

---

# From Traditional Robotic Deployments towards Assisted Robotic Deployments in Nuclear Decommissioning

Erwin Jose Lopez Pulgarin<sup>1,\*</sup>, Dave Hopper<sup>2</sup>, Jon Montgomerie<sup>2</sup>, James Kell<sup>2</sup>,  
Joaquin Carrasco<sup>1</sup>, Guido Herrmann<sup>1</sup>, Alexander Lanzon<sup>1</sup>, Barry Lennox<sup>1</sup>

<sup>1</sup>*Department of Electrical and Electronics Engineering, The University of  
Manchester, Manchester, UK*

<sup>2</sup>*Jacobs, Warrington, UK*

Correspondence\*:  
Corresponding Author  
erwin.lopezpulgarin@manchester.ac.uk

## 2 ABSTRACT

3 The history around teleoperation and deployment of robotic systems in constrained and  
4 dangerous environments such as nuclear is a long and successful one. From the 1940s,  
5 robotic manipulators have been used to manipulate dangerous substances and enable work in  
6 environments either too dangerous or impossible to be operated by human operators. Through  
7 the decades, technical and scientific advances have improved the capabilities of these devices,  
8 whilst allowing for more tasks to be performed. In the case of nuclear decommissioning, using  
9 such devices for remote inspection and remote handling has become the only solution to work  
10 and survey some areas. Such applications deal with challenging environments due to space  
11 constraints, lack of up-to-date structural knowledge of the environment and poor visibility, requiring  
12 much training and planning to succeed. There is a growing need to speed these deployment  
13 processes and to increase the number of decommissioning activities whilst maintaining high levels  
14 of safety and performance. Considering the large number of research and innovation being done  
15 around improving robotic capabilities, numerous potential benefits could be made by translating  
16 them to the nuclear decommissioning use cases. We believe such innovations, in particular  
17 improved feedback mechanisms from the environment during training and deployments (i.e.,  
18 Haptic Digital Twins) and higher modes of assisted or supervised control (i.e., Semi-autonomous  
19 operation) can play a large role. We list some of the best practices currently being followed in the  
20 industry around teleoperation and robotic deployments and the potential benefits of implementing  
21 the aforementioned innovations.

22 **Keywords:** teleoperation, robotics, robot deployment, haptic digital twin, semi-autonomy, training

## 1 INTRODUCTION

23 The use of robots in the nuclear industry has a long and rich history (Bogue, 2011), going from teleoperated  
24 serial manipulator mechanisms to some of the latest research in the use of Remotely Operated Vehicles  
25 (ROV) and mobile robots (Tsitsimpelis et al., 2019) for robotic inspection. Most devices referred to as  
26 robots for the nuclear industry are used to perform tasks where human presence is either limited or not  
27 possible due to environmental factors such as radioactive hazards. Some of their uses happen in different  
28 stages of a reactor's lifecycle, including its commissioning and construction, during maintenance operations,

