

Clean-in-Place Monitoring of Different Food Fouling Materials using Ultrasonic Measurements

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Abstract:

Clean-in-Place is an autonomous technique used to clean the internal surfaces of processing equipment in the food and drink sector. However, these systems clean for a longer time than required with negative economic and environmental impacts. In this work, an ultrasonic sensor system was developed to monitor the cleaning of different food fouling materials at laboratory scale. The fouling removal of three different food materials was also studied at different cleaning fluid temperatures. The three food materials had different cleaning mechanisms, which could be monitored successfully with the ultrasonic system. Tomato paste and gravy appeared to be cleaned by mechanical forces whereas malt extract dissolved into the cleaning water. The results yielded from the cleaning of the malt was found to be repeatable whereas the tomato and gravy were more variable between repeat experiments. It was found that changes in recorded ultrasonic signals were mainly affected by the area of fouling that covered the transducer's active element.

Keywords: Ultrasonic Measurements, Clean-in-Place, Process Analytical Technologies, Food Materials

1. Introduction:

Cleaning of processing equipment is an essential operation in sectors such as food and drink, pharmaceutical and Fast Moving Consumer Goods (FMCG). Cleaning is performed to ensure processing equipment remains hygienic and to remove any internal surface fouling which may have accumulated during processing. Any remaining fouling can have negative effects on the performance of processing equipment (e.g. heat transfer rates in heat exchangers) and can cause cross

35 contamination issues if it becomes dislodged during subsequent processing. This is of critical
36 importance when processing equipment is used to produce numerous different products, some
37 containing allergens, as is often the case in the food and drink sector. Most processing facilities clean
38 their equipment using a method called Clean-in-Place (CIP). This is an autonomous process which
39 utilises water and chemicals to perform the cleaning and does not require equipment disassembly or
40 manual intervention by factory workers. CIP processes clean the equipment by a combination of: time,
41 chemicals, temperature and mechanical force. Different cleaning chemicals and CIP processes are
42 utilised depending on the physical chemical properties and volume of the fouling material. Most CIP
43 processes are designed specifically for the equipment they are cleaning but generally feature five
44 stages: 1) pre rinse 2) chemical cleaning 3) post rinse 4) sanitisation 5) final rinse (Fryer, Christian, &
45 Liu, 2006). CIP processes are routinely validated to ensure the equipment is clean and free of
46 microorganisms after a cleaning cycle. This validation is performed using Adenosine Triphosphate
47 (ATP) kits or growth and microbial enumeration tests (Fratamico, Annous, & Guenther, 2009).

48 Within the food and drink manufacturing sector most processing equipment is used to manufacture
49 numerous different products each with a unique formulation and fouling behaviour. In general, CIP
50 systems are designed for the worst case fouling product and over-cleaning is a common occurrence.
51 This over-cleaning has significant negative economic and environmental impacts resulting from the
52 unnecessary overuse of resources such as water and energy and lost production time. It has been
53 reported that cleaning is responsible for 30% of the energy used in dairy processing (Eide, Homleid, &
54 Mattsson, 2003), and approximately 35% of the water use in beer production (Pettigrew,
55 Blomenhofer, Hubert, Groß, & Delgado, 2015). This problem is increased in the current climate where
56 consumer preferences are driving mass customisation of products (Vries et al., 2018). This mass
57 customisation leads manufacturers to produce more batches (each of lower volume) requiring a
58 cleaning operation between every batch. Therefore, the total time spent cleaning equipment rather
59 than in production, is increasing.

60 Research performed to optimize CIP processes has generally focused on understanding the effects of
61 different cleaning parameters (e.g. water flow rate) (Fan, Phinney, & Heldman, 2018), or studying
62 fouling and cleaning using a range of different measurements (Fratamico et al., 2009). The most
63 common measurement techniques include simple sensors monitoring either temperature (Vieira,
64 Melo, & Pinheiro, 1993) or pressure (Riverol & Napolitano, 2005) at different locations in the
65 equipment or methods that detect the presence of fouling or chemicals in the cleaning water
66 (Lyndgaard, Rasmussen, Engelsen, Thaysen, & Van Den Berg, 2014) (Berg, Ottosen, Berg, & Ipsen,
67 2017) (Van Asselt, Van Houwelingen, & Te Giffel, 2002). However, in many cases the fouling material
68 will remain adhered to the internal walls of the processing equipment and measurements of the

