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1 **Cross-continental comparative experiences of wastewater surveillance and a**  
2 **vision for the 21st Century**

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25 There are no competing interests to declare.

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## 31 **Author contributions**

32 Funding acquisition ID, AJW, OJ

33 Conceptualization ID, CCM, WBP, MIZS, MCC, AJW, IBJ

34 Project administration WBP, IBJ

35 Investigation WBP, ID, MCC, MRFB, MIZS, OJ

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38 Writing - review & editing all authors.

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43

## 44 **Abstract**

45 The COVID-19 pandemic has brought the epidemiological value of monitoring wastewater into sharp  
46 focus. The challenges of implementing and optimising wastewater monitoring vary significantly from  
47 one region to another, often due to the array of different wastewater systems around the globe, as well  
48 as the availability of resources to undertake the required analyses (e.g. laboratory infrastructure and  
49 expertise). Here we reflect on the local and shared challenges of implementing a SARS-CoV-2  
50 monitoring programme in two geographically and socio-economically distinct regions, São Paulo state  
51 (Brazil) and Wales (UK), focusing on design, laboratory methods and data analysis, and identifying  
52 potential guiding principles for wastewater surveillance fit for the 21<sup>st</sup> century. Our results highlight the  
53 historical nature of region-specific challenges to the implementation of wastewater surveillance,  
54 including previous experience of using wastewater surveillance, stakeholders involved, and nature of  
55 wastewater infrastructure. Building on those challenges, we then highlight what an ideal programme  
56 would look like if restrictions such as resource were not a constraint. Finally, we demonstrate the value  
57 of bringing multidisciplinary skills and international networks together for effective wastewater  
58 surveillance.

59 Key words: COVID-19, SARS-CoV-2, São Paulo, Brazil, Wales, One Health, Wastewater Based  
60 Epidemiology

## 61 **1. Introduction**

62 A vital tool in the global response to the COVID-19 pandemic has been wastewater surveillance (table  
63 1). Monitoring viral (SARS-CoV-2) load in the sewers provides estimates of the infection rate in a  
64 community and trends in virus circulation in the population. Unlike traditional public health surveys  
65 based on individual testing, viral load in wastewater has been shown to provide a relatively low-cost  
66 estimate of disease prevalence that are not biased by testing capacity or behaviours, and that often  
67 precede public health data by a few days (Kumar et al., 2021). Efforts to implement wastewater  
68 surveillance to track SARS-CoV-2 across the globe have been met with different types of challenges.  
69 In some regions, such as São Paulo state, Brazil, there has been a long tradition of utilising wastewater

70 surveillance to track the prevalence or the outbursts of serious diseases, in particular, poliovirus, as a  
 71 supplementary approach to the Global Polio Eradication Initiative (De Melo Cassemiro et al., 2016;  
 72 Martins et al., 1983; WHO, 2014, 2022). However, in many other countries, such as the Wales,  
 73 wastewater monitoring for public health surveillance had never been routinely implemented other than  
 74 for environmental monitoring. This is surprising considering the demonstration as an effective  
 75 monitoring tool for infection outbreaks in the UK over 75 years ago (Moore, 1948, 1950).

76 **Table 1. Examples of SARS-CoV-2 monitoring programmes from around the globe. Details of**  
 77 **these programmes are provided, such as the number of sample sites, the population coverage**  
 78 **(where available) as well as the frequency of sampling per week. Data was extracted from various**  
 79 **sources for each of the programmes, as outlined in the reference column. Some of the estimates**  
 80 **are not current, and instead reflect peak coverage. End date is characterised by the national**  
 81 **programme finishing, being drastically reduced, or no longer providing data to their public**  
 82 **dashboard.**

Country or region	Reference	Number of sites	Population coverage (%)	Frequency of sampling per week	End date
England	(UK Health Security Agency, 2022)	302	74	3	March 2022
Scotland	(Fang et al., 2022)	120	80	3 to 4	Ongoing
Northern Ireland	(S. Bell et al., 2022)	31	62	2	March 2023
Netherlands	(van Boven et al., 2023)	300	99.6	1 to 4	Ongoing
Switzerland	(SWI swissinfo.ch, 2022)	117	~70	3	Ongoing
Austria	(BMSGPK, 2022)	48	58	2	Ongoing
Germany	(Robert Koch Institute, 2023)	175		2	Ongoing
Turkey	(Turkish Ministry of Agriculture and Forestry, 2023)	189	67	1 to 2	Ongoing
Israel	(Bar-Or, 2022; Bar-Or et al., 2022)	135	>55%	2	July 2023
Pune, India	(The Pune Knowledge Cluster, 2024)	32		2	Ongoing
Hong Kong, China	(Chui, 2023)	154	80	3	Ongoing
South Africa	(SAMRC, 2023)	76		1	April 2023
Ottawa, Canada	(Delatolla et al., 2024)	2	92	7	Ongoing

83 Besides previous experience, local or regional factors are likely to have also played significant roles in  
84 the way this type of surveillance system is implemented or optimised. Examples of such factors include:  
85 the nature and extent of existing wastewater infrastructure, the governance infrastructure and data  
86 sharing, the nature of the relationship between different actors that need to interact (e.g., water utilities,  
87 researchers, public health organisations), or the availability of financial resources, testing facilities and  
88 trained personnel. The success of wastewater surveillance programmes depends on their capacity to  
89 innovate and overcome challenges to adapt to a changing world. A 21<sup>st</sup> century wastewater surveillance  
90 system will be faced with significant global scale challenges. Some are linked intrinsically to water and  
91 wastewater management systems, including climate change, rising sea levels, emerging pollutants,  
92 rising demographics and concerns over public health, biodiversity loss and ecosystem services. Other  
93 challenges will be prompted by epidemiological concerns, such as new SARS-CoV-2 variants, the shift  
94 from pandemic to endemic status of COVID-19, the emergence of future pandemics as well as the  
95 routine monitoring of other diseases and biomarkers of human health. Beyond challenges, water  
96 surveillance programmes also hold great opportunities, namely the potential for meeting the United  
97 Nations Sustainable Development Goals, contributing to sustainable cities and communities; industry,  
98 innovation, and infrastructure; clean water and sanitation; as well as good health and wellbeing.

99 Here we reflect on the local and shared implementation challenges of the SARS-CoV-2 monitoring  
100 programmes of two geographically and socio-economically distinct regions: São Paulo state (Brazil's  
101 most populous state) and Wales (one of the four countries making up the United Kingdom), to identify  
102 potential guiding principles of a wastewater surveillance programme fit for 21<sup>st</sup> century scenarios.

103 The benefit of using these two examples is that they have both approached wastewaters surveillance in  
104 systems with differing structural constraints. These structural constraints include historical context that  
105 have led to the present-day wastewater infrastructure, as well as social, governance and economic  
106 landscapes. Therefore, understanding the evolution of these two wastewaters surveillance programmes,  
107 created under different selection pressures, can provide valuable insights into overcoming common  
108 challenges while also providing a roadmap to achieve the characteristics of an ideal international  
109 wastewater surveillance programme fit for the 21<sup>st</sup> century.

110 To structure this reflection, here we:

- 111 1. Provide a brief overview of the historical development, current infrastructure and governance  
112 that shape the current wastewater surveillance programmes of São Paulo state and Wales.
- 113 2. Compare SARS-CoV-2 wastewater surveillance in terms of design, laboratory methods, data  
114 analysis and utilisation.
- 115 3. Imagine what an ideal SARS-CoV-2 wastewater surveillance system would look like if  
116 resources and logistics were not a constraint, reflecting on opportunities for future wastewater  
117 monitoring.