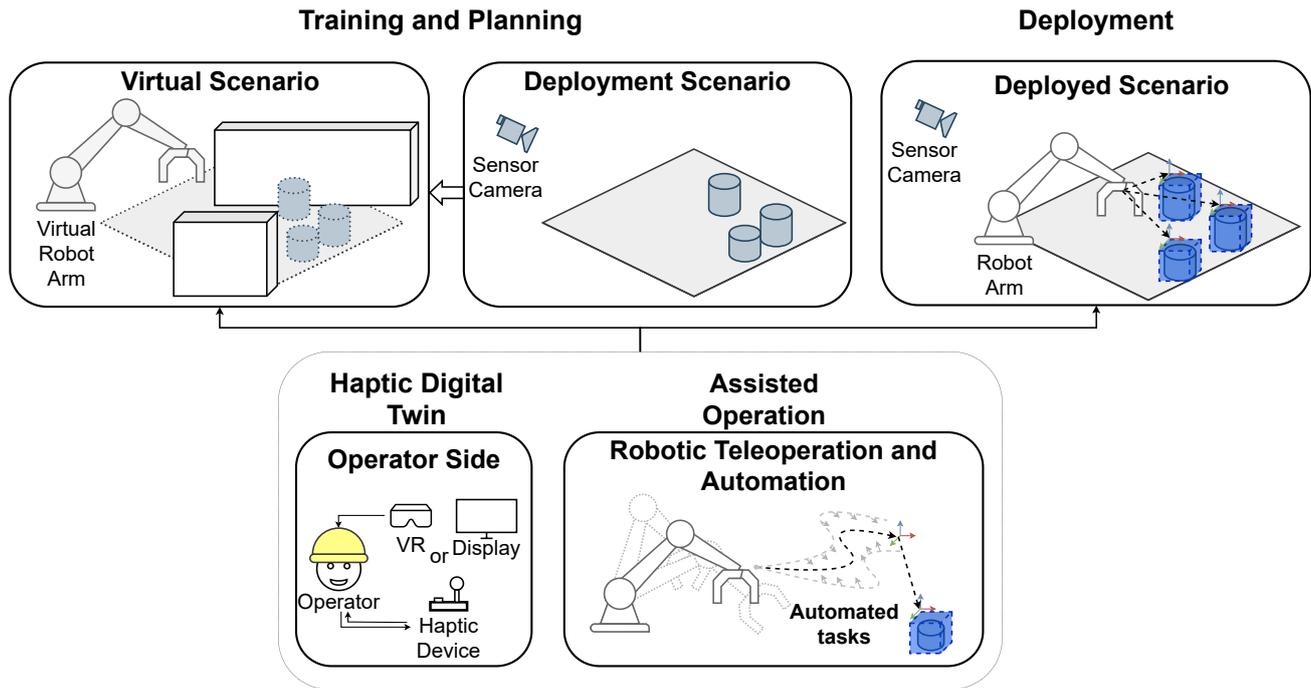
29 waste disposal services and during its decommissioning. The benefits of such technologies are still being  
30 explored, and are maturing into products ready for deployments in nuclear sites such as Sellafield (UK)  
31 (Sellafield, 2023) and Fukushima Daiichi (Japan) (Tugal et al., 2022; Zhang et al., 2025).

32 Beyond mechanical manipulators and exploratory vehicles, other technologies and robots could help  
33 with tasks that are still being performed manually during any of the reactor's lifecycle stages. Maintenance  
34 and decommissioning tasks are a clear example, as they involve expert human operators performing tasks  
35 whilst wearing protective outfits or using gloveboxes to manipulate dangerous substances in an isolated  
36 environment. The use of different robots such as teleoperated robotic manipulators for decommissioning can  
37 bring safety and operational benefits, such as for operations currently being performed inside gloveboxes  
38 (Tokatli et al., 2021). However, many challenges remain for widespread adoption of these tools, and  
39 particularly to reach the task performance levels that an expert human operator can achieve (Pulgarin  
40 et al., 2022). Regardless of the growing need to have larger capacity to perform tasks at a faster pace, the  
41 demanding requirements around safety and performance (i.e., repeatability, accuracy), slow down and  
42 increase costs of producing tested and certifiable platforms regardless of the technology.

43 The term digital twin has gained popularity in the recent decade (Cryer et al., 2023), as it encompasses  
44 many technologies used to sense, store, display and manipulate information related to a remote asset (Cryer  
45 et al., 2023). Such systems are designed to mimic the remote assets, and tend to include simulation and  
46 visualization capabilities that allow for greater control of the process (Tu et al., 2023; Cryer et al., 2023).  
47 Feedback from and towards the digital twin can come from different modalities (e.g., visual, sound), being  
48 haptics and/or force (Elsner et al., 2022) one with large potential and interest. The creation of Haptic Digital  
49 Twins (HDT) that allow for realistic representations of remote sites, either as offline mock-ups or online  
50 representations, could be used during training and deployments in the context of nuclear decommissioning.

