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Citation for published version:

Valero Rodriguez, E, Bosché, F & Bueno Esposito, M 2022, 'Laser Scanning for BIM', *Journal of Information Technology in Construction*, vol. 27, pp. 486-495. <https://doi.org/10.36680/j.itcon.2022.023>

Digital Object Identifier (DOI):

[10.36680/j.itcon.2022.023](https://doi.org/10.36680/j.itcon.2022.023)

Link:

[Link to publication record in Edinburgh Research Explorer](#)

Document Version:

Peer reviewed version

Published In:

Journal of Information Technology in Construction

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1 LASER SCANNING FOR BIM

2 SUBMITTED: December 2021

3 *Enrique Valero, PhD, Post-Doctoral Associate,*
4 *Institute for Infrastructure and Environment, School of Engineering, The University of Edinburgh;*
5 e.valero@ed.ac.uk

6 *Frédéric Bosché, PhD, Senior Lecturer,*
7 *Institute for Infrastructure and Environment, School of Engineering, The University of Edinburgh;*
8 f.bosche@ed.ac.uk

9 *Martin Bueno, PhD, Post-Doctoral Associate,*
10 *Institute for Infrastructure and Environment, School of Engineering, The University of Edinburgh;*
11 martin.bueno@ed.ac.uk

12 **SUMMARY:** *Obtaining useful data from reality capture devices, such as Terrestrial Laser Scanners (TLS), for*
13 *the extraction of semantic information and its subsequent use to support Building Information Modelling (BIM)*
14 *use cases (e.g. Scan-to-BIM or Scan-vs-BIM -based use cases) is a complex task that requires planning and*
15 *execution expertise. Point clouds of quality need to be produced following a conscientious planning and execution*
16 *of scanning. And once the point clouds are acquired, methodical pre-processing operations are vital to ensure the*
17 *point clouds finally are of high quality. This paper summarises some guidelines to surveyors for a successful data*
18 *acquisition campaign, especially when these data will be employed for automatic processes involving point clouds*
19 *and BIM, such as Scan-to-BIM or Scan-vs-BIM. The guidelines are also useful to the recipients of the point clouds*
20 *involved in those processes, such as BIM modellers or Quality Control (QC) managers.*

21 **KEYWORDS:** *laser scanning, BIM, best practice.*

23 1. INTRODUCTION: FROM DATA TO INFORMATION

24 The rapid evolution of 3D reality capture technologies, such as Laser Scanning (LS), supports digitalisation in
25 various industries. The output of LS devices is under the form of point clouds, which are unstructured sets of often
26 coloured 3D points. Although raw point clouds may be useful as is (even as pieces of art (Chapman et al., 2017)),
27 these datasets are usually processed to extract meaningful information. In the particular case of the Architecture,
28 Engineering and Construction (AEC) sector, point clouds obtained by LS devices have accelerated the generation
29 of 2D and 3D drawings, are used to measure volumes (Porrás-Amores et al., 2019), detect objects (Dimitrov and
30 Golparvar-Fard, 2015, Maalek et al., 2019, Perez-Perez et al., 2021a, Perez-Perez et al., 2021b), and ultimately
31 produce semantically-rich 3D models (Valero et al., 2016) or Building Information Models (BIM) (Bassier and
32 Vergauwen, 2020, Valero et al., 2021). The process to generate BIMs from point clouds is commonly called *Scan-*
33 *to-BIM*. In general, point clouds provide accurate representations of the scanned environments and are employed
34 by modellers as a reference from which they manually produce semantically-rich BIM models.

35 Another use case of point clouds in the context of Building Information Modelling is to compare them against
36 BIMs to, for example, monitor construction progress (Braun et al., 2020) or construction (or fabrication) quality
37 (Bosché et al., 2009, Bosché and Guenet, 2014, Kim et al., 2016). These comparison use cases fall under the
38 increasingly used umbrella term *Scan-vs-BIM* (Bosché et al., 2014). *Scan-to-BIM* and *Scan-vs-BIM* process use
39 cases collectively illustrate the importance of 3D point clouds for the generation and management of building-
40 related information.

