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Experimental study of effectiveness of robotic cleaning for fish-processing plants

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ABSTRACT

This paper presents the development and experimental testing of the effectiveness of a robotic cleaning system for fish processing plants. The processing of fish introduces a substantial risk of bacterial contamination, which can cause the spoilage of fish and pose a threat to consumers' health. Good operational hygiene and precautions, in addition to regular cleaning of the processing plants, are necessary for the reduction of the risk of contamination. The state-of-the art cleaning techniques currently include manual cleaning operations of fish processing plants. The experiments of robotic cleaning presented in this paper were performed in two rounds. First, a test using a conventional low-cost industrial robot mounted on a vertical linear axis was used. As the results from this test seemed promising, a second robotic system was built aiming at a more industrialized version. This system consisted of a serial manipulator, tailored for the task, mounted on a horizontal transportation system, and a comparison was conducted between the cleaning performed by human operators and that performed by the robotic system. An electrical stunner with a connected conveyor belt, which is a typical installation for salmon processing plants, was experimentally inoculated with a cocktail of fish-spoilage bacteria that were allowed to develop a biofilm. Back-to-back cleaning trials with biofilms of *Pseudomonas fluorescens*, *Pseudomonas putida*, and *Photobacterium phosphoreum* confirmed that the industrialized robotic prototype performed equally well or better than the conventional manual cleaning procedure currently used in the industry. The results demonstrate that a robotic system can deliver satisfactory results in the cleaning of fish processing plants, thereby minimizing the potential for the spread of contamination. The proposed robotic concept allows for an automated cleaning system, reduced human labor, increased profitability for the industry, and better stability of the cleaning process.

Keywords: Robotic cleaning; Fish processing; Serial manipulator; Listeria; Bacteria; Aquaculture Innovation

1 INTRODUCTION

In this paper, the results from a research project in the Norwegian aquaculture industry are presented. The aim of the project is to develop a robotic system for cleaning fish processing plants, whose performance is equal to or better than that of the manual cleaning procedure

46 that is currently followed.

47
48 The Norwegian aquaculture industry has a yearly revenue of over EUR 6 billion for salmon
49 alone (Statistics Norway, 2018). Owing to the fact that there will be an increasing need for
50 protein food sources to accommodate the anticipated growth in population toward 2030
51 (FAO Food and Agriculture Organization of the United Nations, 2016; World Bank, 2013),
52 the salmon aquaculture industry is expected to grow as it is an important protein food source.
53 However, the salmon industry faces critical challenges that may limit its further growth. One
54 of these challenges is the contamination by the human pathogenic bacterium *Listeria*
55 *monocytogenes* during production; as of yet, the pathogen has not been fully controlled in
56 food production (Buchanan, Gorris, Hayman, Jackson, & Whiting, 2017). This has led to
57 strict requirements from the Norwegian Food Safety Authority (Mattilsynet - The Norwegian
58 Food Safety Authority, 2016), which is the governing body for safe food production in
59 Norway. Additionally, there is an increasing demand for fresh, chilled fish. Microbiological
60 control of spoilage bacteria, such as *Pseudomonas* and *Shewanella*, determines the quality
61 and shelf life of fresh fish (Gram & Huss, 1996; Trond Møretrø, Moen, Heir, Hansen, &
62 Langsrud, 2016). In view of these challenges and the anticipated growth of aquaculture, there
63 is a need for the industry to find new ways to improve its procedures in all stages of the value
64 chain, from breeding to slaughtering and processing, including the improvement of the
65 cleaning procedures, to reduce the risk of bacterial contamination that may pose a potential
66 risk to human health.

67
68 Efforts have been made by researchers and companies to automate and rationalize the
69 different production processes, such as the use of robots and automated systems in different
70 gripping and handling tasks, slaughtering operations, as well as de-heading and filleting
71 (Aadland, 2018; Asche, Cojocaru, & Roth, 2018; Buljo & Gjerstad, 2013; Mikkelsen, 2017;
72 Paluchowski, Misimi, Grimsmo, & Randeberg, 2016; Sandvold & Tveterås, 2014; Sund,
73 2016). Efforts of implementing machine learning in fish processes, such as segmentation of
74 fish and species identification (Hassanien, Tolba, Elhoseny, & Mostafa, 2018) are also a part
75 of the exertions to automate more of the fish processing industry. Despite this, the total
76 amount of industrial robots in the food industry reached 9700 units in 20017, less than 3% of
77 the total supply (International Federation of Robotics, 2018), and most are used for
78 packaging/palletizing operations. The contemporary salmon industry has access to advanced
79 equipment and systems for all stages within its value chain; however, the processing speed is
80 negatively affected by several manual interventions, such as cleaning, that continue to be
81 necessary (Asche et al., 2018). Furthermore, regarding the value chain of the fish, the
82 automation of the open cleaning process has not yet been investigated; nevertheless, efforts
83 have been made in other cleaning aspects, such as the cleaning of tanks that are used in
84 aquaculture (Mcrobbie & Shinn, 2011). Systems for Cleaning-In-Place (CIP-systems) are
85 common and well developed to clean pipes and other closed systems (Cramer, 2013).
86 Cleaning is the last process step during daily fish processing.