51 The large interest around increased levels of autonomy (i.e., robots adapting to environmental changes  
52 to fulfil a goal with little to no human input) enables further opportunities for imbuing robots with the  
53 capabilities to operate in constrained and complex environments. A popular example of autonomy in  
54 highly complex systems is the one for autonomous vehicles (Wang et al., 2020), where parts of the driving  
55 task are performed by the vehicle itself with only supervisory input from the driver. In the context of  
56 remotely controlling or interacting with a robot, any assistance provided that eases the control of the robot  
57 or improves any operational performance metric (Li et al., 2023a) (i.e., Assisted Operations (AO)) can  
58 prove beneficial. Such levels of autonomy and assistance require human input (Li et al., 2023b; Pruks and  
59 Ryu, 2022) for all cases where full autonomy is not possible, which for the current state-of-the-art includes  
60 many industrial use cases. The use of Assisted Operations for teleoperated robotic control could improve  
61 task performance to expert operator-level without exposing operators to hazardous environments.

62 This perspective paper introduces our approach to achieve Assisted Robotic Deployments. We strongly  
63 believe that improved multimodal feedback mechanisms (i.e., Haptic Digital Twins (HDT)) and adaptive  
64 modes of assisted or supervised control (i.e., Assisted Operation (AO)) can play a large role in improving  
65 nuclear decommissioning operations. HDTs can be used as training and deployment platforms, replicating  
66 real-world deployment scenarios with realistic haptic-enabled visualizations; AOs are used to automate  
67 tasks to be performed in the deployment environment and improve teleoperation. A visual description of  
68 these can be seen in Figure 1. Such Assisted Robotic Deployments would produce safe and performing  
69 deployments, increasing capacity and reducing deployment times. The remaining of the paper is organized  
70 as follows: Section 2.1 and 2.2 include the technical challenges behind deploying robotic devices for  
71 nuclear decommissioning and best practices in the industry. Section 2.3 builds on the previously mentioned



**Figure 1.** High-level depiction of Haptic Digital Twins (HDT) and Assisted Operations (AO) through different stages of a robotic deployment.

72 technical challenges to discuss how HDT and AO can play a role in improving overall operation. Section 3  
 73 discusses current limitations and research aims.

## 2 ROBOTIC DEPLOYMENTS IN THE NUCLEAR INDUSTRY

### 74 2.1 Complexities of robotic deployments in nuclear decommissioning

75 The use of remote handling systems, particularly robotic ones, is beneficial to nuclear decommissioning  
 76 in several ways. Primarily, robotics enables to perform tasks that simply may not be possible with any other  
 77 means, since the target may be in a location too difficult or hazardous to reach with manned access. An  
 78 additional significant benefit is the reduction in exposure to harmful radiation to operators who might have  
 79 had to enter an environment if robotic alternatives did not exist.

80 Regardless of its potential benefits, the application of robotic systems in this environment is not simple.  
 81 There are considerable challenges to overcome, both physical and regulatory, for any deployment to be  
 82 successful. Arguably the biggest issue facing the system, and particularly for any operator, is the nature  
 83 of the environment. Knowledge about the environment is crucial to scope the hardware to be deployed,  
 84 and to perform all necessary planning. It is common that historical information in the form of design  
 85 documentation and historical records is incomplete or not up to date. Often the challenge is confirming  
 86 that what is inside a remote cell is what is expected – and it is very common that this is not the case. This  
 87 uncertainty inevitably leads to additional work required to overcome the gaps in knowledge. Some studies  
 88 suggest (Brotherhood et al., 2022) that the level of uncertainty in remote cells can reach up to 300% when  
 89 comparing preliminary decommissioning plans and the actual work required.