## 118 **2. History of wastewater and wastewater surveillance in São Paulo state and Wales**

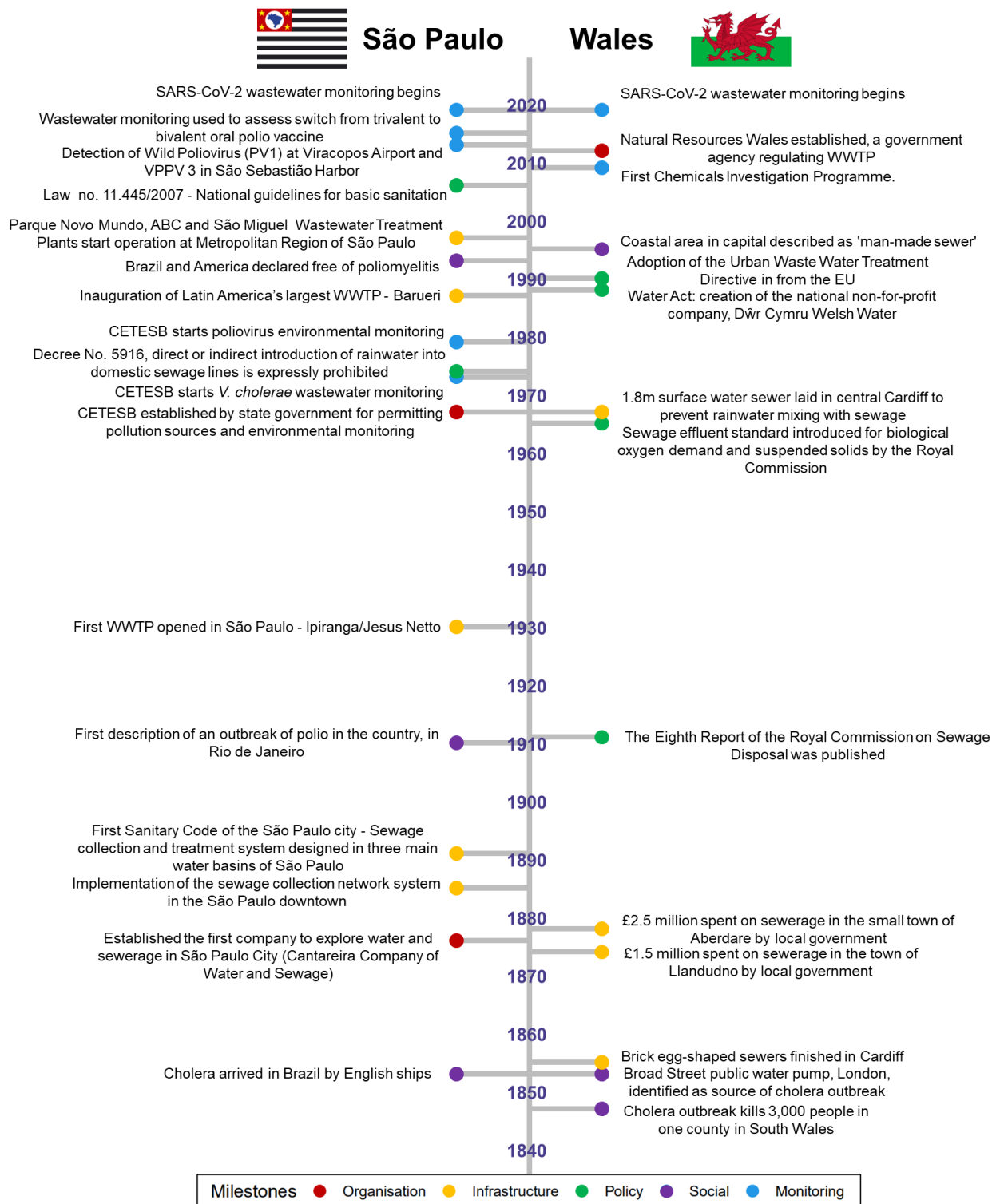
119 In response to the SARS-CoV-2 pandemic, wastewater surveillance programmes have been put in place  
120 across the globe, each example having been shaped by wastewater infrastructure, socio-economic  
121 context and governance (Arora et al., 2022; Carcereny et al., 2021; Izquierdo-Lara et al., 2021; Tihagale  
122 et al., 2022). São Paulo state (Brazil) and Wales (United Kingdom) provide models to assess how these  
123 structural constraints may shape wastewater surveillance programmes and thus Wastewater Based  
124 Epidemiology (WBE).

### 125 **2.1 Evolution of two contrasting wastewater systems – historical context for surveillance**

126 By the turn of the 20<sup>th</sup> Century, both São Paulo state and Wales had emerged from the Sanitary  
127 Enlightenment with the foundations of a modern wastewater system, forged by shifts in organisation,  
128 infrastructure, policy, society, and monitoring (Fig. 1). Due to high levels of rainfall in Brazil, especially  
129 in summer, the direct or indirect introduction of rainwater into domestic sanitary sewer branches is  
130 expressly prohibited (Decree No. 5916/75 in the State of São Paulo). Although, in many areas a fully  
131 separated system is not always achieved (FUNASA, 2019) due to illicit discharges of rainwater into the  
132 wastewater system and water entering old wastewater pipes in times of high rainfall (Fig. 2b,e,h). In  
133 São Paulo, whenever there is a public sewer system in service conditions, industrial effluents must be

134 discharged into it, in which case it is necessary to comply with specific legislation (*Decreto n.8.468,*  
135 1976) (Fig. 2b). If it is not feasible to connect to the public system, the effluent may be released into  
136 the water body if it does not alter its conditions and meets specific emission standards (*Decreto n.8.468,*  
137 1976; CONAMA, 2011)(Fig. 2j). Alternatively, Wales has a largely combined wastewater system,  
138 where wastewater (e.g. blackwater, greywater, industrial (Fig. 2b), clinical (Fig. 2a)) are actively  
139 combined with rainwater. This creates a large volume of wastewater with a cocktail of contaminants,  
140 whose removal can be highly variable (Comber et al., 2019) due to dilution capacity, outdated facilities  
141 (Gardner et al., 2012) and the expense of effective methods of contaminant removal (Rout et al., 2021).





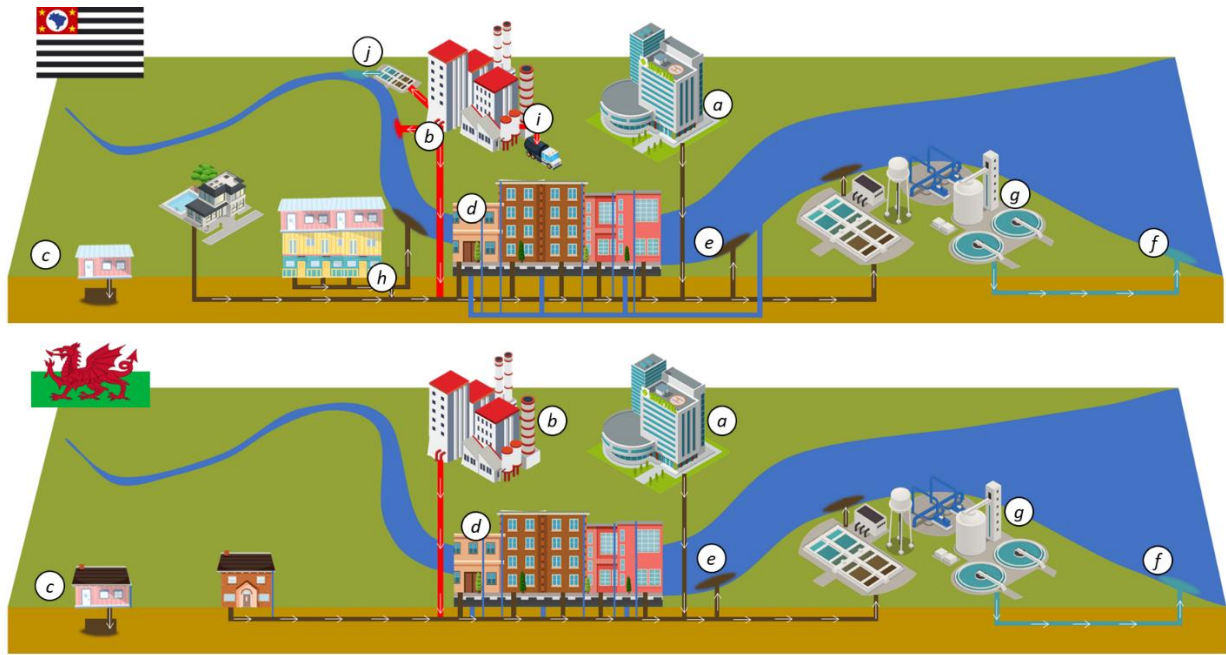
142

143 **Figure 1 Timeline of milestones (coloured by the nature of the event) in the establishment of the**  
 144 **wastewater systems in São Paulo state (left) and Wales (right).**