69 cleaning water would be unsuitable to determine the actual cleanliness of the equipment. To monitor
70 the degree of surface fouling numerous different techniques have been investigated. In a 2012 review
71 of fouling detection methods (E. Wallhäußer, Hussein, & Becker, 2012a), in heat exchangers, it was
72 stated that the main sensor techniques were based on either electrical (e.g. (X. D. Chen, Li, Lin, &
73 Necati, 2003) (Guérin, Ronse, Bouvier, Debreyne, & Delaplace, 2007) (Tlili, Rousseau, Ben Amor, &
74 Gabrielli, 2008)) or acoustic methods (e.g. (Pereira, Mendes, & Melo, 2009) (P. M. Withers, 1996)
75 (Úbeda, Hussein, Hussein, Hinrichs, & Becker, 2016)). Optical sensors utilising ultraviolet fluorescent
76 techniques have also been used to monitor cleaning processes (P. M. Withers, 1996) (Simeone, Deng,
77 Watson, & Woolley, 2018) (Simeone et al, 2016) and a fibre optical device has been used to monitor
78 biofilm fouling in a brewery water pipe (Tamachkiarow & Flemming, 2003). However, optical imaging
79 technologies require lighting to enable imaging of the surface under investigation so they would not
80 be suitable in pipes or processing equipment where suitable illumination would be extremely
81 challenging (e.g. heat exchangers).

82 Various researchers have investigated acoustic techniques to measure fouling and cleaning in systems
83 representative of processing equipment. These techniques can generally be split into three methods:
84 1) guided waves, 2) low frequency vibrations (<20 KHz), 3) Ultrasonic (US) techniques. Guided wave
85 techniques utilise shear (transverse) waves, which propagate along the surface between a solid and
86 fluid, or through a solid material located between two fluids. For inspection and measurements in
87 pipes, guided wave techniques generally use separate transducers for wave generation and detection
88 and can propagate waves either along a length section of a pipe or around its circumference. Guided
89 wave techniques have been used to measure a range of different fouling materials in pipes (Hay &
90 Rose, 2003), (Lohr & Rose, 2003), (Jaidilson Jó da Silva, Lima, & da Rocha Neto, 2007), (J J Silva, Silva,
91 Lima, & Neto, 2008). The group of Rose demonstrated that guided wave techniques were capable of
92 detecting the presence of internal surface fouling of materials such as grease, oils and fats in pipes
93 filled with water (Hay & Rose, 2003) (Lohr & Rose, 2003). The group of Silva performed experiments
94 in flow conditions and demonstrated the capabilities of guided wave techniques for monitoring fouling
95 growth in pipes using a variety of different signal processing methods (Jaidilson Jó da Silva et al., 2007)
96 (J J Silva et al., 2008). Vibration techniques feature two different components. The first is some type
97 of mechanical or electro-mechanical instrument, which causes a vibration in the system under
98 inspection. The second component is a sensor to detect these vibrations (e.g. piezoelectric
99 transducer). The principle behind fouling detection using vibration methods is that any fouling on the
100 internal surface of the pipe or test section will have an effect on vibrations within the material, which
101 can be detected by the sensor. Pereira et al. used a mechatronic sensor to monitor dairy fouling, and
102 found that the amplitude of the detected vibrations reduced as the thickness of the fouling layer

103 increased (Pereira, Rosmaninho, Mendes, & Melo, 2006). The same group used a similar technique to
104 study the cleaning of shampoo residue and also studied the effects that temperature and flow rate
105 had on the fouling layer removal (Pereira et al., 2009). Merheb et al., 2007 studied dairy fouling in a
106 heat exchanger and found that the recorded acoustic power reduced as the amount of surface fouling
107 increased (Merheb, Nassar, Nongaillard, Delaplace, & Leuliet, 2007). Acoustic hammer and
108 microphone techniques have shown that the frequency content and decay in received acoustic signals
109 both reduced when the pipe test sections were fouled with a paraffin resin (Jaidilson Jó da Silva, Lima,
110 Neff, & da Rocha Neto, 2009) (Jó, Marcus, Lima, Neff, & Sérgio, 2009) .