41 Current practice in *Scan-to-BIM* and *Scan-vs-BIM* processes is predominantly manual. While point clouds used
42 as reference for generating BIMs improve modelling quality and efficiency in comparison to the use of single,
43 unstructured measurements from distometers, the modelling part remains a mainly manual process that is tedious,

44 repetitive (and therefore error-prone), and time consuming, with outcomes significantly impacted by the expertise
45 of modellers. Therefore, many research teams have been working to automate steps involved in the modelling
46 phase (Son et al., 2015, Dimitrov and Golparvar-Fard, 2015, Valero et al., 2016, Maalek et al., 2019, Perez-Perez
47 et al., 2021a, Perez-Perez, 2021b, Bassier and Vergauwen, 2020). However, despite great strides in this area, the
48 quality of outputs remains highly dependent on the quality of the input point cloud. Ensuring that the input point
49 clouds are of adequate quality is critical.

50 The acquisition of point clouds with the right quality for effective processing (both manual and automated) is
51 challenging and requires experience and, importantly, knowledge of the study to be performed subsequently. This
52 process of effectively acquiring point clouds to be used for BIM-related purposes is what we call here *Scan4BIM*.

53 In the following, we present ten simple rules for producing high quality point clouds to be used in BIM-related
54 processes such as Scan-to-BIM, Scan-vs-BIM – although the majority of these rules remain broadly relevant when
55 considering other point cloud processing tasks. These rules are grouped into four sections: the device (section 2),
56 the environment (section 3), data acquisition (section 4), and pre-processing (section 5).

57

58 2. THE DEVICE

59 2.1 Rule 1: Type of Scanner

60 Laser scanners are utilised for reality capture purposes and deliver data on the shape (i.e., 3D geometry) and
61 appearance (i.e., colour and texture) of the environment surrounding them. According to the state of the devices
62 during the scanning works, these can be:

- 63 • **Stationary scanners:** these devices are placed at strategic locations in the environment, from which the
64 scans are taken to maximise the documented volume. These can subsequently be classified in two sub-
65 groups (Angelopoulou et al., 1999): 'phase-based' (mid-range, up to around 100m, e.g.,
66 <https://www.faro.com/en/Products/Hardware/Focus-Laser-Scanners>), often used for interiors and facades
67 of not very tall buildings; and 'Time of Flight' (long range, beyond 100m, e.g.,
68 <http://www.riegl.com/nc/products/terrestrial-scanning/produktdetail/product/scanner/48/>), used for
69 larger environments, such as quarries or infrastructure.
- 70 • **Mobile scanners:** these devices are mounted on mobile platforms (a person or a vehicle) that are moving
71 during the scanning process. The methodology behind these devices typically requires further advanced
72 sensing and data processing algorithms (i.e., Simultaneous Localisation and Mapping (SLAM) (Cadena
73 et al., 2016), to deliver the 3D data, and they can be used indoors (e.g., <https://geoslam.com/solutions/zeb-revo-rt/>) or outdoors (e.g., <https://leica-geosystems.com/products/mobile-sensor-platforms/capture-platforms/leica-pegasus-backpack>).

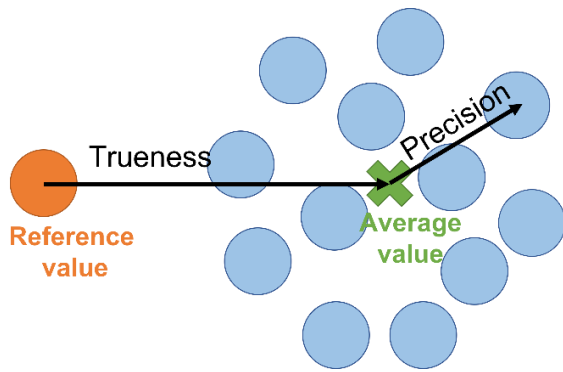
76 The specifications of stationary mid-range scanners have made them the tool of choice for most *Scan4BIM* use
77 cases. However, it is the surveyor who will decide the tool to be employed according to the context, their expertise,
78 and previous experiences.