87
88 To cope with the risk of bacterial contamination, processing plants must be thoroughly and
89 frequently cleaned (Christi, 2014; Windsor & Tattersson, 2001); more specifically, salmon
90 processing plants need to be cleaned daily. Cleaning is performed by cleaning crews after the
91 production has stopped; for several processing plants, this takes place during the night owing
92 to double processing shifts. The cleaning costs up to EUR 1 million in labor per year for a
93 processing plant owing to high wages (including bonuses related to the poor working
94 conditions and working during the night). In addition, there are high expenses related to
95 chemicals and water. Moreover, the chemicals produce a spray cloud inside the processing
96 plants during cleaning, which poses health hazards to the cleaning personnel. A typical “spray
97 mist” can be seen in

98 Table 1 Comparison of the manual and robotic cleaning time in minutes
 99 Figure 1. Furthermore, manual cleaning causes significant strain to the body from repetitive
 100 movements. The hoses that are used are heavy, and owing to high-pressure water, they are
 101 difficult to handle. Cleaners may also be required to climb on equipment to reach inaccessible
 102 areas. Overall, manual cleaning of fish processing plants requires considerable heavy lifting.
 103 A robotic cleaning system could reduce the overall cost by reducing the cost of labor and by
 104 potentially reducing the amount of chemicals and water used during cleaning. In addition, it
 105 could improve the health, safety and environment (HSE) compliance for the workers by
 106 reducing their exposure to the hazardous cleaning environment. Furthermore, a robotic
 107 solution could stabilize the cleaning process as it would perform the task in the same manner
 108 each time, thus removing the “human element,” where different cleaners may perform the
 109 tasks in a different manner. Finally, it is likely that a robotic cleaning system would perform
 110 the task faster than manual cleaners. Robot technology in general, not just for cleaning, is
 111 foreseen to play an important role in intelligent food manufacturing, replacing manual work
 112 operations in several steps along the food processing chain (Khan, Khalid, & Iqbal, 2018). As
 113 mentioned, robotic technology is implemented on some operations in the salmon industry,
 114 however, cleaning of salmon processing plants are still subject to time consuming and costly
 115 manual labor (Løvdal, Giske, Bjørlykhaug, Eri, & Mork, 2017) and problems with bacteria do
 116 occur.

117
 118 Busby and Roberts (2009) estimated that the worldwide cost of all foodborne diseases was
 119 \$1.4 trillion per year. *L. monocytogenes* is, next to *Salmonella*, by far the most frequently
 120 reported pathogenic microorganism in the Rapid Alert System for Food and Feed;
 121 notifications owing to this pathogen have increased in the EU since 2009 (European
 122 Commission, 2015). The product categories that dominated the *L. monocytogenes*
 123 notification reports were fish and fish products, often leading to trade embargoes of these
 124 products (EFSA, 2013; Nielsen et al., 2017). Recalls, consumer complaints, and bad public
 125 relations due to *L. Monocytogenes* contamination in commercial food products significantly
 126 contributed to economic losses in the food industry. An illustrating example is the 2008
 127 Canadian listeriosis outbreak linked to cold cuts from a Maple Leaf Foods (MLF) plant in
 128 Toronto, Canada. Although MLF instituted a voluntary recall before the outbreak was linked
 129 to their plant, the outbreak cost the company in excess of \$50 million including market
 130 losses, as well as lawsuits and compensations for victims and their relatives (Greenberg &
 131 Elliott, 2009). Since 1999, the EU, Norway, Switzerland, Canada, Australia, and New Zealand
 132 have all introduced a quantitative legal limit of 100 *L. monocytogenes* colony forming units
 133 (cfu) per gram, which is applied for a wide range of food products, including the most
 134 susceptible ready-to-eat (RTE) products where *L. monocytogenes* is able to proliferate, such
 135 as cold-smoked salmon products (Løvdal, 2015). USA has an even stricter legislation (i.e.
 136 zero-tolerance) for RTE products resulting in an extremely high rate of recalls from the
 137 market for potential listeriosis hazard (Goetz, 2013). Thus, measures to reduce the risk of
 138 bacterial contamination in general, particularly contamination due to *L. monocytogenes*, are
 139 imperatively necessary and sought after by the salmon industry. To safeguard food safety, it
 140 is crucial that the proposed robotic systems can perform equally well and, preferably, better
 141 than the present manual cleaning practices. The objective of the present study is to first
 142 develop and optimize, and then to evaluate the performance of a robot prototype in
 143 comparison with a contemporary manual cleaning practice in a controlled set-up using
 144 inoculation with relevant salmon spoilage bacteria that formed artificial biofilms.