111 Ultrasonic techniques feature one or more transducers operating in the US frequency range (> 20
112 KHz), and were one of the first techniques proposed to monitor fouling in processing equipment.
113 Withers used two US transducers operating in a transmission mode system to measure the thickness
114 of a range of food and non-food fouling materials (P. Withers, 1994) (P. M. Withers, 1996). This work
115 was performed with the fouling material applied to the internal surface of a pipe section, which was
116 then flooded with water and showed that US velocity can be used to measure fouling layer thicknesses
117 between 0.5 – 6 mm. Other researchers used a single US transducer operating in reflection mode and
118 an Artificial Neural Network (ANN) to monitor the thickness of dairy fouling (Wallhäußer, Hussein,
119 Hussein, Hinrichs, & Becker, 2011). Their measurements were performed in a bespoke experimental
120 container with the fouling material located on one internal surface before being filled with water. They
121 recorded US waves reflected initially from the wall/fouling layer interface and those which travelled
122 through the water before been reflected back. These reflected signals were analysed in terms of their
123 acoustic impedance, acoustic energy, and logarithmic decay. These features were then used as inputs
124 for an ANN, which was capable of detecting the presence of fouling with an accuracy around 99%. This
125 group continued this work on dairy fouling in a static system and performed a sensitivity analysis that
126 showed that the recorded US results were more dependent on the acoustic impedance of the fouling
127 material than the fouling layer thickness (E. Wallhäußer, Hussein, & Becker, 2012b). They also
128 performed work where they created fouling layers of either protein or mineral materials (E.
129 Wallhäußer, Hussein, Hussein, Hinrichs, & Becker, 2013) and investigated different classification
130 methods for determining the presence of fouling. They showed that Support Vector Machines (SVM)
131 performed better in classification than ANN, with higher accuracy when predicting the presence of
132 protein than mineral fouling. This group also recorded US measurements during the dynamic cleaning
133 of dairy material. They showed that with different features extracted from the recorded US signals,
134 classification methods could be developed with accuracies of 98% in predicting the presence of dairy
135 fouling (Eva Wallhäußer et al., 2014) (Úbeda et al., 2016). Chen et al., developed a method which
136 utilised an US transducer in contact with the external wall of a rectangular duct section (B. Chen et al.,

137 2019). On the internal surface of this duct, adjacent to the transducer, a 6mm thick layer of wax was
138 placed. This wax was representative of a layer of fouling material. They flowed water through the duct
139 for a period of approximately three hours and showed that coda wave interferometry techniques,
140 applied to the recorded US signals, could successfully be used to monitor the removal of this wax layer.
141 Although there has been progress in the development of US techniques to detect the presence of
142 fouling the majority of this work has focussed on either dairy fouling or model materials.
143 In this work, an US system, that is capable of monitoring the cleaning of different fouling materials
144 relevant to the food and drink manufacturing sector is described. A range of cleaning fluid
145 temperatures and different signal and data processing methods were investigated. Images were
146 recorded during cleaning and used to determine which aspects of the fouling layers had the largest
147 effect on the recorded US signals.

148 **2. Material and Methods:**

149 **2.1 Ultrasonic methods**

150 Ultrasonic techniques operate by transmitting low amplitude, high frequency acoustic waves through
151 the system under investigation. They are an attractive sensing technology due to their small size, low
152 cost and ability to perform measurements non-invasively on opaque systems (Watson, 2015). This
153 work utilised a single US transducer used for both transmitting and receiving the US signal. This type
154 of system is called a reflection mode system and have been used for a variety of industrial applications
155 including corrosion monitoring (Cheong, Kim, & Kim, 2017) and flow monitoring (Al-Aufi et al., 2019).
156 The US system developed during this work features an US transducer attached to the bottom of a test
157 section (Figure 1). The US wave was transmitted from the transducer through the solid pipe wall. At
158 the wall/fouling layer interface a proportion of the wave was reflected. The wave then propagated
159 back through the wall before being detected by the same transducer. The proportion of the wave that
160 was not reflected from the wall/fouling layer interface propagated through the fouling layer until it
161 reached the fouling layer/water interface where another proportion was reflected, and the remainder
162 transmitted into the water. Figure 1 depicts the fouling layer/water interface as a solid line but in
163 reality this would be much more complex due to constantly changing geometry and localised material
164 properties, resulting from the flowing water.