79

80 2.2 Rule 2: Precision

81 Accuracy, as illustrated in FIG. 1 and described in ISO 5725-1:1994 (ISO, 1994), is a combination of two
82 parameters: trueness and precision. Generally, the trueness of point clouds delivered by Terrestrial Laser Scanning
83 (TLS) devices (or by means of photogrammetric techniques in adequate conditions) is acceptable, meaning that
84 3D coordinates of points are very close, on average, to the right values. Campanelli et al. (2015) have evaluated
85 and compared the accuracy of both low- and high-cost laser scanners. Trueness remains acceptable as long as
86 devices are frequently calibrated, as specified by the device manufacturers. Precision, in contrast, impacts each
87 measurement, and importantly varies from one device to another. So, precision is a crucial parameter to bear in
88 mind when choosing a scanner for *Scan4BIM* use cases.

89



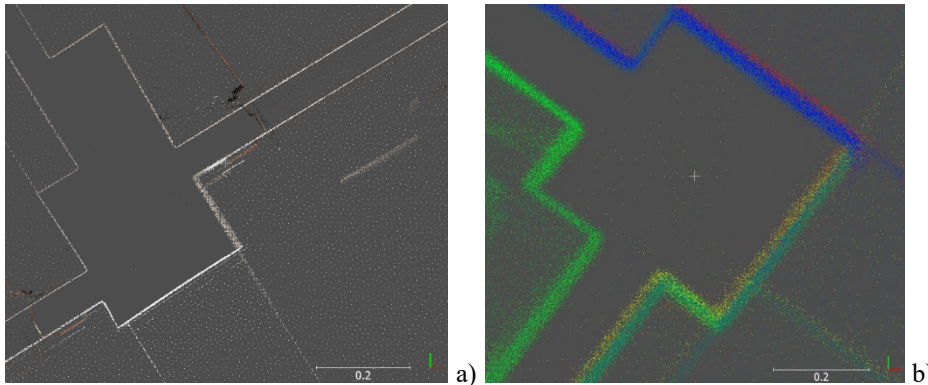
90

91 *FIG. 1: Graphical definition of accuracy*

92

93 **Error! Reference source not found.** illustrates the differences in precision of two point clouds, one delivered by
 94 a Faro Focus S 150 mounted on a tripod (stationary) and placed at key locations, and the other one produced by a
 95 Geoslam ZEB Revo RT (mobile). Although a BIM modeller could approximately visually detect walls or other
 96 structural components in both clouds, modelling, for example, the wall on the right in the GeoSlam cloud is still
 97 challenging to achieve within typical tolerances, and even more so for an automatic algorithm.

98



99

100 *FIG. 2: Top view of the interior walls of a building scanned with two different devices: a) Faro Focus S150. b)*
 101 *Geoslam ZEB Revo RT*