145 In both São Paulo state and Wales, not all wastewater reaches a wastewater treatment plant (WWTP)  
146 (Fig. 2g). The main indicators for São Paulo state shows that 64.5% of the sewage is collected and  
147 treated (Fig. 2f,g), 22.6% is collected but not treated, 9.2% is not collected nor treated (Fig. 2e,h), while  
148 the remaining 3.7% correspond to individual local solutions (Agência Nacional de Águas, 2021) (e.g.  
149 septic tanks, rudimentary pits, open sewers, the launching of wastewater into watercourses and  
150 rainwater galleries (Stepping, 2016)). Therefore, a significant portion of the population is ‘off-grid’,  
151 and left without adequate sanitation services, especially in the many of the irregular/unplanned  
152 settlements and favelas, which have high population densities and levels of deprivation.

153 Like São Paulo, Wales is home to many ‘off-grid’ settlements, but these settlements are rarely found in  
154 urban zones, but located in low density rural areas like the uplands. In addition to these rural settlements,  
155 there are also many irregular settlements which are primarily used for tourism. These include large areas  
156 of static caravans and camp sites, which have a highly seasonal population, often found on the coast,  
157 which also rely on holding infrastructure such as septic tanks. In total, there are 7,116 registered septic  
158 tanks in Wales, however, little is known about the number of unregistered tanks, or how much waste is  
159 released from leaky tanks. Based on household mapping in rural regions it is estimated that ca. 10% of  
160 the population in Wales are not connected to mains sewerage. Finally, in Wales, outdated brick-lined  
161 Victorian networks still support a large portion of the wastewater network (Heathcote et al., 2003),  
162 which due to increasing wastewater volumes, have exceeded capacity. When capacity is exceeded  
163 combined sewer overflows (CSOs) are used to remove excess wastewater, allowing the wastewater to  
164 flow, untreated, into the environment (Perry et al., 2024), a process which is likely to worsen with  
165 climate change (Abdellatif et al., 2015; Petrie, 2021; Zan et al., 2023).

166



167

168 **Figure 2 Diagram of wastewater systems in Wales (UK) (top) and São Paulo state (Brazil)**  
 169 **(bottom), highlighting wastewater sources (e.g., (a) hospitals, (b) industry, (c) off-grid domestic,**  
 170 **(d) high density domestic with surface runoff, (h) informal high density domestic with wastewater**  
 171 **misconnections) and outputs (e.g., (c) individual local solutions, (e) untreated release, (f,g) WWTP**  
 172 **treated release, (i) transport to specialty treatment and (j) on-site WWTP).**

173 **2.2 Social, governance and economic contexts for wastewater surveillance**

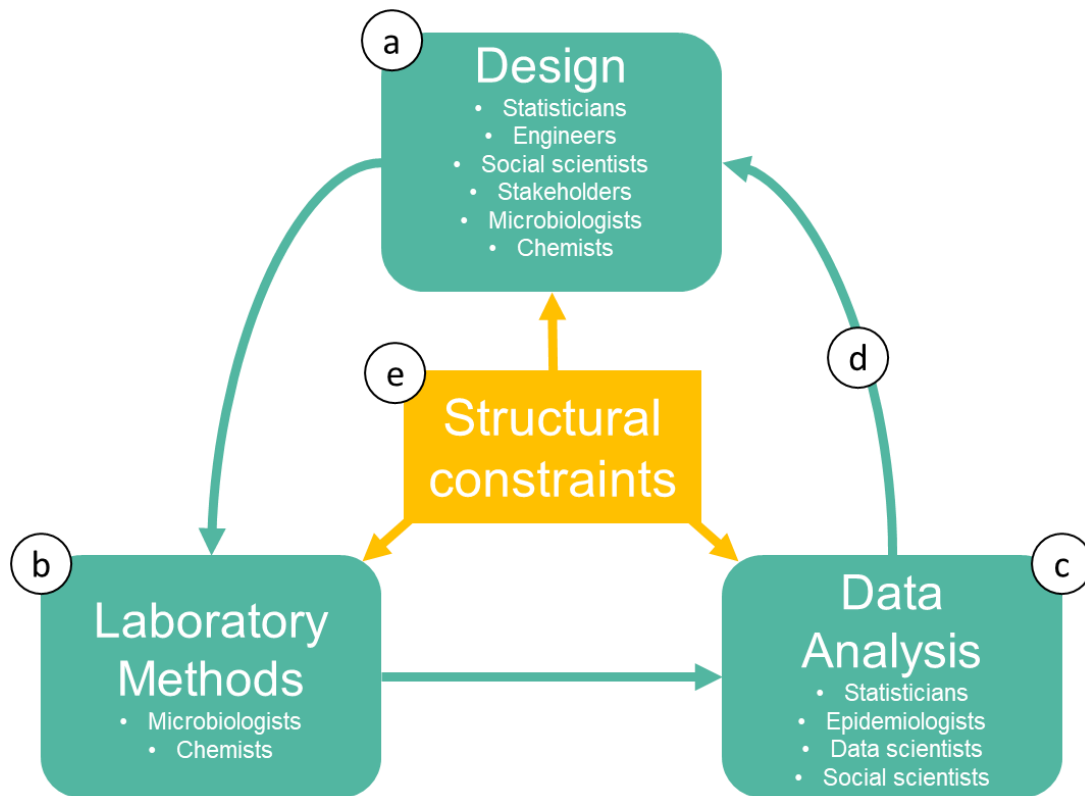
174 Geographic size, demographics and history have shaped different contexts for wastewater surveillance  
 175 in São Paulo state and Wales. Consequently, the governance around wastewater and public health is  
 176 very different between them. In São Paulo state, multiple branches of government are involved in policy  
 177 related to wastewater which is not the case in Wales. São Paulo state also has a multitude of  
 178 governmental regulators, each of which will have different priorities, including sanitation services,  
 179 monitoring of waterborne infectious diseases or wastewater discharges into the environment. Wales  
 180 only has two government sponsored bodies to regulate wastewaters, one mainly focused on the  
 181 environmental impact of wastewater discharge and the other on consumer prices. Economically,  
 182 wastewater service providers are also diverse in São Paulo state with 60% of municipalities serviced by  
 183 a government owned company, and the other 40% of municipalities serviced by local public and private

184 services. In contrast, in Wales, most of the population is served by one not-for-profit company (Dŵr  
185 Cymru Welsh Water), with a smaller second private company (Hafren Dyfrdwy) providing service to a  
186 population approximately 7% of the population served by Dŵr Cymru Welsh Water (Hafren Dyfrdwy,  
187 2019).

### 188 **3. Comparison of SARS-CoV-2 wastewater surveillance programmes**

189 Building on the body of knowledge on WBE accrued from countries and regions where surveillance  
190 programmes have been in place for a long time, such as São Paulo, SARS-CoV-2 wastewater  
191 surveillance programmes have required a step-change to meet the needs for near real-time and country-  
192 wide assessments. The development of a wastewater surveillance programme, be that São Paulo state  
193 or Wales, can be split into three broad categories, design, laboratory methods and data analysis (Fig.  
194 3a-c). These fundamental components will be used to structure the comparison here, acknowledging  
195 the fact that they are all interconnected and influenced by structural constraints of the system (e.g.  
196 funding, resources, logistical constraints, desired outcomes and type of wastewater system) (Fig. 3e).  
197 Structural constraints cannot be easily changed, and will not be the focus here.