90 To understand and agree on the concept of operations for a decommissioning exercise, the most up-to-date  
 91 knowledge of the target area is required. Therefore, the first step is to perform an initial survey of the

92 area to acquire data that will fill in the gaps in knowledge. If attempting to decommission using a robotic  
93 system, this might suggest the use of a specific robot that is equipped with a range of scanners. Such  
94 scanners include visual cameras, thermal cameras, chemical analysis, radiation sensors, LiDAR, and  
95 photogrammetry for geometric shape reconstruction. The acquisition of these data sources – by whatever  
96 means – and subsequent compilation onto a single existence or map of the environment can provide the  
97 operator with the information required to start planning a decommissioning task. This plan will then  
98 lead to the collation of requirements and ultimately a specification of a robotic system that will meet the  
99 decommissioning goals of the site owner. Given the unstructured nature of the decommissioning target, it  
100 is very unlikely that off the shelf robotic systems will be able to meet the specification. This may lead to  
101 the customisation of a system or possibly a bespoke design and build to meet the needs.

102 It should not be understated that the skill and experience of the team behind the design and deployment  
103 of such a robotic solution for nuclear decommissioning is of utmost importance. The ability to draw from  
104 prior examples of interventions that went well — and crucially, what to avoid — is of significant benefit  
105 to all involved and often is the difference between a successful application or otherwise. These operators  
106 can make sense of multiple, often suboptimal data sources such as poor-quality video feeds, or having  
107 to control multiple manipulators simultaneously and still completing the task required in the majority of  
108 instances.

## 109 2.2 Best practices for robotic deployments in nuclear decommissioning

110 Considering the previous statement around how challenging the environment is during nuclear  
111 decommissioning, any strategy around robotic deployment should limit time spent inside the cell,  
112 but particularly ensure benign and retrievable systems. This can be explained as ensuring that during  
113 deployment, the decommissioning task does not become more difficult to perform, or additional waste is  
114 not produced in the form of a robotic system that cannot be retrieved. Rule of thumb concepts such as  
115 “do not make the situation worse” and “do not get stuck” apply during any deployment, regardless of its  
116 complexity.

117 Professional teams take advantage of various formal tools to ensure the successful application of robotics  
118 in nuclear environments, such as:

- 119 • **The Hazard and Operability process:** The HazOp process is a structured and systematic examination  
120 of a complex system, such as a site listed for decommissioning. This process enables the user to  
121 identify hazards to personnel, equipment, or the environment, as well as operability issues that may  
122 affect the operations efficiency – with a primary focus on safety. The International Electrotechnical  
123 Commission (IEC) published an application guide (Standard and IEC61882, 2001) that provides a  
124 framework for operation, and the application of this process can provide a good start to planning any  
125 activity.
- 126 • **Systems Engineering:** This provides a structured set of desk-based tools for the collection and  
127 understanding of requirements and subsequent specifications for devices to be built and used. It  
128 provides a way of ensuring that all interdisciplinary stakeholders are engaged and aware of the  
129 intentions of a robotic application.
- 130 • **Physical mock-ups:** These mock-ups are an effective way of providing a safe test area for the training  
131 and development of a robotic system. They can also help to describe or explain both the concept  
132 of operation and the safety case to the asset owner and the regulatory body. The obvious benefit to

133 physical mock-ups is the additional views that an operator could be provided, which would not be  
134 available in the real application, but are nevertheless of use for development purposes.

- 135 • **Fault Tree Analysis:** This is a type of failure analysis that examines the possibilities of what might  
136 result from an undesirable fault. This is a structured approach to understanding the logic leading up to  
137 a failure. From this analysis, mitigation strategies can be put in place to either prevent the failure from  
138 occurring, or at least to reduce the impact of such a failure taking place. In all instances, the design  
139 needs to adhere to As Low as Reasonably Possible (ALARP) principles (Hurst et al., 2019), which  
140 involves weighing a risk against the trouble, time, and money needed to control it.
- 141 • **Safety Case:** A safety case (ONR, 2020) should include all documentation that demonstrates high  
142 standards of nuclear safety and radioactive waste management to satisfy both the asset owner and the  
143 regulatory body. It facilitates relevant discussions, as it captures useful information for the design and  
144 deployment of robotic systems.