102

103 3. THE ENVIRONMENT

104 3.1 Rule 3: Completeness and Occlusions

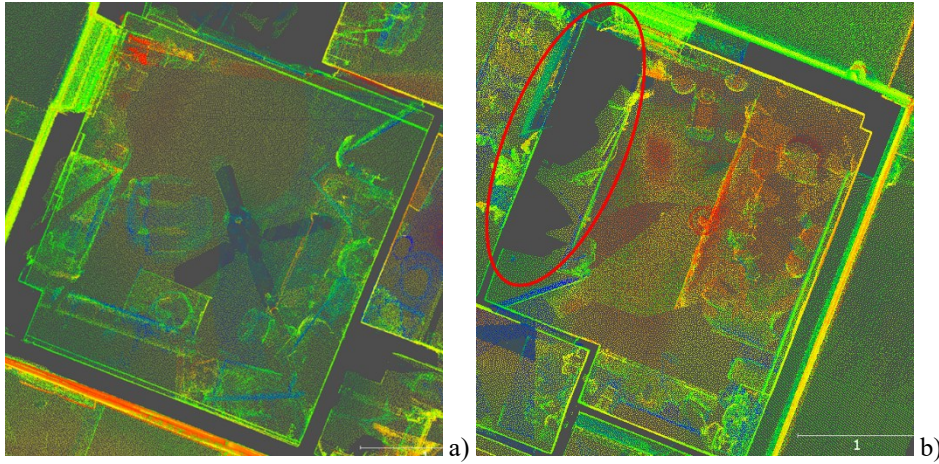
105 *“Gestalt is a theory of perception that describes the manner in which humans perceive the components of an image*
 106 *and organize them into broader structures or interpretations”* (Tait, 2018). Amongst its principles, proximity,
 107 similarity, closure, and continuity are particularly relevant in point cloud analysis. Points that are next to each other
 108 and have similar orientation (i.e., normal vector) are most likely part of the same entity (e.g. plane).

109 Object detection or recognition build on such basic observation and as a result completeness of point clouds is
 110 paramount to successful scan-to-BIM or scan-vs-BIM -based processes. Holes in the data (see FIG. 2b) can at
 111 times be successfully handled by humans thanks to their extensive cognitive capability. In contrast, missing data
 112 can more significantly reduce the effectiveness of algorithms for detecting features or objects (e.g. walls), as shown
 113 in (Adan and Huber, 2010).

114 Delivering (sufficiently) complete point clouds requires detailed planning. Planning for Scanning (P4S) (Aryan et
 115 al., 2021) is the process of identifying the right set of locations from which the target objects can be scanned as

116 required, avoiding occlusions produced by other objects present in the scene, as well as self-occlusions. For
117 example, when scanning the interior of a building, attention should be paid to the impact of furniture on the
118 scanning of walls and openings (see Subsection 3.2 – Rule 4: Openings) when the goal is to create a 3D BIM of
119 that interior environment.

120



121 a) b)
122 FIG. 2: Top view of two scanned spaces, where the point clouds are complete (a) and incomplete (b).
123

124

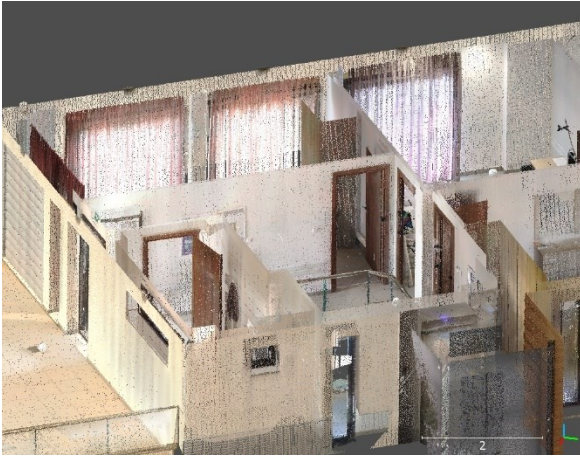
3.2 Rule 4: Openings

125 Openings usually provide access to spaces. Although these are, in general, cleared to allow movement across the
126 building (doors) or let the light in (windows), at times doors and windows can be closed or occluded by other
127 entities. For example, in FIG. 3, while doors are open, curtains are drawn over windows, preventing their easy
128 detection in a point cloud and subsequent modelling. A human may be able to guess that there is likely a window
129 behind a curtain, but they may still not have enough information to model that opening correctly. As can be
130 appreciated in many research works (Díaz-Vilariño et al., 2015, Assi et al., 2019), researchers working on
131 automatic detection of openings usually ensure during data acquisition that doors or windows are not or minimally
132 occluded, although some researchers have developed opening detection algorithms that aim to be more robust to
133 occlusion using various strategies (Quintana et al., 2018, Nikoohemat, 2018).

