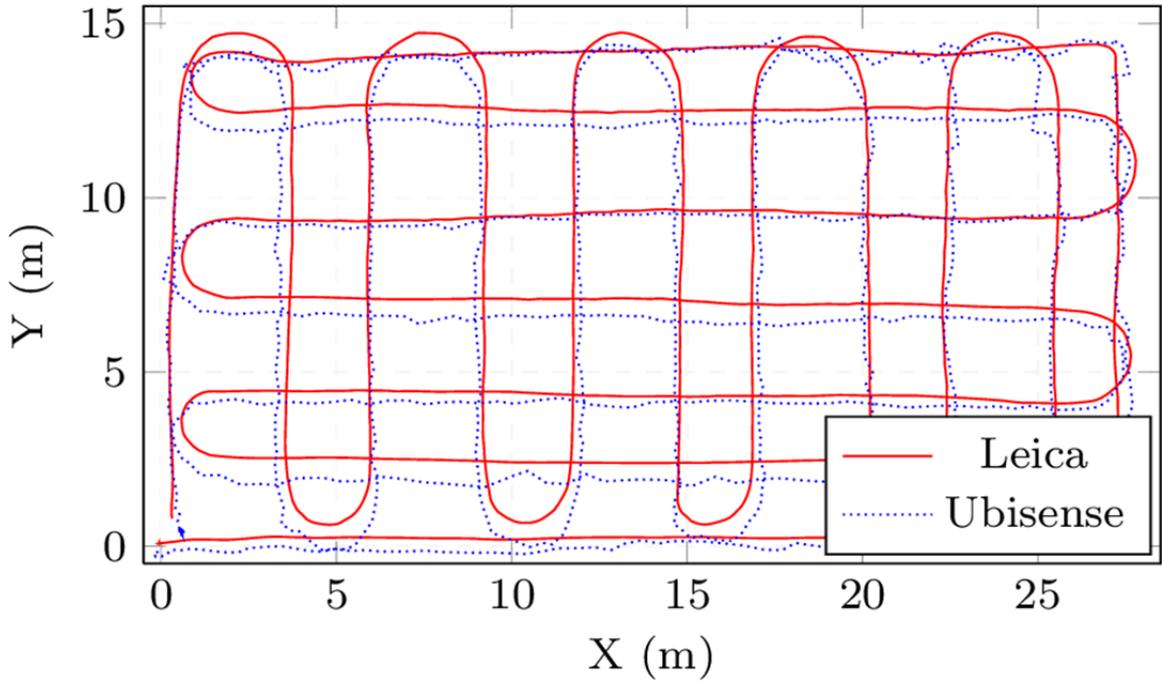
**Figure 4: Tags and prism mounted on a wheelchair**