165 When an US wave becomes incident at an interface, the ratio of the reflected pressure amplitude to
166 the incident pressure amplitude can be determine by the reflection coefficient (R):

$$167 \quad R = (Z_2 - Z_1)/(Z_2 + Z_1) \quad (1)$$

168 Where Z is the acoustic impedance define by:

$$169 \quad Z = \rho v \quad (2)$$

170 Where ρ is the density of the material and v is the velocity of a longitudinal US wave in the material.
171 The above equation indicates that reflection increases with the size of the impedance mismatch.
172 However, strictly, R is a function of incident angle θ and can change markedly with θ . The reflection
173 coefficient and impedance of the materials (e.g. fouling layer) can be calculated by measuring the
174 magnitude of the reflected US waves. Previous research has shown that acoustic impedance
175 calculations of material adjacent to a test section wall can be used to determine the presence of
176 fouling ((Eva Wallhäußer, Hussein, Hussein, Hinrichs, & Becker, 2011) (E. Wallhäußer et al., 2013)).
177 However, these methods generally require the reflected wave to be separated in time from the
178 transmitted wave and this is only true when the wall thickness ($\delta_{w,min}$) is greater than the value
179 specified by:

$$180 \quad \delta_{w,min} \geq Nv/2f \quad (3)$$

181 Where N is the number of wave cycles in the ultrasonic pulse, v is the speed of US wave in the wall
182 material and f is the frequency of US wave. If the wall thickness is greater than this value, the reflected
183 waves are superimposed upon one another and signal and data analysis becomes challenging. Some
184 proportion of the US waves may be reflected from the fouling layer/water interface. However, this
185 proportion is likely to be extremely small due to the changing geometry and properties described
186 above, and the similar values of acoustic impedance for the water and fouling material. Chen et al.,
187 2018 were able to monitor the removal of a fouling layer when reflected US signals were overlapped
188 using a coda wave interferometry technique (B. Chen et al., 2019). During the formation and
189 subsequent cleaning of a fouling material the adhesion between the fouling layer and the surface will
190 change. It has been shown that US techniques can be used to determine the degree of adhesion
191 between a surface and a whey protein fouling layer (Collier et al., 2015). Ultrasonic wave propagation
192 is highly dependent on the temperature of the system in which it propagates (Al-Aufi et al., 2019). To
193 eliminate or minimise any temperature effects when utilising US techniques the temperature should
194 be kept constant during measurements, or recorded and accounted for during subsequent data
195 analysis.

196 **2.2 Experimental rig and materials**

197 An experimental laboratory rig was built to reproduce the cleaning process of a pipe inner wall (Figure
198 2). It was a square, closed channel rig constructed from Perspex with a Stainless Steel (SS) 430 bottom
199 wall. Stainless steel was selected for the fouling wall to represent industrially relevant materials. The
200 remainder of the rig was constructed from Perspex to enable visual access to the progression of the
201 cleaning. Although the majority of industrial piping is cylindrical, a flat section was used for these
202 experiments to aid the imaging of the cleaning process. One side of the rig was connected to a tap and

203 the other side to a drain. The main dimensions (in millimetres) and the parts of the rig are shown in
204 Figure 2.

205 Three different food materials were used to generate the fouling. These were tomato paste, gravy and
206 concentrated malt extract. The tomato paste was Napolina Double Concentrate Tomato Pure and the
207 ingredients were: tomatoes, acidity regulator (citric acid), density: 0.84 kg/m³. The Gravy was Bisto
208 Favourite Gravy Granules with the following ingredients: potato starch, maltodextrin, palm oil, salt,
209 wheat flour, colour (E150c), sugar, flavour enhancer (E621, E635), emulsifier (E322), density once
210 dissolved in water: 0.95 kg/m³. The concentrate malt was taken from a Coopers Real Ale beer kit with
211 the following ingredients: malted barley, hops, yeast and water, density: 1.12 kg/m³. The gravy was
212 prepared by mixing 10 grams of granules and 10 ml of tap water in a beaker at 70°C for 1 minute with
213 continuous stirring. The other materials did not require preparation. The fouling film was created by
214 depositing 15 grams of one of the food materials on to the centre of the bottom plate of the rig. It was
215 then spread evenly with a spatula to form a uniform layer of approximately 5 mm thickness. It was left
216 to dry (and in the case of the gravy, also cool down) for ten minutes before beginning any cleaning
217 experiments.