### 145 2.3 Assisted Training and Modes of Operation for Deployments

146 Digital twins (Cryer et al., 2023), and in particular Haptic Digital Twins (HDT) can have a large  
147 role in providing assistance in robotic deployment at both training and remote deployment stages. An  
148 HDT to assist remote deployments would integrate live sensor data from different sources, such as 3D  
149 cameras and radiation sensors, for its visualization (see Figure 1). It would integrate a physics-based robot  
150 simulations with kinematic constraints replicating and allowing to predict the robot whilst interacting with  
151 the environment (Tu et al., 2023). It would provide the capability to remotely control the robot and to  
152 provide haptic feedback (Pruks and Ryu, 2022) related to the state of the robot and its operation (e.g.,  
153 robot motion and collisions with the environment). Such HDTs would take advantage of modern computer  
154 graphics technologies such as Neural Radiance Fields (NeRF) (Tancik et al., 2023) and Gaussian Splatting  
155 (Kerbl et al., 2023); these allow to use visual and spatial information (e.g., 2D from cameras and 3D point  
156 clouds) to create 3D representations of spaces with high-definition visualizations, including its surface  
157 information (Yu et al., 2022) and the reconstructed geometry (Millane et al., 2024). Due to the availability  
158 of sensors that produce visual and spatial information with capabilities suited for different lightings and  
159 operation conditions (e.g., Global Shutter, Polarizer, stereo or time-of-flight depth perception) obtaining  
160 high-quality visual and spatial information is achievable. Modern computing with dedicated Graphical  
161 Computing Units (GPU), high-speed sensor interfaces and optimized physic engines for robotics are  
162 needed to make use of NeRF and haptic rendering techniques for visuo-haptic rendering. Visually rich,  
163 haptic-enabled representations of a real decommissioning environment would make the HDT into a useful  
164 tool for robotic deployment activities.

165 There are benefits of using HDTs for training and deployment beyond having a more complete and  
166 realistic user interface (Tugal et al., 2023). An HDT could enhance the current practice of using physical  
167 mock-ups for extensive operator training under a specific deployment plan. Operators require training in  
168 a representative and realistic environment, using the same or similar robotic system to be deployed (i.e.,  
169 Graphical User Interface, local control device and remote robot). However, there are usually only one or  
170 two sets of the robotic system to deploy, as these systems are costly and time-consuming to commission. By  
171 using HDTs for training operators, we can use the same interface designed for deployments and training,  
172 whilst creating realistic environments without the need for physical mock-ups.

173 Creating Assisted Operations (AO) to improve task performance (i.e., any metrics of success that  
174 measures task completion and quality) can be realised in many ways, based on the level of autonomy  
175 given to the remote robot. Initial levels of assisted operation include constrained Cartesian motion and

176 velocity compensation, which allow for motion per axis or motion on a plane defined in space, either  
177 whilst holding or changing the orientation of the robot's end-effector. Additional levels of assistance would  
178 include automatic collision avoidance between all the parts of the remote robot and the environment,  
179 removing the robot from dangerous configurations whilst providing useful feedback to the user about the  
180 robot's motion. Automated actions would add another level of assistance, by including pre-programmed  
181 motions (e.g., open entry hatch, pick-up tool, return to initial position, scan glovebox floor) or, sequences of  
182 motions starting from a relative position (e.g., move around a specific geometry or object in the scene, bring  
183 objects to a desired position) (see Figure 1); such actions would aid during teleoperation when complete  
184 sensor feedback from the environment is not possible, which is a common occurrence in a constrained and  
185 hazardous environment such as a glovebox. Using AOs would expand the current practice of using Fault  
186 Tree Analysis, by explicitly avoiding faulty states or state combinations (e.g., collisions), and would help  
187 create the Safety Case by automated testing of the low risk achieved in the fault analysis.