542

543

Figure 5: Workflow of the processing used in this study

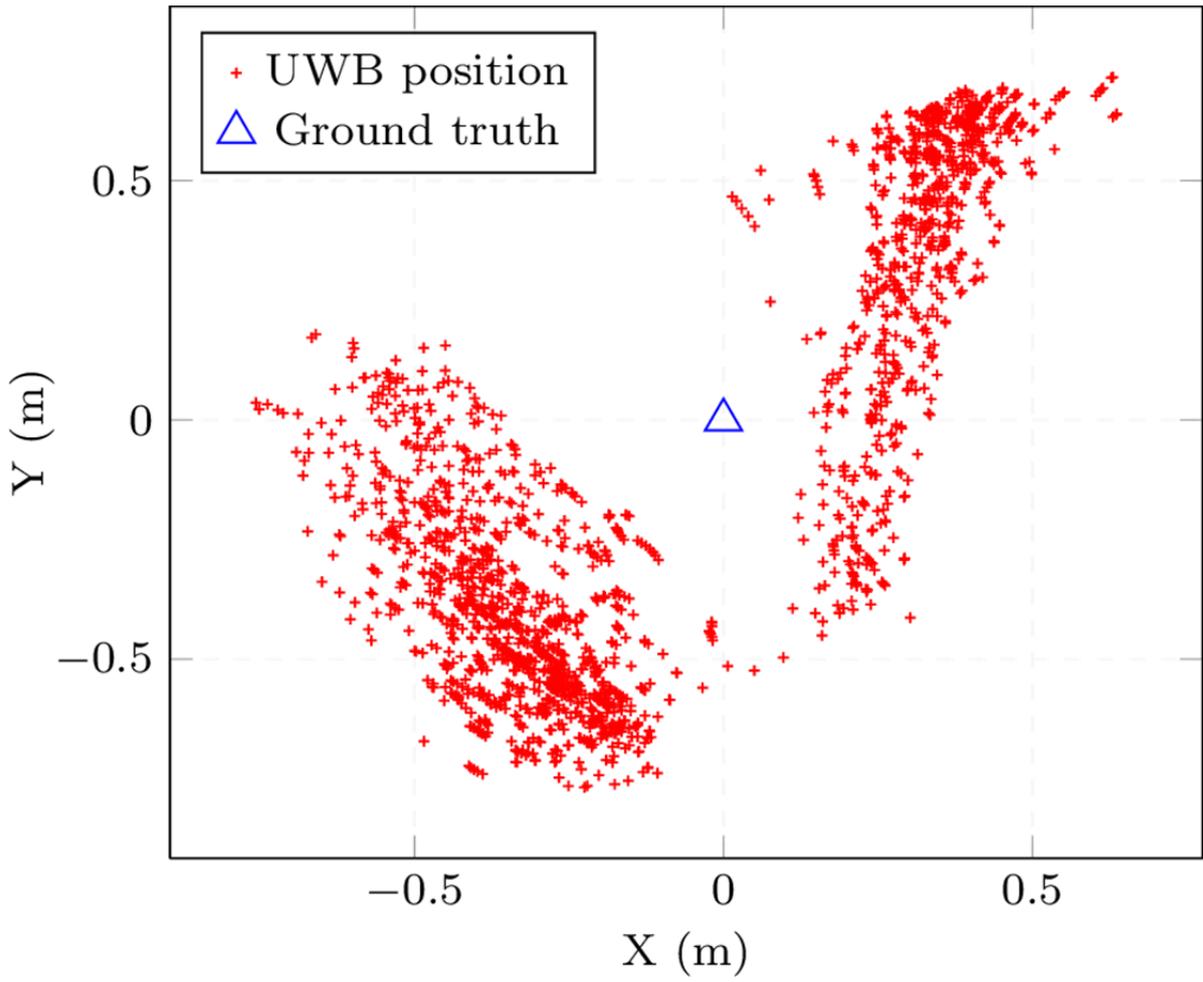


544

545

**Figure 6: Trajectory pattern used for sensors spatial configuration analysis**

546



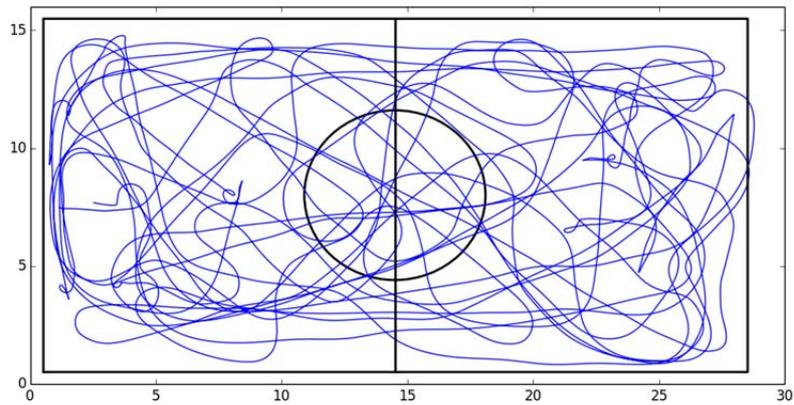