218 **2.3 Experimental method**

219 Once the fouling film was prepared, the rig was flooded slowly with water at the desired temperature.
220 This was performed to ensure the rig was the same temperature as the cleaning water to reduce the
221 effects of temperature variation on the US measurements during the acquisition of the data. After
222 one minute, the inlet tap was opened and water flowed through the experimental rig. The flowing
223 water was allowed to remove the fouling layer and the experiment finished when all of the fouling
224 was removed. The water flow rate was calculated by measuring the volume of water exiting the
225 experimental rig during a known time period. Nearly all CIP systems utilise a chemical cleaning stage.
226 As the fouling materials used in this work were not burnt onto the pipe section, chemicals were not
227 required to remove them. This current work is more representative of the initial pre-rinse stage of a
228 CIP cycle, which generally uses water at ambient temperature without the addition of chemicals.

229 **2.4 Instrumentation**

230 The ultrasonic transducer was a 5 MHz magnetic contact transducer supplied by Olympus. This
231 transducer had a circular active element of diameter 1.27 cm and an area of 1.27 cm². A US Box
232 provided by Lecoer Electronique was used to excite the transducer with an electronic pulse and
233 digitise the received signals. Temperature was recorded by a RTD PT1000 attached to a Pico
234 Technology PT-104 data logger. Images were recorded using two Logitech C270 3MP web cameras.
235 The US Box, PT-104 and web cameras were all attached to a laptop. Bespoke MATLAB software was
236 developed to control the hardware components and acquire the data. The two web cameras were

237 located above and at the side of the US transducer location on the rig to image the cleaning of the
238 fouling materials from two different perspectives (Figure 2).

239 **2.5 Signal and data processing**

240 Figure 3 displays a reflected signal, recorded from the US system. As can be seen the reflected signals
241 are not separated in time from the transmitted signal due to the conditions of Equation 3 not being
242 true. Figure 3 (a and b) shows that there is very little difference in the reflected signal from a clean
243 pipe or one fouled with food materials. However, small changes can be observed when smaller
244 sections of the received waves are studied (Figure 3 c-e). The clean pipe has the largest amplitude of
245 the reflected wave. The gravy and tomato have a slightly reduced amplitude when compared to the
246 clean pipe. The malt fouling has a larger difference than the other two fouling materials. Differences
247 between clean and fouled surfaces can be determined in all of the window locations in Figure 3 (c-e).
248 However, the greatest different is observed in the 4-4.5 μs location where the reflected signal has the
249 largest amplitude in general (Figure 3 c). Signals before 4 μs could not be analysed as these had
250 saturated. It would be possible to reduce the gain applied to the received signal but this would reduce
251 the duration of the received signal due to attenuation, which may make it indistinguishable from
252 noise.

253 Several different signal and data processing methods were utilised to analyse the received US signals
254 during the cleaning of the fouled material. The first two methods studied the amplitude and energy in
255 a windowed portion of the signal. The third method compared any variation in the signal to a signal
256 from a known clean pipe. The peak-to-peak amplitude (PPA) was calculated using the following:

$$257 \quad PPA = \max_{w_1 < t < w_2} (V(t)) - \min_{w_1 < t < w_2} (V(t)) \quad (4)$$

259 Where $V(t)$ is the voltage at time t , and w_1 and w_2 are the starting and ending times of the processing
260 window. Examples of different signal windows can be seen in Figure 3 (c-e). Relative peak-to-peak
261 amplitude (PPA_r) is the peak to peak amplitude when the wall is dirty (PPA_d) compared to when it
262 is clean (PPA_c). This was used to determine the most suitable signal window location and length.

$$264 \quad PPA_r = \text{abs}(PPA_c - PPA_d) / PPA_c \quad (5)$$

266 The energy E in the windowed signal was calculated using:

$$268 \quad E = \sum_{t=w_1}^{w_2} V(t)^2 \quad (6)$$

270 The relative energy (E_r) was used to determine the most suitable window location and length and
271 calculated using:

$$272 \quad E_r = \text{abs}(E_c - E_d) / E_c \quad (7)$$

273
274
275

276 Where E_c is the energy for a clean pipe and E_d is the energy for a dirty pipe. Root Mean Square Error
277 (RMSE) was used to compare the difference between two waveforms. This was calculated using:
278

$$279 \quad RMSE = \sqrt{\frac{1}{(w_2-w_1)} \sum_{t=w_1}^{w_2} (V_d(t) - V_c(t))^2} \quad (8)$$