198 Each component of the wastewater surveillance programme requires multidisciplinary expertise (Fig.  
199 3a,b,c), for example, molecular biologists are required to develop the methodology for detecting and  
200 quantifying pathogens of interest in wastewater; their knowledge will influence the design of the  
201 programme, such as where samples can be taken and how long they can be stored and best transported.  
202 Another example is statisticians and bioinformaticians, who can extract and interpret relevant trends  
203 through data analysis, but those insights should also feedback into the design (Fig. 3d) of a programme,  
204 as they can influence how variance is better accounted for. In this section, we compare the outcome of  
205 these multidisciplinary interactions in the programmes in São Paulo state and Wales.



206

207 **Figure 3 Schematic representation of the main interacting components needed to build an**  
 208 **operational wastewater surveillance programme: (a) design, (b) laboratory methods, (c) data**  
 209 **analysis, as well as the (e) feedback loop finalised between data analysis and design. Also included**  
 210 **is the influence of (e) structural constrains on each of those components.**

211 **3.1 Design**

212 **3.1.1 Stakeholders**

213 Decisions on choosing sample sites are made by a combination of stakeholders in both São Paulo state  
 214 and Wales. In São Paulo state, sampling design was conceived between the wastewater utility, São  
 215 Paulo State Sanitation Company (SABESP), who provided logistics and wastewater flow data, the  
 216 Epidemiological Surveillance Centre, who brought their knowledge of transmission routes, and the  
 217 Environmental Company of São Paulo State (CETESB), who provided environmental surveillance  
 218 expertise. University of São Paulo and UNICAMP participate in the process in the support of data  
 219 analysis. Wales followed a similar multi-stakeholder development, which included Dŵr Cymru Welsh

220 Water and Hafren Dyfrdwy, who provided access to WWTPs and wastewater flow data, Public Health  
221 Wales, who provided clinical expertise, academics at Cardiff University and Bangor University, who  
222 provided environmental surveillance. Finally, the end users, the Technical Advisory Cell of Welsh  
223 Government, who used insights gathered to advise on public health policy, with insights from the  
224 wastewater surveillance ultimately preventing two lockdowns in 2021 and 2022 during the Christmas  
225 period.

226 Criteria driving the decision-making regarding sampling locations were similar in Wales and São Paulo:  
227 size of the population represented by the sample, and logistics such as site accessibility. Previous  
228 monitoring programmes in São Paulo state had focused on faeco-orally spread enteric pathogens, with  
229 transmission through contact and contamination of water and food (e.g. enteroviruses). Therefore,  
230 previous wastewater sampling designs reflected the previous transmission types, focusing on areas with  
231 poor sanitation and low immunisation rates, as well as reflecting the endemic, rather than pandemic,  
232 nature of the pathogens, focusing on entrance sites (e.g. airports, harbours, bus stations). SARS-CoV-2  
233 required a different sampling approach, as it is primarily a respiratory disease and its presence was far  
234 more ubiquitous, which meant sampling efforts were largely concentrated on WWTPs, although not  
235 entirely. In Wales, there had been no previous wastewater monitoring for health, but like São Paulo,  
236 Wales focused on WWTPs, even more so than São Paulo.

### 237 **3.1.2 Sampling locations**

238 In São Paulo, out of a total of 24 sites, 14 were located within the conurbation of the MRSP as half of  
239 the population of the São Paulo state live in the MRSP and since the beginning of the pandemic the  
240 largest number of cases of COVID cases was concentrated in this region. Seven sites were located on  
241 the Tietê river system that crosses the city of São Paulo which are directly influenced by the releases  
242 from the MRSP sanitary drainage basins, and three at the WWTPs of three large cities in the interior of  
243 the State. However, due to the organic growth of the RMSP and the complexity of its watersheds, the  
244 selection of sampling points was a challenge, and high priority areas included high-density urban areas  
245 and regions with vulnerable populations, such as slums, where people are concentrated in a small indoor

246 space (Barbosa et al., 2022). Effluents from hospitals and WWTPs from cities that received large  
247 immunisation campaigns (Serrana and Botucatu) were also monitored for a year. More recently, human  
248 entry sites such as international ports, airports and bus stations were included in the wastewater  
249 surveillance program. A focus on transmission routes is also what prompted the São Paulo state  
250 programme to choose a sampling site on the coast of São Paulo city, where the largest port in Latin  
251 America is located, and where the base population of around 800,000 doubles in the summer season.  
252 When lockdown measures are suspended, sites that see high levels of population influx such as tourism  
253 hotspots are key to follow epidemiological sources and transmission routes.

254 The programme in Wales initially started with 19 WWTP sites, and in January 2022, the number of  
255 sites increased to 47, representing approximately 66% of its population (2,070,883 people). Due to  
256 resource constraints, the number of sites was however decreased again in November 2023 to 11  
257 representative WWTPs, after almost four months of no sampling. The largest of these sites, and the  
258 capital, Cardiff, serves an estimated 602,990 people, and the smallest site, Betws-y-Coed in north Wales  
259 servicing an estimated 375 people. Wales also regularly monitored four prisons, due to the congregation  
260 of people in a small indoor space, as well as it being an isolated community of vulnerable people. In  
261 addition to these sites, near to source sampling has previously conducted on an ad hoc basis, including  
262 sea/airports due to their role in the entry of new SARS-CoV-2 strains. Further, the programme was  
263 expanded in 2022 to include the routine monitoring of wastewater from 10 hospitals covering ca. 40%  
264 of the patient population in Wales, although this was cut completely in July 2023 due to resource  
265 constraints. Periodic sampling of other near-source environments has also been undertaken on a limited  
266 basis (e.g. meat processing plants, mass quarantining hotels, university halls of residence). Unlike  
267 England, a decision was not made to monitor other near-source environments of perceived high  
268 transmission risk (e.g. schools, care homes) due to a lack of perceived benefit relative to the cost.

269 The trade-off between site representativity and site accessibility were important considerations for both  
270 programmes. In the São Paulo state programme, 8 sites out of 24 were located within WWTP basins:  
271 the influent of the five WWTP that served the RMSP, and the WWTPs of the three inland cities, unlike  
272 Wales, where all sites were located at the influent of WWTPs, other than some sampling of ports and

273 prisons. This is linked to the difference in scale between São Paulo state and Wales, both in terms of  
274 population size (Wales = 3.14 million people, São Paulo state = 46.65 million people) and geographic  
275 size (Wales = 20,779 km<sup>2</sup>, São Paulo state= 248,220 km<sup>2</sup>), and so even though Wales focused almost  
276 entirely on WWTPs, it could cover a greater proportion of the population, including sparsely populated  
277 areas. The Wales programme is different from many other WBE programmes that solely focused on  
278 capturing large urban areas, as it actively sought to understand how the disease evolved and moved  
279 between urban and rural communities. Wales has a high density of WWTPs (total = 833) over a small  
280 geographic area and population. On average, there is a WWTP for every 3,765 people (Dŵr Cymru  
281 Welsh Water, 2023), a granularity which facilitates the accessibility and surveillance of smaller  
282 populations, unlike São Paulo state, where relatively fewer WWTPs (total = 897) service a larger  
283 population, equating to a WWTP for every 49,097 people on average (Agência Nacional de Águas,  
284 2019). Sampling within the network of a large WWTP catchment to capture smaller sections of the  
285 population can be more difficult due to accessibility. In addition, in Wales, there is, on average, one  
286 WWTP for every 25km<sup>2</sup>, whereas in São Paulo state there is one WWTP for every 276km<sup>2</sup>.

287 To ensure population representativity, the São Paulo state programme samples sewage pumping  
288 stations, accessible sites in the sewage network through manholes, accessible sewage sites from  
289 hospitals, and accessible sewage or open river flows where infrastructure is scarce or absent such  
290 vulnerable favela communities. For example, SARS-CoV-2 wastewater surveillance was carried out in  
291 São Remo (May to November 2020) and Paraisópolis (May 2020 to June 2021) communities in São  
292 Paulo city, which have a population of approximately 8,000 and 51,000 inhabitants, respectively. These  
293 communities are supplied with treated potable water, but not connected to a sewer system. Wastewater  
294 is therefore discharged into streams which run to Rio Pinheiros. Samples collected from these streams  
295 revealed a correlation between cases of COVID-19 cases in the communities and SARS-CoV-2  
296 concentrations in the streams (Barbosa et al., 2022; Pepe Razzolini et al., 2021). A similar study in  
297 Wales has also shown the presence of SARS-CoV-2 RNA in WWTP effluent with hydrodynamic  
298 modelling predicting that it can travel many kilometres downstream from the WWTP eventually  
299 reaching the coastal zone (Hillary et al., 2021; Robins et al., 2022).