134 Besides, the presence of closed curtains in a point cloud can result in additional planes that can be confused as
135 wall segments by algorithms. Similarly, blinds, air conditioning units and other objects can affect the size of
136 detected openings (Quintana et al., 2018).

137 Having doors open while scanning not only facilitates the detection of openings, but it also increases the overlap
138 between consecutive scans (see Subsection 4.3 – Rule 8: Overlapping) and, therefore, eases the registration process
139 (see Subsection 5.2 – Rule 10: Registration).

140



141

142 *FIG. 3: Example of windows covered with curtains*

143

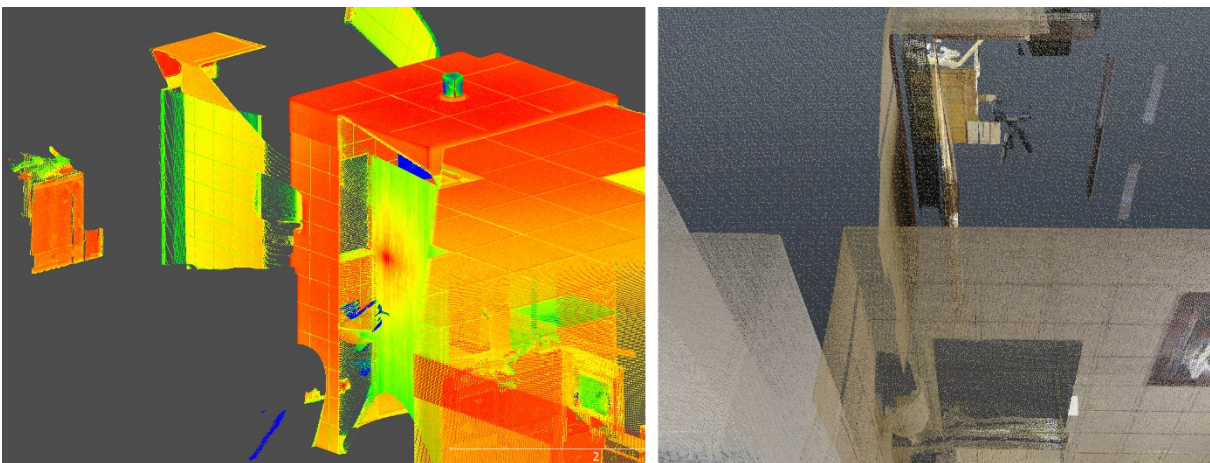
144 **3.3 Rule 5: Mirrors**

145 All the elements reflected in a mirror are considered to be *Through the Looking-Glass* and, as illustrated in FIG.
 146 4, a mirrored scene is added to the real one. These artefacts may also occur with glass, metals and other polished,
 147 reflective surfaces. Some research has been done on the identification of rectangular mirrors to remove erroneous
 148 points (Käshammer and Nuchter, 2015). If such highly reflective objects cannot be removed from the scene or
 149 covered (e.g. with a cloth), the affected point clouds should be removed (Gao et al., 2022) (see Subsection 5.1 –
 150 Rule 9: Cleaning) before subsequent processing, starting with registration.

151 A particular exception to this is when mirrors are actively used to scan the back surfaces of objects alongside their
 152 visible surfaces, as done by (Li and Kim, 2021) to obtain nearly complete point clouds of pre-fabricated concrete
 153 components from a single scan.

154 Note that, in the case of transparent surface, such as windows, the diffraction of laser beams when traveling through
 155 them similarly lead to incorrect measurements. Those points should similarly be removed from the cloud.

156



157

158 *FIG. 4: Effects of mirrors in point clouds*

159

160 4. DATA ACQUISITION

161 4.1 Rule 6: Resolution

162 One of the parameters to be selected before starting a scan is the resolution. Resolution is defined by two angles
163 that are the horizontal and vertical angular intervals between successive scanned points. Resolution is also
164 commonly expressed as the distance between consecutive points in the cloud at a given distance from the scanner,
165 such as “*x millimetres @ y metres*” (Faro, 2020). Note that the same value is commonly used for the horizontal
166 and vertical scanning resolutions.