188 A more advanced level of AOs would integrate all previous assistance modes, allowing for manual or  
189 automatic switch between assistance modes and full manual teleoperation. Such levels of assistance are  
190 called shared control in the literature, acting at either the control level of the robot or at the user input and  
191 feedback level (i.e., haptic guidance (Li et al., 2023a)). Achieving these levels of assistance requires sensor  
192 feedback from the robot and its environment, similar to the one needed for HDTs. However, the main  
193 challenge lies in dynamically computing safe trajectories and control policies to handle object grasping,  
194 tool handling and other tasks, considering an operator controlling or supervising the whole operation. This  
195 is still under research, earning it different names, including semi-autonomy, shared autonomy, and shared  
196 teleoperation among others (Elsner et al., 2022; Pruks and Ryu, 2022; Li et al., 2023b). The final objective  
197 of any AOs should be to either improve task performance directly with better control capabilities, or to  
198 positively influence this metric by providing more intuitive and configurable capabilities to the operator.

### 3 DISCUSSION

199 There are many benefits to nuclear decommissioning deployments if embracing increased levels of digital  
200 technology, among them improved overall safety and performance. However, challenges remain. For  
201 instance, adding technology such as Haptic Digital Twins and Assisted modes of Operation to robotic  
202 systems would add complexity to the system, making its management and validation harder. Furthermore,  
203 training operators to use such systems would require updating traditional training methodologies to include  
204 these new technologies; as HDTs and AO are designed with training in mind, this could potentially  
205 reduce entry-level requirements to operate a robotic system. For such sophisticated and complex systems,  
206 simplicity of use and having user interfaces that make operations as effortless as possible are essential. This  
207 increases the need for research to enable new interaction schemes between highly automated and complex  
208 systems, and human operators.

209 Robotic systems enabling remote operation in normal environments are the most readily available devices  
210 in the market, but their operational lifespan can be considerably reduced (Zhang et al., 2020) if used in  
211 hazardous environments without further protection. In contrast, bespoke mechanical designs that reduce  
212 the number of electronics and plastics in the joints of the robots can improve the operational lifespan of  
213 a robot, making them suitable for nuclear decommissioning environments. The use of radiation-proved  
214 robotic devices and Commercial Off the Shelf (COTS) robots together with AO and HDT technologies for  
215 training and deployment should be a focus for R&D.

216 Current regulations for most nations like the United Kingdom require that the operator performing the  
217 decommissioning task to be in complete control of any remotely operated vehicle (ROV) or tool during its  
218 deployment. By the time assisted and semi-autonomous modes of operations become readily available,  
219 such modes would need to include the capability of seamless transition from assisted to manual operation  
220 for reconfiguration and recovery in case of failure. Such capability would require an efficient Human Robot  
221 Interaction scheme to seamlessly enable transition and ensure optimal task performance.

222 The introduction of novel digital technologies should be done in such a way to ensure that new concepts  
223 or procedures are not detrimental to the success of a deployment, giving time for erroneous ideas to be  
224 identified and discarded. Such introduction should include the overall nuclear decommissioning community,  
225 including researchers, developers, deployment specialists, regulators, and site owners of different nations.  
226 This integrated and paced approach would help developers to build trust in the systems whilst creating a  
227 safety case that regulators can validate, and site owners implement.

228 Considering the state of HDTs and AO, and its rapid adoption rate outside the nuclear industry, we believe  
229 the benefits of such technologies will be realized in the coming decade. There is already evidence of nuclear  
230 site operators (Sellafield, 2023) and solution providers (UKAEA, 2023; AtkinsRealis, 2024) aligning  
231 their priorities around digital technologies and robotics. Increased funding for industry and academia to  
232 transition from Low Technology Readiness Level (TRL) to mid and high TRL is underway, with initial  
233 prototypes and use case engagement pushing for initial active demonstrators (RAICo, 2024).