300 Many rural households in Wales are ‘off-grid’ and rely on septic tanks or rudimentary pits for  
301 wastewater collection (Fig. 2c), however, the percentage this makes up of overall wastewater volume  
302 is unknown. A definitive publicly available estimated percentage of wastewater that is released directly  
303 into the environment is also unknown (Fig. 2e). The Welsh programme does not sample ‘off-grid’  
304 communities, which would involve sampling cesspits, or untreated wastewater released into rivers. This  
305 has not been considered as a priority because ‘off-grid’ communities, while often the most deprived, are  
306 also living in the most sparsely populated areas of Wales where transmission rates are lowest. In São  
307 Paulo, on the other hand, ‘off-grid’ communities live in some of the most densely populated areas of  
308 the region where transmission rates are likely to be the highest.

### 309 **3.1.3 Sampling timings**

310 Given previous expertise, initial monitoring of SARS-CoV-2 started 39 days after the first case of  
311 COVID-19 was recorded for the São Paulo state programme. In Wales, the first wastewater samples  
312 were taken for analysis a few days after citizens were evacuated from Wuhan, China and arrived back  
313 in the UK (March 2020), however, this was confined to 3 locations in Wales (Cardiff, Wrexham, Bangor  
314 (Hillary et al., 2021)). At this stage the use of WBE for tracking SARS-CoV-2 was unproven and as  
315 such this was the validation phase that led to the development of the national programme in both  
316 England and Wales (Tlhagale et al., 2022). It took a further 6 months for the programme to be expanded  
317 more widely in Wales, and there was reliance on grab sampling. Two WWTPs in the São Paulo state  
318 programme (ABC and São Miguel) were also already equipped to provide 24-hour composite samples,  
319 which meant that they could be sampled immediately. However, in Wales, composite samplers were  
320 not being installed until November 2021, approximately 1 year and 8 months after the first case of  
321 COVID-19 in Wales, which meant the programme was reliant on grab samples for a large portion of  
322 the pandemic, which are a snapshot in time, and less representative than the longer time periods captured  
323 by composite samples. Currently, both programmes utilise a mix of grab and composite sampling, but  
324 due to constraints at some of the São Paulo state sites, they are unsuitable for permanent composite  
325 samplers. The Welsh programme, however, is predominantly based within WWTPs, and has now  
326 installed refrigerated composite sampling across all its sites. In terms of sampling frequency, the São

327 Paulo state programme samples bi-weekly whereas the programme in Wales did samples five times a  
328 week and was reduced to three times a week in November 2023, but this choice was primarily resource  
329 driven.

## 330 **3.2 Laboratory methods**

### 331 **3.2.1 Concentration**

332 The initial step in processing wastewater for molecular methods is concentration and removal of large  
333 debris. Originally, the São Paulo state programme used an ultracentrifugation and glycine elution  
334 method (Pina et al., 1998), unlike in Wales where an overnight polyethylene glycol (PEG) precipitation  
335 is used (Farkas et al., 2021). The latter method was chosen in Wales based on previous studies isolating  
336 a range of viruses from wastewater and the ability to process large volumes of wastewater (100-250 mL  
337 (Farkas et al., 2018)). Both ultracentrifugation and PEG precipitation methods offer similar viral  
338 recovery, but each have advantages and disadvantages (Ahmed et al., 2020; Crocetti et al., 2021). An  
339 advantage of overnight PEG concentration is that the only laboratory equipment required is a centrifuge  
340 capable of reaching 10,000 *g*, which is a standard piece of equipment in most molecular laboratories,  
341 unlike ultracentrifugation which can require speeds of upwards of tenfold greater. Further,  
342 ultracentrifuge filters are expensive and their supply chain was erratic at the start of the pandemic. Due  
343 to the high speeds required and expensive equipment, it is also more difficult to process larger volumes  
344 (> 50 mL) using ultracentrifugation (Lu et al., 2020). However, the ultracentrifugation method is less  
345 time consuming, and concentration can be achieved in 3 hours, rather than the 24-hour process required  
346 for overnight PEG precipitation, which can be an obstacle for the rapid turnaround time required to  
347 make wastewater monitoring useful. In addition to this, PEG precipitation can also suffer from co-  
348 concentration of inhibitors, which can impact on SARS-CoV-2 detection (Scott et al., 2023), which is  
349 not such an issue for ultracentrifugation (Warish Ahmed et al., 2020). Recently, however, due to the  
350 decrease in the number of COVID-19 cases in São Paulo, the ultracentrifugation method was replaced  
351 by the electronegative membrane filtration method (W. Ahmed et al., 2015) which allows the  
352 concentration of high volumes of wastewater (100-200 mL) when compared with ultracentrifugation

353 method (40-50 mL). Importantly, methods used in São Paulo state and Wales can concentrate viruses  
354 from both solid and liquid fractions of the wastewater, which can be important for maximising SARS-  
355 CoV-2 recovery (Ahmed et al., 2020; Kaya et al., 2022). Both methods also have a low cost per sample  
356 once the laboratory equipment has been acquired, which makes the scaling of operations more  
357 economic, vital for regular and widespread monitoring.

### 358 **3.2.2 Extraction**

359 Following concentration, viral RNA is extracted. In São Paulo state, a spin column extraction kit is  
360 used, whereas in Wales magnetic silica beads are used. Much like the differing concentration steps,  
361 both methods offer similar viral recovery of SARS-CoV-2, but each have advantages and  
362 disadvantages. There is evidence to suggest that SARS-CoV-2 recovery is marginally lower using  
363 magnetic silica beads when compared to spin columns, however, magnetic silica beads have also been  
364 shown to provide greater sensitivity and the process can be automated (Pérez-Cataluña et al., 2021).  
365 Another important factor in selecting which extraction methodology to adopt includes availability of  
366 equipment, and because the production of silica magnetic beads is relatively easy, it means that they  
367 are more readily available, unlike specialised commercial plasticware involved in the spin column  
368 extractions (Klein et al., 2020). Plasticware supply chain issues, like those seen during pandemics,  
369 brought on by factors such as high demand, mean that reliance on specialised commercial plasticware  
370 can hinder the progress of smaller wastewater programmes that are unable to stockpile. Finally,  
371 protocols involving magnetic silica beads can be easily automated using liquid handling robots, unlike  
372 those involving spin columns, due to the requirement of centrifugation. The latter allows the  
373 simultaneous processing and preparation of 96 samples within a few hours.

### 374 **3.2.3 SARS-CoV-2 measurement**

375 The focus for both São Paulo state and Wales has been SARS-CoV-2 prevalence in wastewater, which  
376 is measured using reverse transcription quantitative real-time PCR (RT-qPCR), involving the  
377 amplification of target regions from the fragmented SARS-CoV-2 genome found in wastewater (Farkas  
378 et al., 2021). The first step in method development is choosing a region of the SARS-CoV-2 RNA

379 genome to target. In São Paulo, both the N1 and N2 regions are amplified, whereas in Wales the N1  
380 gene is used. The multiple marker approach allows for multiple regions of the fragmented SARS-CoV-  
381 2 RNA genome to be characterised. To assess the presence of PCR inhibitors, in São Paulo, the N1  
382 assay is performed in a multiplex reaction with a synthetic oligonucleotide RNA positive control. In  
383 Wales, a pseudomonas virus phi6 positive control is measured. The final element of the RT-qPCR is  
384 the addition of standards which are used to produce standard quantification curves, which, in São Paulo,  
385 consist of serial dilutions of a SARS-CoV-2 plasmid control. In Wales, however, a synthetic positive-  
386 strand SARS-CoV-2 RNA is used.