167 Higher resolution angles result in sparser point clouds. This can enable accurate scanning of smaller object, but it
168 can also negatively impact data processing performance. To ensure adequate resolution of a given target object
169 (i.e. enough but not too high), the surveyor needs to know or estimate the scanning distance to select the correct
170 resolution settings. But, it must also be highlighted that the incidence angle of the laser beam on the scanned
171 surface also affects the resolution of the obtained point cloud. The lower incidence angle (i.e. the scanning direction
172 is more perpendicular to the scanned surface), the higher the resolution. Surveyors may also have to consider this,
173 for example when scanning structures that are high above ground, such as upper parts of high-rise buildings.

174 4.2 Rule 7: Colour

175 The surveyor needs to know if colour will be required for further processes. This is because colour acquisition can
176 slow overall data acquisition, depending on the scanning technology used.

177 Colour acquisition is done from the scanning device with reasonable, but not necessarily, high-quality cameras. If
178 high-quality imagery is required, then scanning may need to be supplemented with additional image acquisition
179 and the colour information transferred to the point cloud using texture mapping, e.g., through alignment with
180 photogrammetric reconstructions (Alshawabkeh et al., 2021, Valero et al., 2019). But this process is both time-
181 consuming and challenging.

182

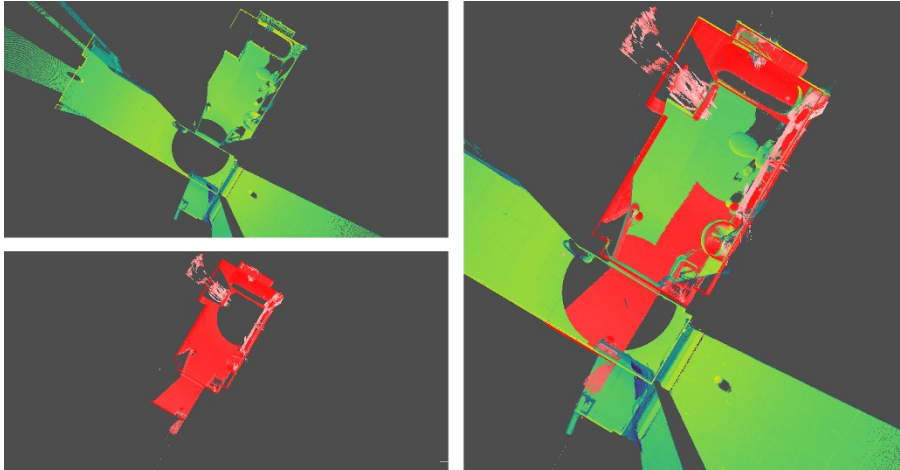
183 4.3 Rule 8: Overlapping

184 As mentioned in Subsection 3.1 – Rule 3: Completeness and Occlusions, a well-designed plan for scanning an
185 environment is crucial to obtain a complete point cloud of that environment. But, the selected scanning locations
186 must ensure not only that the target objects are scanned with the right levels of quality and completeness, but also
187 that the resulting scans can be effectively registered together in a unified point cloud (see Subsection 5.2 – Rule
188 10: Registration). This requires adequate overlap between consecutive clouds, as considered by many research
189 works (Ahn and Wohn, 2016, Chen et al., 2018, Huang et al., 2021, Li et al., 2020, Aryan et al., 2021).

190 As illustrated in FIG. 5, when scanning indoors, placing the scanner in doorways (see green point cloud in FIG.
191 5), enables capturing data from the two connected spaces and facilitates the co-registration of other point clouds
192 acquired in those two adjacent spaces and beyond.