280
281 Where $V_d(t)$ is the voltage when the wall is dirty at a time t and $V_c(t)$ is the voltage when the wall is
282 clean at the same time t . It was important to determine which portion of the received signals to
283 analyse using the methods proposed above. A section of signal can be analysed by applying a window
284 in the time domain. Example windows can be seen in Figure 3 (c-e). To determine the most suitable
285 location and size of this window analysis were performed using the RMSE (Equation 8) and relative
286 US peak to peak amplitude and energy values described above (Equations 5 and 7), for each of the
287 fouling materials. A received signal was recorded from the clean test section, flooded with water and
288 the fouled test section also flooded with water. Different window locations and sizes were then
289 applied to the received US signals to determine which window locations and sizes gave the greatest
290 difference for the three US analysis methods, when comparing the clean to dirty test section (Figure
291 4).

292 The results for the most suitable window location (Figure 4) do not show a clear result and ideal
293 location, with differences identified for the different fouling materials and US data analysis methods.
294 For relative amplitude, a location above 6 μ s appears to make no difference for the tomato and gravy
295 but does for the malt. For relative energy a peak in difference is present at 6 μ s however after this the
296 results show different trends for the three materials. The results for RMSE generally show a reduction
297 between 5 and 9 μ s. As no clear ideal location could be identified from this approach, a location of 6
298 μ s was selected as this appears to be a suitable compromise. The results for the window length show
299 that this has very little effect on the results so a value of 3 μ s was selected. The results in Figure 4 also
300 show that the presence of fouling has a slightly larger effect on the US energy than the US amplitude.
301 This is expected as the US energy uses all the points within the windowed signal whereas the US
302 amplitude only uses the maximum and minimum values.

303 **3. Results and Discussion:**

304 The US amplitude, the US energy and the RSME were normalized in between two values, in order to
305 plot them together using a single axis and compare the results between the different fouling materials.
306 The US amplitude (Equation 4) was normalised between 0-5000, the energy (Equation 6) was
307 normalised between 0 – 2.8×10^8 and the RMSE (Equation 8) was normalised between 0-300.

308 **3.1 Tomato paste**

309 The results for the US data recorded during the cleaning of tomato paste are shown in Figure 5. The
310 results show that as the tomato paste was cleaned the US amplitude and the energy increased whilst
311 the RMSE reduced. All three US data features changed at the same locations in time, which was
312 approximately 6:30 – 8:30 minutes. This indicates that all US data features calculated are sensitive to
313 the same aspects of the fouling and there is no benefit in calculating more than one US data feature.
314 The change in the US amplitude during cleaning was the lowest and approximately 55-60%. The
315 change in the US energy was approximately 70-80%, whereas the change in RMSE was approximately
316 18-2%. The changes in the US amplitude and energy were as expected; as the fouling materials have
317 different physical properties to the flowing water, so would reflect the ultrasonic waves differently
318 (Equations 1 and 2). The results in Figure 5 were consistent with those of Wallhäußer et al., 2013 who
319 found that the US energy in a reflected signal was lower in the presence of dairy fouling when
320 compared to a clean test section. In the first seven minutes of cleaning the coverage area and thickness
321 of the tomato fouling could be seen to reduce in the images but no changes were observed in the US
322 results (Figure 5). It should be noted that during the first seven minutes of the experiment the fouling
323 completely covered the area on the plate opposite to the US transducer's active element area (1.27
324 cm²). Therefore, the only change in that region was the fouling layer thickness. This suggests that the
325 US technique utilised within this work is not sensitive to changes in fouling layer thickness in the early
326 stages of cleaning or with fouling not directly above the sensor footprint. This is consistent with the
327 findings of E. Wallhäußer et al., 2012b. The RMSE technique is similar to the decorrelation coefficient
328 method used by B. Chen et al., 2019. However the decorrelation coefficient was used to compare
329 recorded signals from two sensors, one where fouling was present and one without fouling. In addition
330 the decorrelation coefficient only takes into account differences in phase or signal location and not
331 amplitude differences. Both the RMSE method used in this work and the decorrelation coefficient
332 used in Chen et al., 2019 showed comparable results. However, the results in Chen et al., 2019 showed
333 changes throughout the entire cleaning processes. This could be due to the nature of the material
334 they were cleaning (wax), the hardware components utilised or their signal and data processing
335 methods.