145 **1.1 Future perspective**

146 Robotic cleaning systems have already been well established in the literature. However, most
 147 robotic cleaning systems were focused on the cleaning of flat surfaces, e.g., floors (Palleja,
 148 Tresanchez, Teixido, & Palacin, 2010), walls (Lee et al., 2018), windows (Houxiang Zhang,
 149 Jianwei Zhang, & Guanghua Zong, 2004), and solar panels (Jaradat et al., 2015). Cleaning
 150 systems may be able to operate in large areas; nevertheless, they are limited to moving in two
 151 dimensions, and they typically do not operate in 3D space. However, there are exceptions.

152 Cleaning systems, such as hull cleaning (Ortiz et al., 2007) and car/truck washers (Yu,
153 Kurnianggoro, & Jo, 2015), can operate in three dimensions and can clean objects of
154 arbitrary shape. However, to the knowledge of the present authors, there are no research
155 works in the literature focused on robotic cleaning for fish processing plants. Conventional
156 robotic manipulator designs do not fulfill the requirements of a robotic cleaning system for
157 fish processing plants (Bjørlykhaug, Giske, Løvdal, Mork, & Egeland, 2017). Several aspects
158 of the robotic design deserve extra attention for a robotic cleaning system focused on fish
159 processing plants. Special consideration regarding the corrosion resistance, the intrinsic
160 contamination, and the transportation system, among others, must be considered to deliver a
161 satisfying operating performance. In addition, a robotic manipulator suitable for cleaning
162 fish processing plants should have a long reach (> 2 m); however, it would have a lower
163 payload requirement than typical industrial robotic manipulators. The robotic manipulator
164 itself must have long reach, be slender, have good dexterity, and provide adequate payload;
165 meanwhile, its weight should be as low as possible. Moreover, the footprint should be kept as
166 low as possible, and the system itself should be unintrusive because modern salmon
167 processing plants often have limited space available for such installations. Moreover, a
168 robotic cleaning solution should impose minimal contamination threats, thus adhering to
169 hygienic design guidelines (EHEDG Secretariat, 2004; Giske, Mork, & Bjoerlykhaug, 2017),
170 to facilitate its efficient cleaning.

171
172 There is no existing literature that documents the effectiveness of robotic cleaning for fish
173 processing plants. Our novel contribution documents that such a system can deliver
174 satisfactory results, enabling this technology to be implemented in the industry. This work is
175 an extension of the work conducted in Bjørlykhaug et al. (2017).

176
177 The remainder of the paper is organized as follows: First, the two robot systems are
178 presented in Section 2. The setups of the experiments are presented in Section 3. In Section
179 4, the results from the two different experiments are presented. Finally, in Section 5, the
180 present work is concluded, and further work is discussed.

181 **2 ROBOTIC SYSTEM**

182 Here, we will present the robotic system used in the experiments. The experiments were
183 conducted in two separate occasions, with two different robotic systems. The systems were
184 designed according to the challenges related to installing such a system in a real-world
185 processing plant. Examples of the equipment layout inside a plant are shown in Figure 2;
186 Figure 3 depicts typical installation locations for a future robotic cleaning system.

187 **2.1 System 1**

188 The first robotic system consisted of a conventional serial robot, namely the UR10, mounted
189 onto a vertical linear axis with a slewing ring, as shown in Figure 4. In addition, a 1 m long
190 lance holding the nozzle was mounted on the end effector. The combination of these factors
191 enabled the robot, which originally had a reach of 1300 mm, to cover a complete electric
192 stunner. This system was manually programmed “online”, jogging the robot from point to
193 point and creating a cleaning path. An overview of the architecture can be seen in
194 Bjørlykhaug et al. (2017).

195 **2.1.1 Cleaning system**

196 The cleaning system was composed of the industrial cleaning station 4K on wheels by System
197 Cleaners. This equipment required 400 V AC and its input was regular pressure water. The
198 output was high-pressure water or a mixture of air, chemicals, and water, used for spraying
199 foam. The end could hold different types of nozzles based on which mixture is sprayed. The
200 end of the accompanying hose was attached to the UR10 robot in Experiment 1 and it was
201 used directly.

202 **2.2 System 2**

203 This system was an upgrade of System 1 and it was an effort to eliminate the drawbacks of
204 System 1. Instead of a conventional robotic manipulator from a commercial supplier, a
205 custom robotic manipulator tailored to the task was constructed, as can be seen in Figure 5.
206 In addition, the robot was mounted on a custom horizontal linear axis suitable for
207 installation in the harsh environment of fish processing plants,
208 with the necessary hygienic considerations. The robot itself is a long-reach, slender robot,
209 with a low payload capability (compared with typical industrial robots of the same reach),
210 thereby maintaining the manipulator weight as low as possible. The robot and its kinematic
211 chain are shown in Figure 6 and Figure 7, respectively. At full horizontal extension, the robot
212 has a reach of 4 m; furthermore, its weight was approximately 220 kg.
213

214 **2.2.1 Cleaning system**

215 The cleaning system of Experiment 2 was identical to the one in Experiment 1, except for the
216 fact that the hose was separated into six different hoses, each of which had their own solenoid
217 valve that could be controlled to be switched on or off. Each of the six hoses continued up to
218 the end-effector of the robot, where six identical nozzles were mounted at an angle with
219 respect to each other, as seen in Figure 8. For different parts of the cleaning procedure, a
220 different number of nozzles was used. This eliminated the need for extra degrees of freedom
221 (DOFs) close to the end-effector of the robot, thus removing the need for a servomotor near
222 the end effector; this minimized the weight and made the robot slenderer.