## ACKNOWLEDGEMENT

234 This work was funded by the SBRI – Digital Technologies for Robotic Nuclear Decommissioning:  
235 Teleoperation with digital twins – C/2064382, and by the Robotics and AI Collaboration (RAICo).

## REFERENCES

- 236 Bogue R. Robots in the nuclear industry: A review of technologies and applications. *Industrial Robot* **38**  
237 (2011) 113–118. doi:10.1108/01439911111106327/FULL/XML.
- 238 Tsitsimpelis I, Taylor CJ, Lennox B, Joyce MJ. A review of ground-based robotic systems for the  
239 characterization of nuclear environments. *Progress in Nuclear Energy* **111** (2019) 109–124. doi:10.  
240 1016/J.PNUCENE.2018.10.023.
- 241 Sellafield. Sellafield ltd ai strategy - gov.uk. Tech. rep., Sellafield Ltd (2023).
- 242 Tugal H, Abe F, Caliskanelli I, Cryer A, Kelly R, Pacheco-Gutierrez S, et al. Haptic digital twin for clean-up  
243 process of the fukushima-daiichi nuclear power plant. *Proceedings of the International Topical Workshop*  
244 *on Fukushima Decommissioning Research* **2022** (2022) 1011. doi:10.1299/JSMEFDR.2022.0\_1011.
- 245 Zhang K, Plianos A, Raimondi L, Abe F, Sugawara Y, Caliskanelli I, et al. Towards safe, efficient long-reach  
246 manipulation in nuclear decommissioning: a case study on fuel debris retrieval at fukushima daiichi.  
247 *Journal of Nuclear Science and Technology* **62** (2025) 1–16. doi:10.1080/00223131.2024.2386478.
- 248 Tokatli O, Das P, Nath R, Pangione L, Altobelli A, Burroughes G, et al. Robot-assisted glovebox  
249 teleoperation for nuclear industry. *Robotics 2021, Vol. 10, Page 85* **10** (2021) 85. doi:10.3390/  
250 ROBOTICS10030085.
- 251 Pulgarin E JL, Tokatli O, Burroughes G, Herrmann G. Assessing tele-manipulation systems using task  
252 performance for glovebox operations. *Frontiers in Robotics and AI* **9** (2022) 323. doi:10.3389/FROBT.  
253 2022.932538.