387 In addition to monitoring persistence, molecular methods are also used to monitor SARS-CoV-2  
388 variants present in wastewater. In Wales, sequencing is used to detected variants using the EasySeq™  
389 RC-PCR SARS-CoV-2 Whole Genome Sequencing kit on the Illumina NextSeq platform, which is  
390 conducted at all WWTPs once a week. In the São Paulo state, no regular sequencing is conducted for  
391 variant detection, although some samples have been sent to project partners in the Karolinska Institute,  
392 Sweden.

### 393 **3.3 Data analysis**

#### 394 **3.3.1 Data processing and normalisation**

395 In São Paulo, data streams include field measurements (temperature, pH and rainfall), wastewater flow  
396 (instantaneous and daily average) provided by Sanitation Companies, and laboratory determinations for  
397 SARS-CoV-2, total suspended solids, CrAssphage and ammoniacal nitrogen. SARS-CoV-2 data are  
398 originally measured in gene copies (gc) per litre but for WWTPs and other sampling sites where  
399 wastewater flow is available, this measure is converted to daily viral load of gc per day. The data are  
400 normalised by population, considering the population of the sewer catchment. Studies are being  
401 conducted to assess the need for data normalisation for sewage dilution, based on CrAssphage and  
402 ammoniacal nitrogen. CETESB laboratory data have shown that total suspended solids (TSS)  
403 concentration interferes with the recovery of enveloped viruses; samples with higher TSS values had

404 lower bovine coronavirus recovery rates (Barbosa et al., 2022). Recovery rates are not used to correct  
405 the concentration of SARS-CoV-2 detected in wastewaters.

406 For the programme in Wales, there are multiple data streams, all of which are used to produce a  
407 normalised signal. The first type of data stream received by the programme is that produced by the  
408 laboratories, which includes SARS-CoV-2 and chemical markers (ammoniacal-N, electrical  
409 conductivity, orthophosphate, and turbidity). The second type of data stream is flow data, which is  
410 provided weekly by Dŵr Cymru Welsh Water, and is recorded in minute intervals. The final data stream  
411 is static and is an estimate of population size within the WWTP catchment. Due to issues with the  
412 accuracy and reliability of flow measurements taken at the WWTPs, chemical markers are measured  
413 and are used to produce an estimated flow, which in combination with the population estimates are used  
414 to calculate viral load per capita. Using this methodology is important as often flow is lost, or capped,  
415 before it is measured at the WWTP through outlets in the system such as diversion channels, storm  
416 tanks and CSOs, but using the dilution of chemical markers, total flow can still be estimated despite  
417 capping. The details flow estimation and SARS-CoV-2 normalisation can be found in Wilde et al.  
418 (2022). 10 day rolling averages are used to smooth the signal over time. Unlike São Paulo, there is no  
419 outlier removal or missing data imputation.

### 420 **3.3.2 Reporting**

421 In Wales, reports are sent weekly to Welsh Government, with both national and regional overviews of  
422 SARS-CoV-2 levels in the wastewater, and were later made publicly available (Welsh Government,  
423 2022), although reports were no longer made public after the budget cuts in September 2023. Regions  
424 of Wales were split into the management units of the main water utility company, Dŵr Cymru Welsh  
425 Water, with the country being divided into 14 regions, but this later changed to four larger geographic  
426 regions based on health boards, due to the cut in the number of sites sampled after budget cuts. Data is  
427 primarily visualised and communicated in line graphs of rolling mean SARS-CoV-2 gc/day per 100  
428 people over time, a national heatmap and before budget cuts, four bullet points of descriptive text. The  
429 text provided nationally, and for each region, contains the following information: four-week trend, trend

430 compared to the previous week, indicators triggered and sampling issues or inconsistencies. One of the  
431 most informative aspects of the text are the indicators, of which there are three, which indicate high  
432 signal level (viral loads exceed half of the highest weekly average recorded in the previous 6 months),  
433 rapid increase (weekly average of the viral load has increased by at least 100% since the previous week)  
434 and increasing signal level (weekly average of the viral load has increased since the previous week for  
435 at least 3 weeks in a row).

436 In São Paulo, reports with N1 and N2 concentrations (gc/L) for all sites monitored are sent fortnightly  
437 to State Epidemiological (CVE) and Sanitary Surveillance (CVS) Centers (Health Secretary) and São  
438 Paulo Municipal Health Surveillance Coordination (COVISA) with a graphic showing the temporal  
439 evolution for each site. This information is also available at CETESB website where it is possible to  
440 follow up the spatial and temporal variation of SARS-CoV-2 concentration at WWTPs, vulnerable areas  
441 and surface water (<https://cetesb.sp.gov.br/sars-cov-2/>). Other than the evolution of time trends, no  
442 indicators are routinely established in the technical report of São Paulo state, but such a tool can easily  
443 be developed and incorporated into the website.

#### 444 **3.4 Funding and cost**

445 Funded through CETESB, the São Paulo state programme received approximately US\$292,000  
446 (R\$1,435,150) over a three-year period, equating to US\$8111 a month. 45% of the spend was on human  
447 resources and 55% was on supplies and materials. At its peak, the Welsh programme received  
448 US\$5,442,000 (£4,270,000) of funding from Welsh Government for a 12-month period, equating to  
449 US\$453500 a month, which was then reduced to US\$791,400 (£621,000) between September to March  
450 2023, equating to US\$113,057 a month. The longest continuous contract awarded at any one time was  
451 12 months.

452 Due to differences in the price of goods and services between São Paulo state and Wales, an exact  
453 comparison of cost is difficult, however, Purchasing Power Parities (PPP), offer a methodology to  
454 compare costs through a "basket of goods" approach (table 2). We accept that there are variations in  
455 standards of living and values of indicators such as gross domestic product per capita within nation

456 states. However, Brazil and UK national figures, in terms of PPP, provide an estimated comparison of  
 457 purchasing power between São Paulo state and Wales. These comparisons highlight the disparity in  
 458 funding between the two programmes and begins to explain differences in frequency and geographic  
 459 spread of sampling, as well as the extent of monitoring conducted on other biological markers other  
 460 than SARS-CoV-2.

461 **Table 2. Comparison of wastewater programmes costs in São Paulo state and Wales. Included**  
 462 **are the actual monthly costs of the programmes in their home countries and local currencies, as**  
 463 **well as an estimated comparable cost, using Purchasing Power Parity from 2022 (OECD, 2024).**  
 464 **Figures are given to the nearest thousand.**

<b>Wastewater programme</b>	<b>Actual monthly cost in source country and currency</b>	<b>Purchasing power parities, National currency units/US dollar</b>	<b>Estimated monthly equivocal cost in US dollars</b>
<b>São Paulo state, Brazil</b>	R\$40,000	2.583	US\$15,000
<b>Wales, UK (scaled down)</b>	£80,000	0.651	US\$136,000
<b>Wales, UK (peak)</b>	£356,00	0.651	US\$547,000

465

## 466 **4. Common challenges**

### 467 **4.1 Design**

468 By comparing the Wales and São Paulo, the main challenges in designing an efficient wastewater  
 469 surveillance program include: lack of existing experience and infrastructure, the level of collaboration  
 470 between stakeholders, the extent to which the sample can provide a representative measurement of  
 471 SARS-CoV-2 levels in wastewater as well as the extent to which these levels can be related to known

472 populations and infectivity. The design of both programmes has involved the participation of a range  
473 of stakeholders including researchers with an understanding of epidemiology and water systems, water  
474 utilities that can provide information and access to sewerage systems and data, governmental  
475 institutions that can provide funding and public health data to support the programme. In São Paulo,  
476 many of these collaborations were already in place because of existing poliovirus and cholera  
477 programmes. However, given the historical nature of these previous programmes, collaborations and  
478 initiatives for environmental monitoring have developed at municipal, state or regional levels. The  
479 SARS-CoV-2 programme has reflected this legacy and a National Wastewater Monitoring Plan for  
480 SARS-CoV-2 has not yet been developed. On the other hand, the urgent need for national decision-  
481 making tools as well as the lack of existing wastewater surveillance systems has required investment at  
482 a scale that could only be directly led and resourced by Welsh Government. Lack of experience in large-  
483 scale routine wastewater surveillance in Wales, as well as the United Kingdom more broadly, has  
484 brought the additional challenge of rapidly developing operational collaborations between environment  
485 agencies, academia, public health and water utilities. While academic institutions often develop  
486 collaborative links, these often operate around research and innovation departments and over short  
487 timescales, thus they rarely involve long term operational demands required for wastewater  
488 surveillance.