223 **2.2.2 Control system**

224 One of the main limitations of System 1 was the manual programming of the robotic system.
225 This proved to be excessively time consuming and tedious; therefore, a better approach was
226 required, particularly considering that a potential industrialized version would be installed in
227 different plants, thus requiring different paths to be programmed. Offline programming of
228 the robot movements was decided to be the preferred approach. Because the robot was built
229 anew, a control system had to be developed. For the control system, a distributed approach
230 was used. Instead of implementing the kinematics of the robot in the programmable logic
231 controllers (PLC), which controls the servos, a computer program calculated the actuator
232 positions for the desired robot pose. A schematic detailing the control system approach is
233 shown in Figure 9. For a more thorough explanation of the trajectory generator, we refer to
234 Bjørlykhaug (2018).
235

236 **2.2.3 Horizontal transportation system**

237 The manipulator was mounted onto a horizontal linear axis to expand the work envelope of
238 the robotic system. Similar to the manipulator, the linear axis was also built specifically for
239 the task. Fish processing plants often have a processing layout that is not in a straight line.
240 Unlike conventional linear axes, which are limited to a straight line of motion, this one has a
241 modular design that was built from sheet metal building blocks, thus enabling curvature.
242 With curvature, the axis was able to navigate the robot base in 2D, which potentially covered
243 the complete processing plant depending on the particular plant layout.

244 **3 METHOD AND TOOLS**

245 Here, we will present how the experiments were set up for both cases. A physical experiment
246 to measure the cleaning effect of robotic cleaning was the chosen method for both systems.

247 **3.1 Experimental setups**

248 **3.1.1 Experiment 1**

249 Regarding the methodology for Experiment 1, we refer to Bjørlykhaug et al. (2017). The robot
250 in action during Experiment 1 is shown in Figure 10.

251 **3.1.2 Experiment 2**

252 Experiment 2 was set up as back-to-back experiments between cleaning by human operators
253 and cleaning by the robotic system. The equipment used to perform the cleaning test
254 consisted of an electric stunner and a conveyor used for gill cutting, both of which can be
255 typically found in fish processing plants. These machines are often considered among the
256 most important machines for cleaning, as they are situated immediately after the fish has
257 been pumped into the processing facility. Typically, the machines are filled with fish residue,
258 fish protein, fish slime, and fish parts (of fins). Because this is where the gills are cut, blood is
259 usually spilled on a large part of these machines. The dimensions of the mini fish
260 slaughtering line cleaned in Experiment 2, as well as the sampling points for microbiological
261 analysis are shown in Figure 12. A length of almost 6 m and a width of a little over 1 m is close
262 to a typical installation at a fish processing facility.

263 **3.1.3 Microbiology analysis**

264 The electric stunner and conveyor were inoculated, as seen in Figure 11, with a bacterial
265 suspension cocktail of *Pseudomonas fluorescens* MF05002 (Trond Møretrø et al., 2016),
266 *Pseudomonas putida* ATCC 49128 from the American Type Culture Collection, and
267 *Photobacterium phosphoreum* CCUG 16288 from the Culture Collection University of
268 Gothenburg. Bacteria cultivation, inoculation, and sampling were performed as previously
269 described (Bjørlykhaug et al., 2017), with only minor modifications. All bacteria were initially
270 grown separately to a stationary phase at 30 °C and 150 rpm in a shaking incubator in a
271 tryptic soy broth with 0.6% of yeast extract (TSBYE; Oxoid). Bacteria were pooled together at
272 250 mL of each strain in 1 L of sterile polyethylene bottles, which were then topped up with
273 250 mL of fresh TSBYE. The bacteria were maintained at ambient temperature (10–20 °C)
274 and were used within 72 h. The inoculation was performed by spraying the bacteria using a
275 household spray flask on all open surfaces. Spraying was repeated once each hour four times.
276 Twenty-four hours after the first spraying, an incomplete biofilm had developed on the
277 surfaces (approximately 10^6 cells·cm⁻²). Prior to washing, eight predefined control points
278 were sampled using Sodibox cloths (Sodibox, La Fort-Fouesnant, France). After the manual
279 washing procedure had been completed and the stunning machine had been air dried, an
280 additional eight predefined control points were sampled using Sodibox cloths (Figure 12).