547

548

549

550

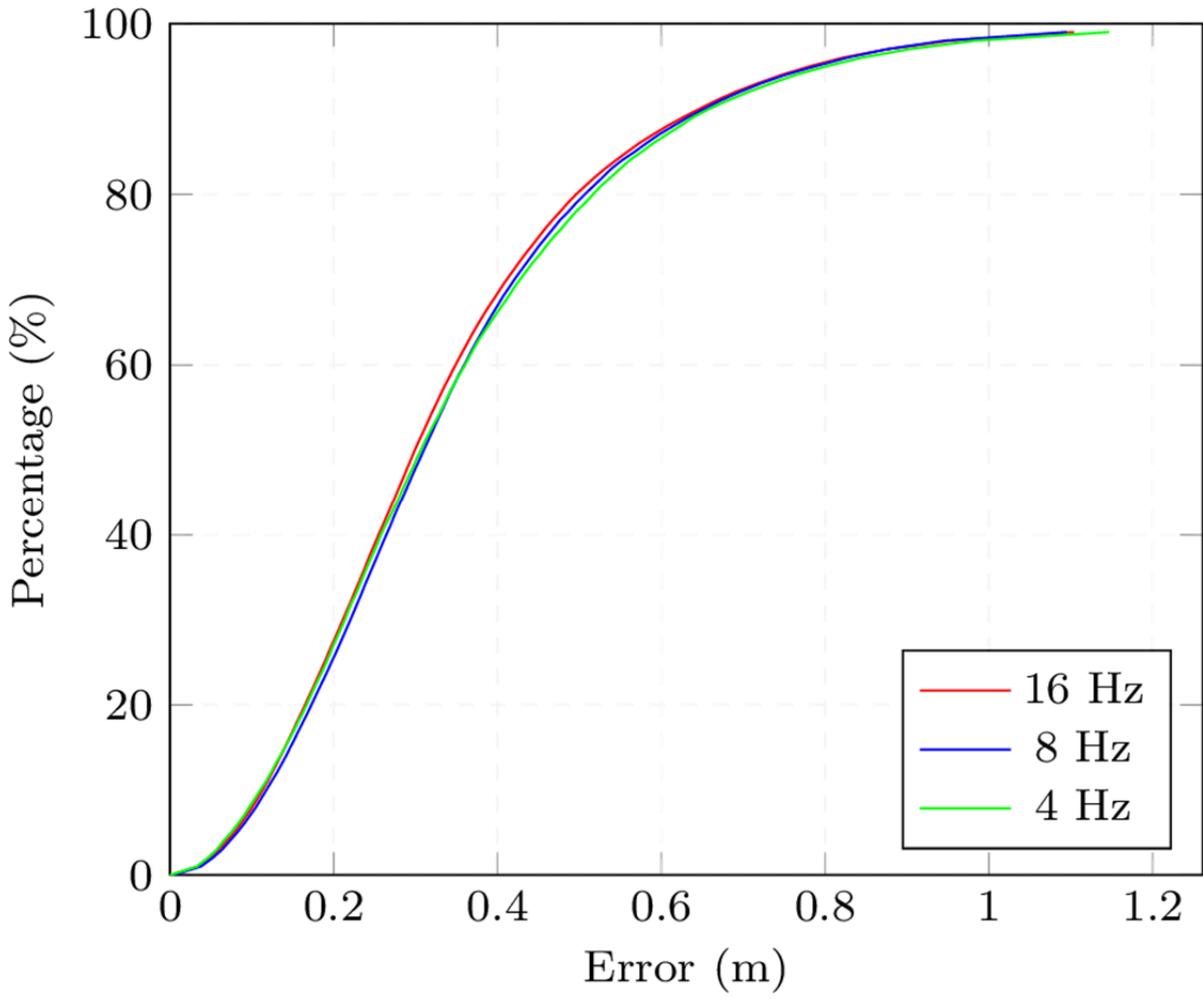
Figure 7: Stationary tag positioning



551

552

Figure 8: Example trajectory during a quarter of a simulated match

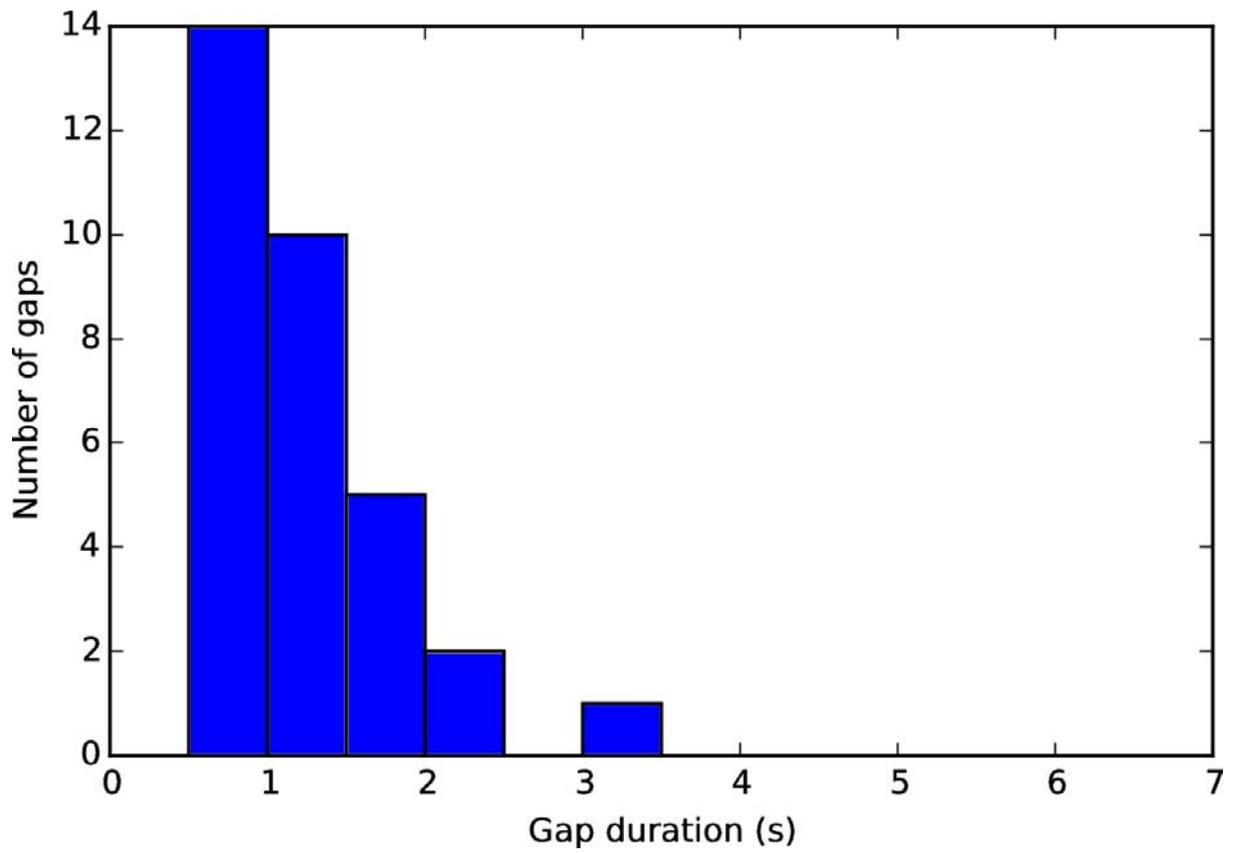


553