336 **3.2 Gravy**

337 The US results for gravy (Figure 6) showed the same trends as the results for tomato paste (Figure 5).
338 These were an increase in US amplitude and energy and a reduction in RMSE during cleaning. All final
339 values for the three US features were also a similar value as the results for tomato (Energy = 90%,
340 Amplitude = 65%, RMSE = 5%). The RMSE showed a significantly larger change throughout cleaning,
341 which began at a normalised value of 90% before increasing during the first three minutes. Following
342 this, the RMSE reduced quickly between three and five minutes before a more gradual decrease to 11

343 minutes when the fouling had been removed. The US amplitude and energy both began at lower initial
344 values than the tomato paste (55% and 42% respectively). Once cleaning began these both reduced
345 for three minutes before increasing until the fouling was removed. From the images, it was observed
346 that during these first few minutes the gravy fouling swelled due to absorption of water, which may
347 be responsible for these initial changes in the US results. The initial swelling of the fouling layer has
348 been identified by numerous researchers and was included in the cleaning model for protein fouling
349 proposed by Xin, Chen, & Özkan, 2004. The recorded images suggest cleaning differences between
350 the tomato and the gravy. The tomato appeared to gradually reduce in volume whereas the gravy
351 swelled became partially detached from the test section, with visible movement, before becoming
352 detached at approximately 11 minutes. The gravy results for all US data features show a significant
353 amount of localised variation during the cleaning. These localised variations could be caused by the
354 movement of the swollen gravy or the presence of small bubbles adhering to the surface and the gravy
355 which, were observed. This localised variation is not likely caused by noise as it was not present at the
356 beginning (0-4 minutes) or end of the results (11-15 minutes).

357 **3.3 Malt**

358 The US results for the cleaning of malt (Figure 7) showed the largest variation in the US data features
359 when compared to the tomato paste (Figure 5) and gravy (Figure 6). The results for malt do appear to
360 follow the same general trends as the tomato, which is an increase in US amplitude and energy and a
361 reduction in RMSE. The initial values for US amplitude was 30%, for US energy 20% and RMSE 90%.
362 The change in the US data features is the smoothest of the three materials studied and occurs between
363 seven and ten minutes. It was observed that the nature of the cleaning of malt was very different to
364 the tomato and gravy. Although mechanical actions performed some role, it appeared from the
365 images that the malt mainly dissolved into the water. Figure 7 presents numerous images from the
366 two cameras between seven and ten minutes to determine which aspects of the fouling layer had the
367 largest effects on the US reflected signals (Images 5-17 in Figure 7). During the first seven minutes the
368 fouling layer appears to reduce in thickness (images 1-4 Figure 7) but no change is observed in the US
369 results. This supports the argument that fouling layer thickness has little effect on the received US
370 signals during the initial stages of cleaning, with the current US system. During the time periods where
371 the US data features change, the images show that the largest change is the coverage area of the
372 fouling material (Images 5-17 in Figure 7). This supports the argument that the US results are most
373 sensitive to the area of fouling covering the transducer's active element area and not the fouling layer
374 thickness. Ultrasonic transducers with larger active areas may therefore be more suitable for
375 monitoring cleaning processes. However, practical challenges remain here as the active element area
376 is usually determined by the frequency of the transducer, with higher frequency transducers preferred

377 in thin walled systems to reduce the effects of overlapping signals (Equation 3). Larger diameter
378 transducer may also have contact issues with small diameter pipework.