281
282 The following day, the same routine was repeated for the robotic cleaning using bacterial
283 suspensions of the same age as for the manual cleaning. The samples were maintained at 4°C
284 and were plated 48 h after sampling. Sodibox cloths were suspended in 100 mL buffered
285 peptone water (Oxoid) and were subject to homogenization using a stomacher machine
286 (Seward) for 2 min. Serial dilutions of the samples were spread-plated in triplicate on tryptic
287 soy agar with 0.6% of yeast extract (TSAYE; Oxoid); then, they were incubated at 30°C for 48
288 h before the bacterial concentrations were calculated as cfu per cm². The data are presented
289 as logarithmic reductions in the plate counts (ΔN) between the counts before (N_o) and after
290 (N_x) cleaning, namely $\Delta N = \log N_x - \log N_o$.

291 **3.2 Cleaning procedures**

292 A general cleaning process for fish processing plants is formulated in (Mariott & Gravani,
293 2006):

- 294 1. Cover electrical equipment.
- 295 2. Remove large debris.

- 296 3. Remove soil deposits from the equipment, walls, and floors, proceeding from top to
297 bottom towards the drains.
- 298 4. Disassemble equipment as required.
- 299 5. Pre-rinse the equipment with water at 40 °C or less.
- 300 6. Apply a cleaning compound effective against organic soil (typically an alkaline cleaner),
301 with a temperature lower than 55 °C.
- 302 7. Wait for approximately 15 min to allow the cleaning compound to work.
- 303 8. Rinse the equipment with water at 55-60 °C.
- 304 9. Inspect equipment and the facility for effective cleaning.
- 305 10. Apply a sanitizer, typically a chlorine compound

306 This procedure coincides with the cleaning procedures typical in Norwegian fish processing
307 plants (Løvdal et al., 2017).

308 For our case, Steps 1, 2, 3, and 4 would not be necessary to be followed in any of the
309 experiments owing to the non-existence of fish debris and the absence of electrical
310 equipment. The test equipment was designed in such a way that the disassembly would not
311 be required prior to cleaning as part of the daily cleaning routine.

312 For the cleaning process in Experiment 1, please refer to Bjørlykhaug et al. (2017). The
313 cleaning process in Experiment 2 was performed as follows: the equipment was hosed down
314 using cold high-pressure water at first. Immediately after hosing down all machines, a thick
315 foam of *Lilleborg Enduro Super* with a diluted pH of 12,5 was sprayed on. This was allowed
316 to stay on for 10 min before being washed off by using cold high-pressure water. Then, a foam
317 layer of *Lilleborg Titan 951*, a disinfectant of pH 7 (diluted), was sprayed on and was allowed
318 to stay on for 10 min. Again, the equipment was washed down with cold high-pressure water.
319 This procedure was performed first manually by one cleaner, as seen in Figure 13; after the
320 equipment was inoculated with bacteria again, the robot system repeated the same cleaning
321 procedure. The cleaner performing the manual cleaning was employed at an undisclosed fish
322 processing plant as the team leader for all cleaners with 15 years of experience in cleaning
323 fish processing plants. He performed the cleaning as he would have normally done.
324 The robot was programmed offline in the simulation software, as shown in Figure 14. The
325 prototype in the test facility is shown in Figure 6.

326 Our cleaning procedure differs slightly from the general procedure detailed above. We
327 followed the general recommendations of the professional cleaner who performed the
328 manual cleaning, which is the industry standard. Moreover, we did not have access to hot
329 water at our test facility. However, it is common practice in the industry to use cold water.

330 4 RESULTS

331 We present the results from both experiments. The robot in Experiment 2 can be seen
332 spraying water and soap in Figure 15 and Figure 17, respectively. In this study, the cleaning
333 times for manual and robotic cleaning were not of primary importance; however, they are
334 listed in Table 1 for comparison.

335 4.1 Microbiology

336 The decrease in the bacteria count in Experiments 1 and 2 can be seen in Figure 16. The
337 decrease in bacteria after robotic cleaning in Experiment 1 (Figure 16A) was promising, and
338 the bacteria count was between 10 and 100 cfu-cm⁻² for all control points, compared with 10³
339 to 10⁵ cfu-cm⁻² prior to cleaning.

340 In Experiment 2, the decrease in microbial count for both manual cleaning and robotic
341 cleaning was substantial, as seen in Figure 16B. In this trial, the inoculated bacterial load was
342 higher than that in Experiment 1, *i.e.*, between 10⁵ and 10⁷ cfu-cm⁻². This difference is

343 presumably attributed to the significantly higher temperatures (~ 20 °C) of growth compared
344 with the ones in Experiment 1 (~ 5 °C), combined with a slightly extended time between the
345 inoculation and the cleaning. In Experiment 2, the bacterial load after both manual and
346 robotic cleaning was between 1 and 200 cfu·cm⁻². Although the data were too sparse for us to
347 perform a statistical comparative analysis, it is clear that the robot performed the cleaning at
348 least as well as or better than the operator in this particular case (Figure 16B).