489 Factors affecting the quality or quantity of genetic material in the sample have also been a challenge to  
490 design, with one of the main factors being dilution. Extreme weather events such as floods and storms  
491 are common in tropical regions, including Brazil, and high rainfall associated with storm events are also  
492 common in Wales, particularly when excess water enters the wastewater system, they cause dilution of  
493 the SARS-CoV-2 signal. Indeed, this problem is only set to get worse, as climate change has been  
494 identified as one of the main challenges faced by urban wastewater systems globally, with increased  
495 frequency of high-intensity storm events (Hughes et al., 2021; Langeveld et al., 2013). Ideally, sewage  
496 would be collected separately from surface-runoff, reducing the impact of dilution. Even where sewage  
497 is separated from wastewater, such as São Paulo, surface-runoff can still enter the sewage network  
498 through misconnections. One way of accounting for the changes in SARS-CoV-2 signal induced by



499 dilution is by normalising by wastewater flow. However, access to reliable flow data is often not  
500 possible, due to logistical, technical, or financial constraints. Even if a flow meter is present at the  
501 WWTP, flow data can become more unreliable during high flow events due to issues caused by debris,  
502 and because wastewater can be diverted before the flow meter into storage tanks or overflow into rivers  
503 or the sea. Not only this, but where the flow meter is placed in the WWTP can influence the flow  
504 measurements, making comparability between plants difficult. When flow is not available, or when  
505 flow data is unreliable, there are methods that can account for dilution based on chemical or biological  
506 constituents of the wastewater (Wilde et al., 2022). However, these methods require further laboratory  
507 capacity, at least some flow data (e.g. historic flow) and may not be as effective in near-source  
508 applications. The measurement of flow within the sewer network (e.g. at manholes) is particularly  
509 problematic, but not impossible, as within Welsh programme flow gauges have recently been installed  
510 for near-source monitoring of wastewater at hospital sites.

511 If dilution can be adequately accounted for, there still remains the issue of dilution and its impact on  
512 viral detection rates (Aguiar-Oliveira et al., 2020), with the possibility that significant dilution events  
513 could reduce the SARS-CoV-2 signal below detection limits, rendering dilution normalisation  
514 techniques redundant in these scenarios. In addition to dilution, decay is another challenge when  
515 designing a wastewater monitoring programme. Sewage infrastructure and sampling design can impact  
516 the length of time it takes for particles to get to a sampling point, and then to reach the laboratory. Given  
517 the nature of SARS-CoV-2 and other pathogens, transit times can affect the decay level of the material  
518 and thus the quality of the sample (Burnet et al., 2023). Other wastewater contributions to the sewage  
519 system can also affect the quality of the sample. For example, industrial contributions can, in addition  
520 to diluting sewage, affect the viability of pathogens in the environment (Bayati et al., 2022) as well as  
521 cause PCR inhibition (Scott et al., 2023).

522 Finally, variation in the population numbers being serviced by a wastewater system is a challenge to  
523 wastewater monitoring and relating a wastewater sample to its shedding population. Populations  
524 serviced by a wastewater treatment plant can be estimated by looking at population census. These can  
525 be at best annual, but always represent a population residing in the area at a static point in time.

526 Movement of individuals during the day to commute, but also for tourism, can significantly affect viral  
527 loads. This is true for example for tourism hotspots like the Welsh coastal towns, or the main port of São  
528 Paulo. Populations not serviced by a wastewater treatment plant are more difficult to estimate.

#### 529 **4.2 Laboratory methods**

530 SARS-CoV-2 concentrations measured in wastewater can be strongly influenced by sampling method,  
531 sample preservation, storage time, concentration method, RNA extraction method, RT-PCR assay  
532 selection and overall performance of each step in the molecular pipeline (Beattie et al., 2022; McClary-  
533 Gutierrez et al., 2021). Yet decision making processes in response to operational constraints have led  
534 to a great diversity of molecular pipelines, with no standardized protocols for the determination of  
535 SARS-CoV-2 in wastewater.

536 Previous poliovirus monitoring has influenced the processing of wastewater samples for SARS-CoV-2  
537 detection in São Paulo. The government agency, CETESB, had existing environmental virology  
538 laboratory infrastructure and well-established collaborative networks from environmental poliovirus  
539 surveillance which began in 1980. In Wales, however, academic environmental virology laboratories  
540 were, and continue to be, the foundation of the national programme. SARS-CoV-2 is an enveloped  
541 virus, unlike poliovirus and other enterovirus, which had been previously monitored by government  
542 laboratories in São Paulo state. Therefore, although there was valuable expertise and equipment  
543 available, new methodologies for SARS-CoV-2 had to be validated, making them more specific, more  
544 sensitive and reducing degradation of the non-enveloped virus, all while working within the parameters  
545 of existing laboratory infrastructure. The academic, and fundamentally experimental, beginnings of the  
546 SARS-CoV-2 monitoring in Wales meant that it was not constrained by the infrastructure of previous  
547 monitoring programmes, which offers benefits, but also came with its own constraints. Laboratory  
548 infrastructure had to be created during a time where equipment was scarce, and molecular methods and  
549 analytical pipelines are constantly evolving, which has consequences for data consistency and long-  
550 term comparisons. The consistency of long-term comparisons is likely to change over time, even  
551 without methodological approaches, as new strains arise, and populations become more widely

552 vaccinated. However, flexibility in the molecular approaches used in Wales has also contributed to a  
553 lack of standardised molecular protocols used in the two laboratories in the Welsh programme, which  
554 at times made comparability between samples difficult. Ultimately this led to the primary processing of  
555 samples in one laboratory with subsequent molecular analysis carried out in two laboratories.

556 In addition to variations in the molecular pipelines, SARS-CoV-2 can be influenced by the properties  
557 of wastewater itself, including matrix composition, physicochemical characteristics of wastewater and  
558 viral form (Kantor et al., 2021; LaTurner et al., 2021; Li et al., 2021). The combined sewage systems  
559 found in Wales has the potential for producing highly complex wastewater due to varied inputs,  
560 including industry and surface runoff. Each of the inputs could contain a different cocktail of PCR  
561 inhibitors, such as multi-ringed polysaccharides (e.g. humic and fulvic acids), salts, fats, proteins,  
562 surfactants, metal ions (e.g. iron and aluminium) and RNases (Warish Ahmed, Simpson, et al., 2022),  
563 all of which can cause false reduced SARS-CoV-2 signals when using RT-qPCR. Not only this, but  
564 inhibitors can vary in both time and space in ways that are often hard to predict, such as human  
565 behaviour (Pons et al., 2020). The properties of the wastewater may also affect downstream sequencing  
566 with some sites routinely failing despite having strong RT-qPCR signals. The same issue is also seen  
567 in the São Paulo state programme, however, the greater separation of industrial waste and surface runoff  
568 means that, in theory, the wastewater matrix should be more consistent in composition.

569 Finally, trying to detect small concentrations of SARS-CoV-2, and establishing an adequate limit of  
570 detection/quantification can also be a challenge, especially when wastewater samples are highly dilute,  
571 because of high rainfall events, or if SARS-CoV-2 cases are low in the population. Increasing sensitivity  
572 in RT-qPCR is one of the primary drivers for selection of molecular methods (Warish Ahmed, Bivins,  
573 et al., 2022), which is also the case in São Paulo state and Wales.