379 Fryer and Asteriadou presented two classification methods for cleaning problems. The first method
380 was based on the fouling material type and the second on the cleaning mechanism (Fryer &
381 Asteriadou, 2009). For the method based on the material type all of the fouling materials reported in
382 this current work are type 1 (of the three types they specify). Type 1 fouling are defined as viscoelastic
383 fluids, which can be rinsed from a process surface with water. For the second method based on
384 cleaning mechanism the different types of cleaning were define as *fluid mechanical removal* and
385 *diffusion reaction* removal. Fluid mechanical removal is the removal of surface fouling using the force
386 provided by the fluid. From our results this appears to be the cleaning mechanism of the tomato and
387 gravy. The results indicate that the US data features for this type of cleaning display a sudden change
388 in the recorded US data features and some localised temporal changes as the material is been
389 removed (Figures 5 and 6). Fryer and Asteriadou, 2009 define diffusion reaction removal as the
390 diffusion of a chemical into the fouling material to aid in its removal. Although no chemicals were used
391 in this current work the fouling removal of the malt appears to follow this concept as the water was
392 dissolving the fouling layer. The US results for this type of cleaning, show a gradually changing curve
393 with little or no localised variation in values. When comparing the three US data analysis methods it
394 appears that all are suitable for detecting the presence of fouling and monitoring certain stages of
395 cleaning. All three methods vary in the same time domains but the RMSE and US energy methods are
396 preferable as they show the largest variation in results during cleaning, potentially making them more
397 suitable for systems with a lower degree of fouling.

398 **3.4 Repeatability and temperature effects**

399 Figure 8 displays the US energy for repeat experiments for the three fouling materials at low (12 °C)
400 and high (45 °C) temperature. All energy values returned to the same value of 1.9×10^8 EU for the
401 experiments at 12°C and 2.1×10^8 EU for the experiments at 45°C, once the fouling materials had been
402 removed. This supports the earlier assertion that the developed US system can repeatedly monitor
403 the cleaning of a range of different fouling materials at ambient and elevated temperatures. Of the six
404 results presented in Figure 8 the only ones that did not consistently return to the same final energy
405 value was the tomato at high temperature. These were the first experiments that were performed
406 and it is likely that the contact between the transducer and the wall had not yet stabilised. The
407 importance of having transducers that are either fixed in place or have been given enough time to
408 form a stable contact before performing experiments cannot be overstated. The malt had the most
409 repeatable cleaning profile with very little variation in cleaning time or data trends between runs
410 (Figure 8). The malt cleaning experiments at higher temperature showed that surface fouling was

411 removed in a much faster time of approximately two minutes compared to approximately 13 minutes
412 for the lower temperature (Table 2). This supports the theory that the malt cleans by dissolving as it
413 is known that dissolution occurs faster as higher temperatures due to the increase in kinetic energy.
414 It has also been reported in the literature that temperature increases the cleaning rate of fouling from
415 numerous experimental studies (Fryer et al., 2006).

416 The effect of temperature on cleaning time for tomato cannot be studied quantitatively as a lower
417 flow rate was used for the high temperature experiments (Table 1). This lower flow rate was used as
418 the tomato cleaned almost instantaneously at the faster flow rate and higher temperature. The gravy
419 was difficult to clean with water at low temperature so only two repeats were performed. These two
420 experiments had very different cleaning times of approximately 40 and 100 minutes. The results in
421 Figure 8 and Table 2 indicate that cleaning processes, which are primarily dependent on mechanical
422 forces for removal, appear to be much more variable than those dependent on dissolution.

423 **4. Conclusions:**

424 Cleaning of processing equipment is essential to ensure hygienic and optimal processing conditions
425 within industries such as the food, pharmaceutical and FMCG. However, current CIP processes are
426 often inefficient, over cleaning equipment, with significant negative financial and environmental
427 impacts. In this work an US sensor technique capable of monitoring the cleaning of different food
428 fouling materials was developed and experiments performed in a bespoke laboratory rig. It has been
429 shown that the developed US technique was capable of monitoring the fouling removal of the three
430 food materials, which all clean from the test section differently. It was shown that the US energy and
431 RMSE of a windowed section of the received US signal were the most suitable data analysis techniques
432 to use. Repeat experiments were performed at two temperatures (12°C and 45°C) and almost
433 identical values of the US data features were obtained at the end of each experiment, once the test
434 section was cleaned. This demonstrates the potential of the US technique for a range of different
435 fouling materials. The tomato paste was found to clean gradually mainly by mechanical force, the
436 gravy swelled due to the presence of water and was removed in a single lump whereas the malt slowly
437 dissolved into the water. The time to clean was relatively similar for repeat experiments of the malt
438 but much more variable for the tomato and gravy. Analysis of the images and the US results indicated
439 that variations in the US signals were more sensitive to the amount of fouling material within the area
440 covered by the US transducer's active element than the fouling layer thickness. This indicates that the
441 current technique may be limited in monitoring the initial stages of cleaning.

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