- 254 Cryer A, Sargent A, Abe F, Baniqued PD, Caliskanelli I, Kivrak H, et al. Digital mock-ups for nuclear  
255 decommissioning: A survey on existing simulation tools for industry applications. *Robotics &*  
256 *Automation Engineering Journal* **5** (2023). doi:10.19080/raej.2023.05.555669.
- 257 Tu X, Autiosalo J, Ala-Laurinaho R, Yang C, Salminen P, Tammi K. Twinxr: Method for using digital twin  
258 descriptions in industrial extended reality applications. *Frontiers in Virtual Reality* **4** (2023) 1019080.  
259 doi:10.3389/FRVIR.2023.1019080/BIBTEX.
- 260 Elsner J, Reinerth G, Figueredo L, Naceri A, Walter U, Haddadin S. Parti-a haptic virtual reality control  
261 station for model-mediated robotic applications. *Frontiers in Virtual Reality* **3** (2022) 925794. doi:10.  
262 3389/FRVIR.2022.925794/BIBTEX.
- 263 Wang W, Na X, Cao D, Gong J, Xi J, Xing Y, et al. Decision-making in driver-automation shared  
264 control: A review and perspectives. *IEEE/CAA Journal of Automatica Sinica* **7** (2020) 1289–1307.  
265 doi:10.1109/JAS.2020.1003294.
- 266 Li G, Li Q, Yang C, Su Y, Yuan Z, Wu X. The classification and new trends of shared control strategies in  
267 telerobotic systems: A survey. *IEEE Transactions on Haptics* **16** (2023a) 118–133. doi:10.1109/TOH.  
268 2023.3253856.
- 269 Li G, Caponetto F, Wu X, Sarakoglou I, Tsagarakis NG. A haptic shared autonomy with partial orientation  
270 regulation for dof deficiency in remote side. *IEEE Transactions on Haptics* **16** (2023b) 86–95. doi:10.  
271 1109/TOH.2023.3239602.
- 272 Pruks V, Ryu JH. Method for generating real-time interactive virtual fixture for shared teleoperation in  
273 unknown environments. *International Journal of Robotics Research* **41** (2022) 925–951. doi:10.1177/  
274 02783649221102980.
- 275 Brotherhood J, Garriga MM, Hope C. Delivering a safer, faster, and better value decommissioning solution.  
276 *Proceedings of the Waste Management Symposium 2022 (ACM)* (2022), no. 22271 in Session 146 —  
277 Application of Innovative D&D Technologies Including Application of Virtual Reality (3/3)(6.7a).
- 278 Standard B, IEC61882 B. Hazard and operability studies (hazop studies)-application guide. *International*  
279 *Electrotechnical Commission* (2001).
- 280 Hurst J, McIntyre J, Tamauchi Y, Kinuhata H, Kodama T. A summary of the 'ALARP' principle and  
281 associated thinking. *Journal of Nuclear Science and Technology* **56** (2019) 241–253.
- 282 ONR. Safety assessment principles (saps). Tech. rep., Office for Nuclear Regulation (2020).
- 283 Tancik M, Weber E, Ng E, Li R, Brent JY, Kerr, et al. Nerfstudio: A modular framework for neural radiance  
284 field development. *ACM SIGGRAPH 2023 Conference Proceedings* (2023).
- 285 Kerbl B, Kopanas G, Leimkühler T, Drettakis G. 3d gaussian splatting for real-time radiance field rendering.  
286 *ACM Trans. Graph.* **42** (2023) 131–139.
- 287 Yu Z, Chen A, Antic B, Peng SP, Apratim MB, Niemeyer, et al. Sdfstudio: A unified framework for surface  
288 reconstruction. Tech. rep., SDFStudio (2022).
- 289 Millane A, Oleynikova H, Wirbel E, Steiner R, Ramasamy V, Tingdahl D, et al. nvblox: Gpu-accelerated  
290 incremental signed distance field mapping. *Proceedings - IEEE International Conference on Robotics*  
291 *and Automation* (2024) 2698–2705. doi:10.1109/ICRA57147.2024.10611532.
- 292 Tugal H, Abe F, Caliskanelli I, Cryer A, Hope C, Kelly R, et al. The impact of a haptic digital twin in the  
293 nuclear industry and potential applications. *2023 IEEE International Conference on Advanced Robotics*  
294 *and Its Social Impacts (ARSO)* (2023), 134–139. doi:10.1109/ARSO56563.2023.10187460.
- 295 Zhang K, Hutson C, Knighton J, Herrmann G, Scott T. Radiation tolerance testing methodology of  
296 robotic manipulator prior to nuclear waste handling. *Frontiers in Robotics and AI* **7** (2020) 499048.  
297 doi:10.3389/FROBT.2020.00006/BIBTEX.
- 298 UKAEA. Longops: Faster and safer decommissioning. Tech. rep., UK Atomic Energy Authority (2023).

- 299 AtkinsRealis. Digital in nuclear: our vision for 2035. Tech. rep., Atkins Realis (2024).  
300 RAICo. Raico review 2024. Tech. rep., The Robotics and Artificial Intelligence Collaboration (RAICo)  
301 (2024).