349 5 DISCUSSION

350 A main limitation of this work is that only one instance of both experiments was performed.
351 Performing additional repetitions of the same experiments would result in a more reliable
352 measurement of the effectiveness. Additionally, these experiments were performed in a
353 closed scenario. Testing in a real-world processing plant might have affected the results in a
354 certain manner. However, Experiment 2 was performed at a technology readiness level (TRL)
355 of 5–6 (Horizon 2020 Work Programme Commission, 2014), and clearly illustrated the
356 validity of robotic cleaning. The sprayed bacteria produced a biofilm that is close to the real
357 biofilm often found in fish processing facilities; the experiments showed that the robot was
358 fully capable of washing the biofilm away, thus inhibiting the establishment of niches known
359 to facilitate growth of spoilage bacteria and human pathogenic bacteria, such as the *L.*
360 *monocytogenes* (T. Møretrø & Langsrud, 2004).

361 An even more industrialized version of the robot is required in the future and tests in a real-
362 world fish processing plant is the only manner to validate if the robot can perform as well or
363 better than a manual cleaner. Furthermore, it would be beneficial to combine the robot
364 solution with a vision system to detect blood or microorganisms; these would eliminate the
365 need for manual control after cleaning.
366

367 It is probable that the optimization of the robot program related to information collected
368 from several trials, both in the laboratory and in real life, in a fish-processing factory will
369 further improve the cleaning results, particularly in corners and places of limited
370 accessibility.
371

372 Corners and spaces of limited robot accessibility can be evaluated for a redesign, layout
373 altering, or a more hygienic design as well (Giske et al., 2017). Robot cleaning can be
374 considered as a method to standardize the cleaning process compared with human operators,
375 as the robot will perform the task equally well on each instance. The standardization of the
376 cleaning process will also stabilize the method, which is one of the core principles in lean
377 manufacturing (Liker & Meier, 2005), thus enabling a more predictable performance of the
378 cleaning process, which can be used in further efforts to further stabilize the fish processing
379 in general. This will allow for incremental improvements from several aspects; the robot
380 operation itself, the design of the equipment, and even the type of materials (stainless steel,
381 plastic, etc.) can be adapted for optimal cleaning processes. In addition, high-temperature
382 steam or new chemicals that may not be allowed or are suitable to be currently used owing to
383 the HSE requirements or considerations, may now be possible to be used.
384

385 Furthermore, the robotic system can store the cleaning operation in an electronic format and
386 offer it as proof of the cleaning quality to the customers, thus substantiating that their fish
387 products are produced in an environment that does not compromise quality. Fish processing
388 companies claim that this is key in competing in the global seafood market.
389 The robot performed the cleaning at a relatively slow, steady pace during Experiment 2. The
390 speed was limited owing to vibrations in the system, which became great if the speed was
391 increased. In this research, only the cleaning results were the focus, and not the cleaning
392 speed. In the next version, the robot design will be upgraded so that the robot may achieve
393 higher speeds and accelerations; and it is still believed that a robotic solution will perform the
394 cleaning tasks faster than manual cleaning operations. Furthermore, it is also possible to use
395 several robots to clean different areas. When cleaning a large area, the waiting time may be

396 neglected because the robot can rinse the first part of the area immediately after the
397 application of the chemicals on the last part of the area.

398
399 In addition, robotic cleaning of fish processing plants may reduce cleaning costs. A manual
400 cleaner in Norway earns EUR 60-70k per year; the suggested solution could easily replace
401 several cleaners. In a robotic cleaning concept, one could possibly even allow increased
402 cleaning time as the main cost, whereas manual work hours would be reduced. Even further
403 savings are predicted when tuning the robot to use exactly the amount of water and chemicals
404 that are needed for the task, instead of consuming random amounts used in modern manual
405 cleaning processes.

406 **6 CONCLUSION AND FUTURE WORK**

407 It can be concluded that although the robot system has its limitations in its current form, a
408 robotic cleaning system can perform the cleaning as well as or better than manual cleaning.
409 Additionally, the repeatability of a robotic system compared with human operators will
410 potentially ensure better hygiene and control of bacteria that develop in fish processing
411 plants over time.

412 **6.1 Future work**

413 Future work related to robotic cleaning will need to focus on installing a robotic cleaning
414 system in a producing fish processing plant. This will enable long-term measurements of the
415 effectiveness of robotic cleaning at a higher TRL. There are still certain challenges that have
416 not been addressed by the experiments conducted in this study, including the fact that in
417 real-life processing plants, certain fish remains are likely to be present on the equipment.
418 This is because the robot only performs a predefined cleaning path and does not consider
419 how to identify and hose away the remains as a manual cleaner would do. The robotic
420 cleaning system must be expanded into including vision and algorithms for the detection of
421 fish remains, as well as be taught how to hose away such residue for it to fully replace manual
422 cleaners. The possibility to develop a vision system to detect blood and biofilm should be
423 investigated, as well as its use for the programming of the robot, and possibly for measuring
424 cleaning results. Machine learning and simulation could also be used to enable the robot to
425 program itself, and to use cleaning results to optimize the cleaning path.