554

**Figure 9: Cumulative Distribution Function of the positioning error**

555

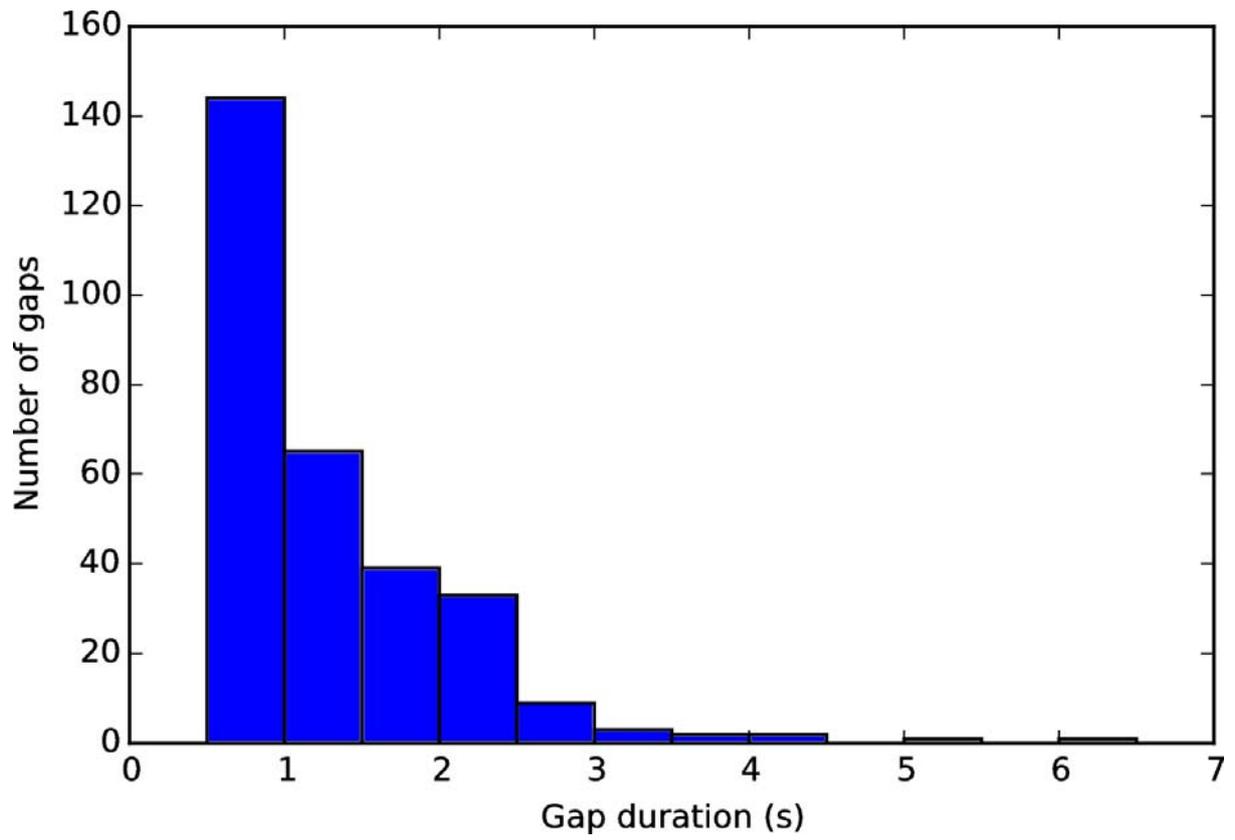


556

557

558

Figure 10: Distribution of the data gaps with one wheelchair on the court



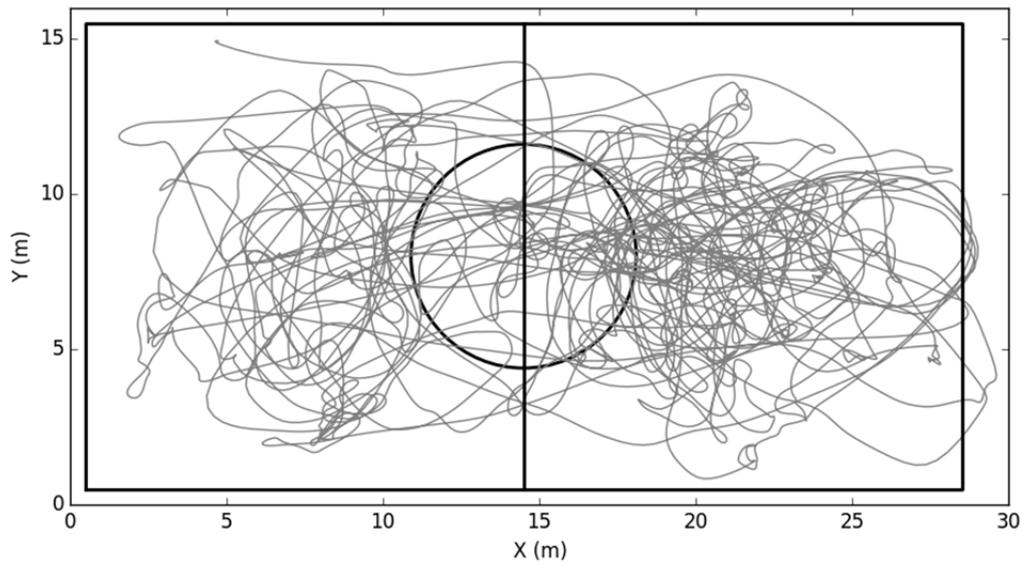
559

560

561

Figure 11: Distribution of the data gaps during a real match with 8 athletes + 1 referee on the court

562



563

564

**Figure 12: Example of an athlete's trajectory during one quarter of a real match**