193 Outdoors, when scanning a building envelope, it is important to connect the individual clouds representing the
194 facades. When using stationary scanning devices, placing the device at the corners of the building will deliver a
195 cloud containing data from two (or more) connected facades, which will enable co-registering other point clouds
196 acquired of those facades.

197



198

199 *FIG. 5: Overlapping point clouds. Advantages of placing a device in a doorway.*

200

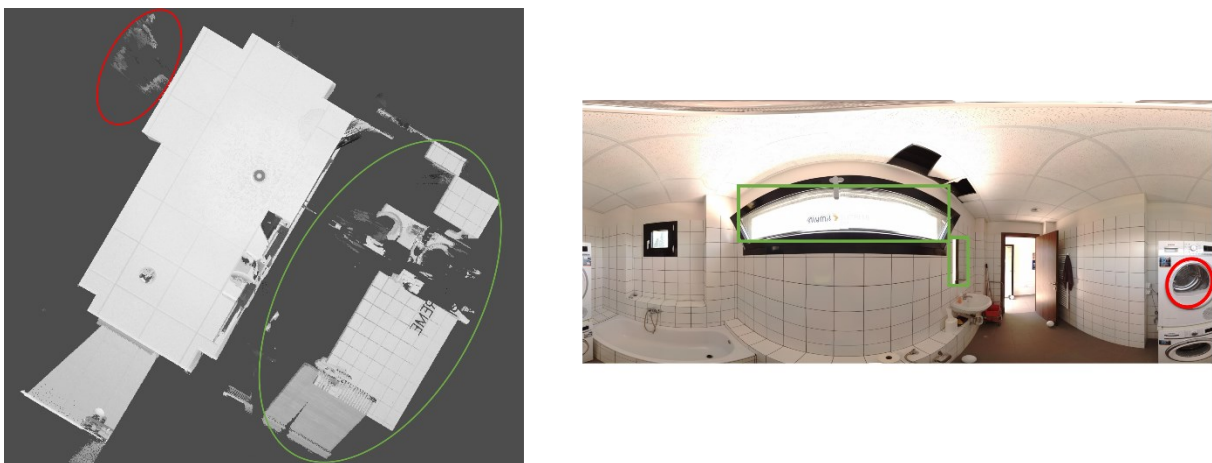
201 **5. PRE-PROCESSING**

202 **5.1 Rule 9: Cleaning**

203 As mentioned in Subsection 3.3 – Rule 5: Mirrors, reflecting and transparent surfaces will introduce spurious
 204 points to a cloud. FIG. 6 shows the case of a scanned bathroom where objects were scanned through reflection on
 205 glass and plastic surfaces, creating 'ghost' 3D points (Gao et al., 2022). Besides, the scene may contain moving
 206 objects, e.g. people, at the time of scanning. These points should be removed as much as possible before co-
 207 registering the individual scans to avoid confusion, especially when running automatic procedures (Cheng et al.,
 208 2021, Hang et al., 2017).

209 Additionally, points corresponding to objects that are not the subject of the study should ideally be removed to
 210 improve processing performance (in terms of both time and quality). Such cleaning effort may involve human
 211 intervention, which may itself be time-consuming and prone to error. The amount of cleaning should thus be
 212 assessed using a cost/benefit analysis.

213



214

215 *FIG. 6: Incorrect points produced by plastic and glass surfaces. The washing machine door introduces incorrect*
 216 *points (highlighted in red), and a window and a mirror on the opposite wall produce a similar effect (highlighted*
 217 *in green).*