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 485 [2020-wp1415-annex-g-trl_en.pdf](https://ec.europa.eu/research/participants/data/ref/h2020/wp/2014_2015/annexes/h2020-wp1415-annex-g-trl_en.pdf)
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555

- 556 *Table 1 Comparison of the manual and robotic cleaning time in minutes*
- 557 *Figure 1 The environment in which a robotic cleaning system will operate*
- 558 *Figure 2 Slaughtering line at facility 1*
- 559 *Figure 3 Slaughtering line at facility 2*
- 560 *Figure 4 A CAD model of the linear axis and the rotational axis assembly. A support frame was manufactured to*
561 *suspend the assembly from the ceiling*
- 562 *Figure 5 A CAD model of robot 2, mounted on the horizontal linear axis*
- 563 *Figure 6 Robotic system 2 in the test facility*
- 564 *Figure 7 Kinematic chain of the custom robot in prototype 2*
- 565 *Figure 8 Nozzle arrangement*
- 566 *Figure 9 Overview of the control system*
- 567 *Figure 10 Prototype 1 during Experiment 1*
- 568 *Figure 11 Spraying of the inoculation mixture in Experiment 2*
- 569 *Figure 12 Control points and dimensions of equipment in Experiment 2. The electric stunner in the cleaning*
570 *position with the conveyor belt connected at the front. Sampling points for microbiological analysis in*
571 *Experiment 2 are indicated as follows: S1 and S2 are steel lamellae on the electrical stunner; P1 and P2 are*
572 *plastic walls inside the electrical stunner; S3 and S4 are steel cross-beams under the belt of the conveyor and*
573 *the stunner, respectively. P3 and P4 are the inside plastic walls on the conveyor and the stunner, respectively.*
574 *(Care was taken not to sample the same areas before and after cleaning)*
- 575 *Figure 13 Manual cleaning in Experiment 2*
- 576 *Figure 14 Simulation of the cleaning process*
- 577 *Figure 15 Prototype 2 spraying water*
- 578 *Figure 16 Bacterial log reduction as an effect of robotic cleaning in Experiment 1 (A) and as an effect of manual*
579 *and robotic cleaning in Experiment 2 (B). S=Steel and P=Plastic. For sampling points in (B), see Figure 12. Data*
580 *in (A) are redrawn from Bjørlykhaug et al. (2017). Error bars represents standard deviation (SD) of n=3*
581 *triplicate plates per sample with error in log Nx and log No propagated ((SD log Nx+SD log No2)^{1/2})*
- 582 *Figure 17 Prototype 2 spraying soap*

Highlights: Experimental study of effectiveness of robotic cleaning for fish processing plants

The cleaning of fish processing plants is in large part a costly operation which is unstable in terms of outcome.

Two cleaning experiments in which a robot is used to do cleaning of bacteria inoculated fish processing equipment is done.

Both experiments show very promising results after testing, indicating that robots could clean fish processing plants.

Robotic cleaning of fish processing plants could enable more stable cleaning results and mitigate bacterial contamination.