218

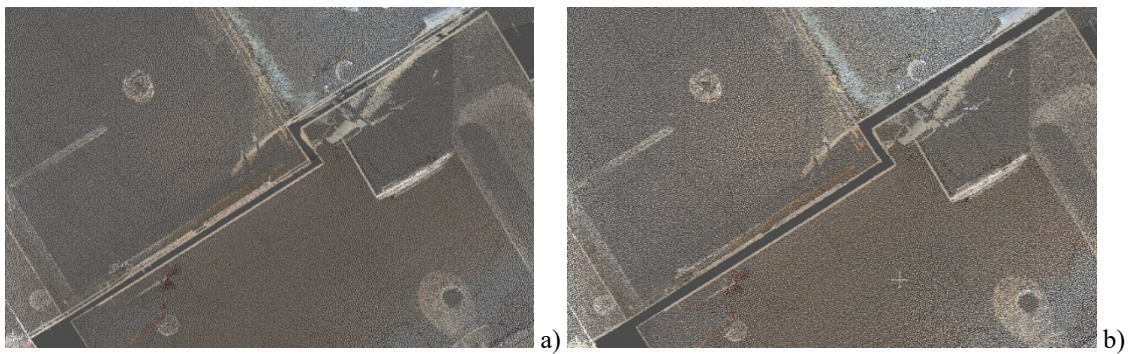
219 **5.2 Rule 10: Registration**

220 A careful registration of consecutive point clouds is crucial to produce a complete cloud that accurately represents
221 the scanned environment. Meticulousness is important in this process, especially when the point cloud is to be
222 used in automatic processes. Indeed, registration errors can result in misalignment that can impact processing far
223 more than single point precision and accuracy – e.g. overlapping scans of a wall that are correctly co-registered
224 will result in two close but distinct planes. Although most software packages devoted to handling point clouds
225 (e.g., Faro Scene <https://www.faro.com/en/Resource-Library/Tech-Sheet/techsheet-faro-scene>, Cyclone
226 <https://leica-geosystems.com/products/laser-scanners/software/leica-cyclone>, Autodesk ReCap
227 <https://www.autodesk.co.uk/products/recap>) can automatically robustly register point clouds through natural and
228 artificial features as well as internal sensor data (e.g., IMU) (Ridene et al., 2013), it is important to verify the
229 results, including by checking that overlapping scans of elements of interest (e.g., walls) are properly aligned
230 (Mora et al., 2021).

231 If the automatic registration of consecutive clouds is not satisfactory (see walls in **Error! Reference source not**
232 **found.a**), a manual registration should be carried out, by selecting at least three (but preferably four or more) pairs
233 of matching points as in (Aiger et al., 2008, Li et al., 2021). **Error! Reference source not found.b** shows an
234 example of the manual correction of mis-registration. Note that geometric features other than points can also be
235 used, e.g. planes (Förstner and Khoshelham, 2017, Kim et al., 2018, Bueno et al., 2018).

236 Note that geometric features used for registration (e.g. points) should be located widely throughout the space and
237 should not lie on the same plane. This makes the registration process more robust.

238



239

240 FIG. 8: Consecutive point clouds registered (a) automatically and (b) manually.

241

242 **6. CONCLUSIONS**

243 This paper presented ten simple rules to surveyors to deliver high quality point clouds to be used in BIM-related
244 processes (e.g., Scan-to-BIM, Scan-vs-BIM). The guidelines are also useful to the recipients of the point clouds
245 involved in those processes, such as BIM modellers or Quality Control (QC) managers. These rules are especially
246 useful when automatic processes are applied to the clouds, because the performance of algorithms can be
247 significantly improved by ensuring that the acquired input clouds are complete and accurate.

248 Although the recommendations summarised in this manuscript can assist in the generation of point clouds for
249 various *Scan4BIM* use cases, it should be emphasised that a good communication of the requirements between the
250 end user (e.g. BIM modeller or QC manager) and the surveyor is very important. Laser scanning is a tailored
251 process, which presents a unique and challenging task with every new environment (e.g. building, infrastructure)
252 and where experience plays an important role alongside the application of best practice.

253

254 **7. ACKNOWLEDGEMENTS**

255 This work was conducted as part of the BIMERR project that received funding from the European Commission's
256 Horizon 2020 research and innovation programme under grant agreement No 820621. The views and opinions
257 expressed in this article are those of the authors and do not necessarily reflect any official policy or position of the
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260

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