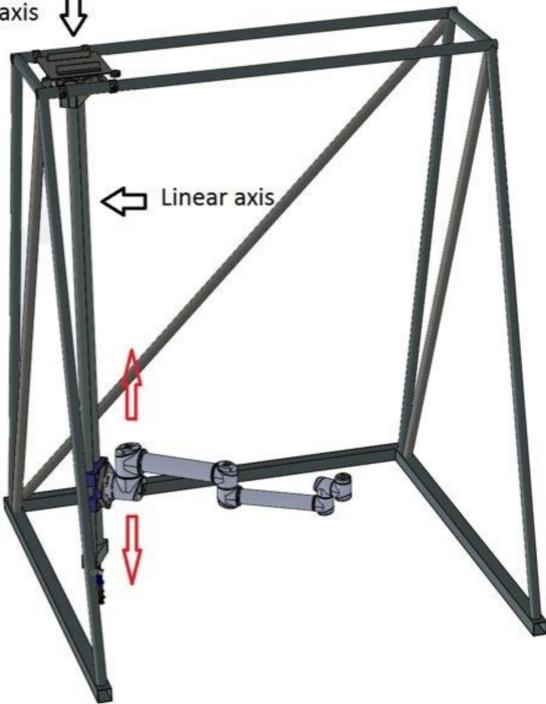




Rotation axis

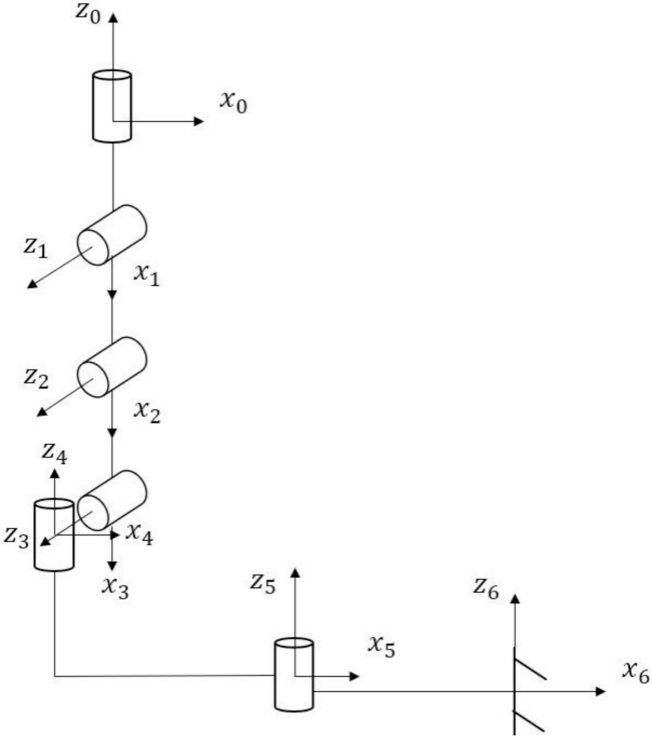


Linear axis











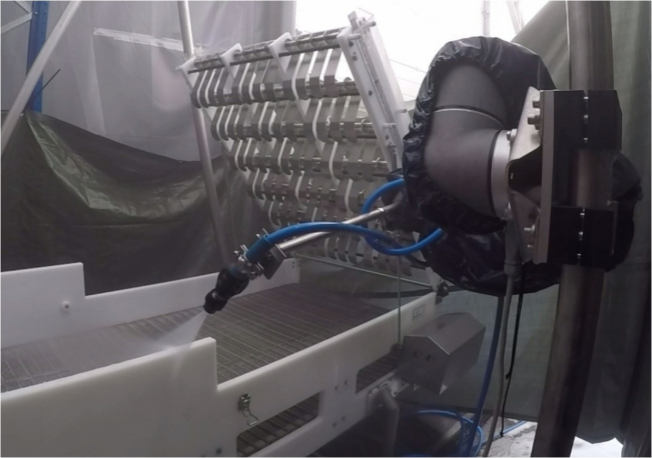
Simulation
software

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graph LR; A[Simulation software] --> B[Trajectory generator]; B --> C[PLC]; C --> D[Servos & I/O]
```

Trajectory
generator

PLC

Servos & I/O







1030mm

S3

P3

5750mm

P1

S1

S2

P2

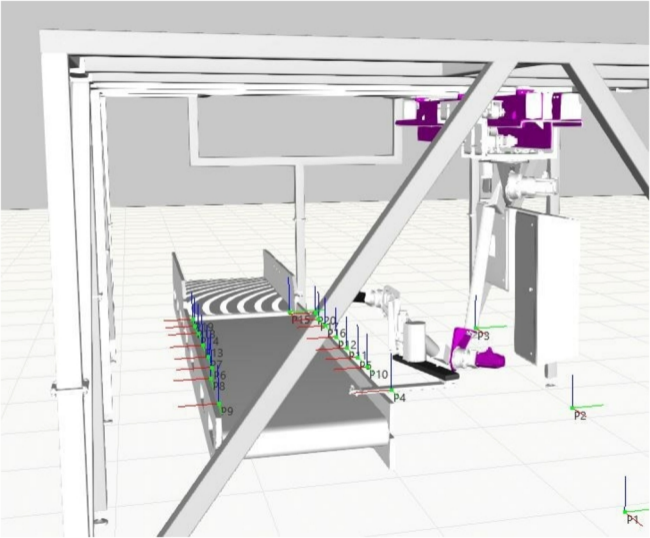
S4

P4

1000mm

30mm







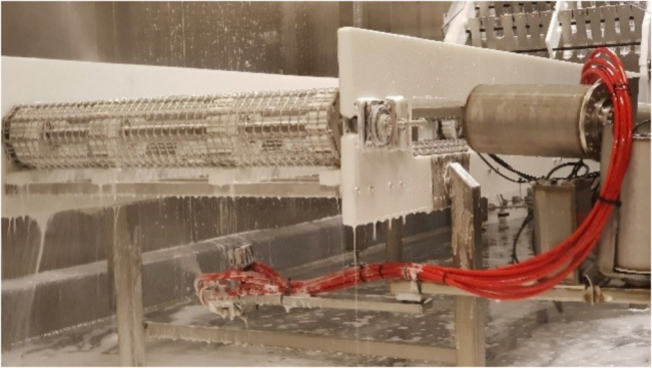


Table 1 Comparison of manual and robotic cleaning time in minutes

Cleaning procedure	Rinsing	Foam soap	Rinsing	Foam disinfectant	Rinsing	Total time
Manual time	6	2	10	1,5	5,5	45*
Robotic time	33	33	33	33	33	185*

**Included in the total time is 2 x 10 minutes of waiting time between applying foam and rinsing to allow the chemicals to work*