#### 574 **4.3 Data analysis**

575 To assess the validity of wastewater surveillance, one of the most straightforward assessments that can  
576 be carried out is a correlation between COVID-19 cases and SARS-CoV-2 signal in the wastewater.  
577 This does come with two assumptions, the first being that case data is complete, reliable and

578 representative. This was not the case at the start of the pandemic when testing was often sporadic and  
579 targeted to outbreaks causing a mismatch between wastewater and clinical surveillance data (Hillary et  
580 al., 2021). The São Paulo State Health Department provides daily data for COVID-19 cases per  
581 municipality which is compared with SARS-CoV-2 levels, however, due to the lack of intensive testing  
582 in Brazil, most reported cases correspond to symptomatic patients and health seeking behaviours,  
583 invalidating any comparison with wastewater SARS-CoV-2 levels. In Wales, although during the height  
584 of the pandemic there was intensive routine testing, however, national policy has now moved away  
585 from providing free testing services, to a reliance on self-reporting and finally no need to report a  
586 positive case. Consequently, comparison of this data with wastewater SARS-CoV-2 levels became more  
587 and more futile.

588 Subsequently, an estimated national percentage of the population with COVID-19, which was modelled  
589 by the Office for National Statistics as part of their COVID-19 Infection Survey (Office for National  
590 Statistics, 2022), was used for comparisons with the SARS-CoV-2 wastewater signal. The COVID-19  
591 Infection Survey estimates had their limitations, however, such as the decreasing number of SARS-  
592 CoV-2 tests they were based on, as well as the lack of regional data. However, even this programme  
593 was terminated in March 2023, and so wastewater provides the only reliable way to estimate national  
594 levels of SARS-CoV-2 infections.

595 The second assumption of comparing SARS-CoV-2 levels in wastewater and case data is whether the  
596 two have a meaningful relationship. The levels of SARS-CoV-2 in wastewater corresponds to the  
597 number of people who are infected with SARS-CoV-2 and their shedding rates. A host of factors can  
598 impact shedding, such as age (Bertels et al., 2022; Jones et al., 2020; Omori et al., 2021; Prasek et al.,  
599 2022), and shedding can last a week after detectable respiratory SARS-CoV-2 (Zhang et al., 2021).

600 Another challenge when utilising levels of SARS-CoV-2 in wastewater is understanding the source of  
601 variation and trying to limit that variation so that a representative signal can be achieved. This starts  
602 with the sample design, for example, grab samples will be inherently less representative of the  
603 wastewater over a 24-hour period than composite samplers, yet in the São Paulo state and Welsh

604 programme, results from both grab and composite samples are compared together. We have also already  
605 outlined how the programmes in both São Paulo state and Wales try and limit variation through  
606 improved laboratory methods, however, data collected in the laboratory on precipitation volumes,  
607 extraction efficiency and qPCR efficiency are not used in downstream data analysis, and so that  
608 explanatory variable for variation is therefore lost.

609 Neither São Paulo state nor the Welsh programme routinely implement an outlier removal method.  
610 Removal of large values in signal could be erasing a genuine rapid increase in SARS-CoV-2 in a  
611 community, which would be important for policy makers. Finding and implementing a robust statistical  
612 approach for understanding outliers in this context has not yet been achieved. If an exceedingly high  
613 value is detected in the Welsh programme, data sources contributing to that signal are investigated (e.g.,  
614 physiochemistry, flow data, qPCR result), and if there is doubt on the quality of that sample, it is  
615 withdrawn and re-analysed. Not having a robust outlier removal method risks the inclusion of  
616 anomalous data.

617 One of the greatest sources of variation on levels of SARS-CoV-2 in wastewater is from the sewage  
618 infrastructure, primarily through dilution and population sizes. The Welsh programme uses a robust  
619 methodology to account for dilution and population size, whilst also overcoming the challenges  
620 imposed by sewage infrastructure (mainly capped flow measurements through CSOs). Yet, despite this,  
621 the Welsh programme only have access to static population estimates, when in some cases, large  
622 proportions of the population are not static, due to phenomenon such as commuting and tourism, which  
623 will lead to unaccounted variation in levels of SARS-CoV-2 in wastewater. In São Paulo, even getting  
624 static population estimates for a wastewater catchment can pose a challenge. The Demographic Census  
625 is the best database, and the information is collected by households, but it is held every 10 years and  
626 the last one, in 2020, has not yet been finalised. The SNIS is a national database on Water and Sewage  
627 services, Urban Solid Waste Management and Drainage and Urban Stormwater Management for all  
628 Brazilian municipalities, and it too has detailed annual information, but the data is self-reported by the  
629 sanitation companies and is not audited or validated. It therefore contains inconsistencies introduced by  
630 errors in filling out official forms, difficulties verifying data by local teams, or even intentional

631 manipulation of information by service providers. For example, population resident in regularized areas,  
632 or areas under their responsibility, as defined in the service concession contract, and not the total  
633 population of the municipality. Therefore, populations that reside in watershed protection areas are  
634 intentionally made invisible. Indeed, households used by the tourist population or commercial activities,  
635 but which maintain the residential register, also pay lower costs and contribute to overestimating the  
636 proportion of the population with access to the water network and sewage.

637 Further progress also needs to be made on characterising the impacts different inputs have on signals  
638 detected at a WWTP, for example, the age of a population, the number of hospitals and population  
639 density. Similarly, better linking results obtained from wastewater with healthcare interventions could  
640 provide a valuable tool in validating results from wastewater programmes, but also validating the impact  
641 of different types of intervention (e.g. transport restrictions, targeted vaccination). Yet these real-world  
642 examples are lacking in both the São Paulo state and Welsh programmes. One barrier in progressing  
643 crossover between wastewater surveillance and public health is that often the areas monitored for  
644 wastewater overlap multiple health boards or administrative boundaries, in addition to the difficulties  
645 of working with public health stakeholders who can have different aims and objectives.

646 On a practical level, working with multiple stakeholders, not only in public health, but also within a  
647 wastewater programme (e.g., water utilities, government, laboratories) can also pose its own challenges.  
648 For example, in Wales, due to the constraints of data sharing between institutions imposed by firewalls,  
649 prior to the implementation of a new data system in 2023, two separate databases existed in the North  
650 and South Wales laboratories, both of which had different data formats and required frequent manual  
651 updates. In addition, accessing regional data on the number of clinical COVID-19 infections and SARS-  
652 CoV-2 variants from individual patients proved almost impossible to obtain due to ethical and data  
653 compliance issues. Fundamental databasing and data sharing issues such as these can cost time and  
654 impact cross collaboration.

655 In the State of São Paulo, data sharing must comply with the Brazilian General Data Protection  
656 Regulation (Law 13.709/2019), however, although data on mortality and hospital admission is available

657 on an open access platform, there is a delay of about two months for data on hospitalizations and two  
658 years for mortality, with the smallest scale of data being entire municipalities, except for the  
659 municipality of São Paulo state where the data are available on the census tract scale. Additional  
660 difficulties with health data involve underreporting; lack of an integrated outpatient database in basic  
661 health units; the use of different administrative divisions for the areas of health, sanitation, environment,  
662 among others. Besides, these databases usually have high dimensionality, missing data and can suffer  
663 from user inputting errors, thus requiring specialist teams with computing, data handling and statistical  
664 skills.

#### 665 **4.4 Novel insights**

666 While other reviews have examined challenges of wastewater surveillance design (e.g. number of  
667 samples taken, geographic spread, degradation and the value of composite samples) (Medema et al.,  
668 2020), laboratory methods (e.g. extraction yields, concentration and purification) (Kumblathan et al.,  
669 2021) and, to a lesser extent, data analysis (e.g. wastewater-clinical comparisons and population  
670 normalisation) (Polo et al., 2020), our cross-continental comparison and use of case studies has allowed  
671 us to identify and go beyond discussing technical challenges. Novel insights into more holistic  
672 challenges are presented, such as the collaboration between stakeholders required to bring an effective  
673 wastewater surveillance system together, how choosing stakeholders can impact long term laboratory  
674 processes and how data sharing issues between stakeholders can seriously limit novel insights and  
675 analyses.

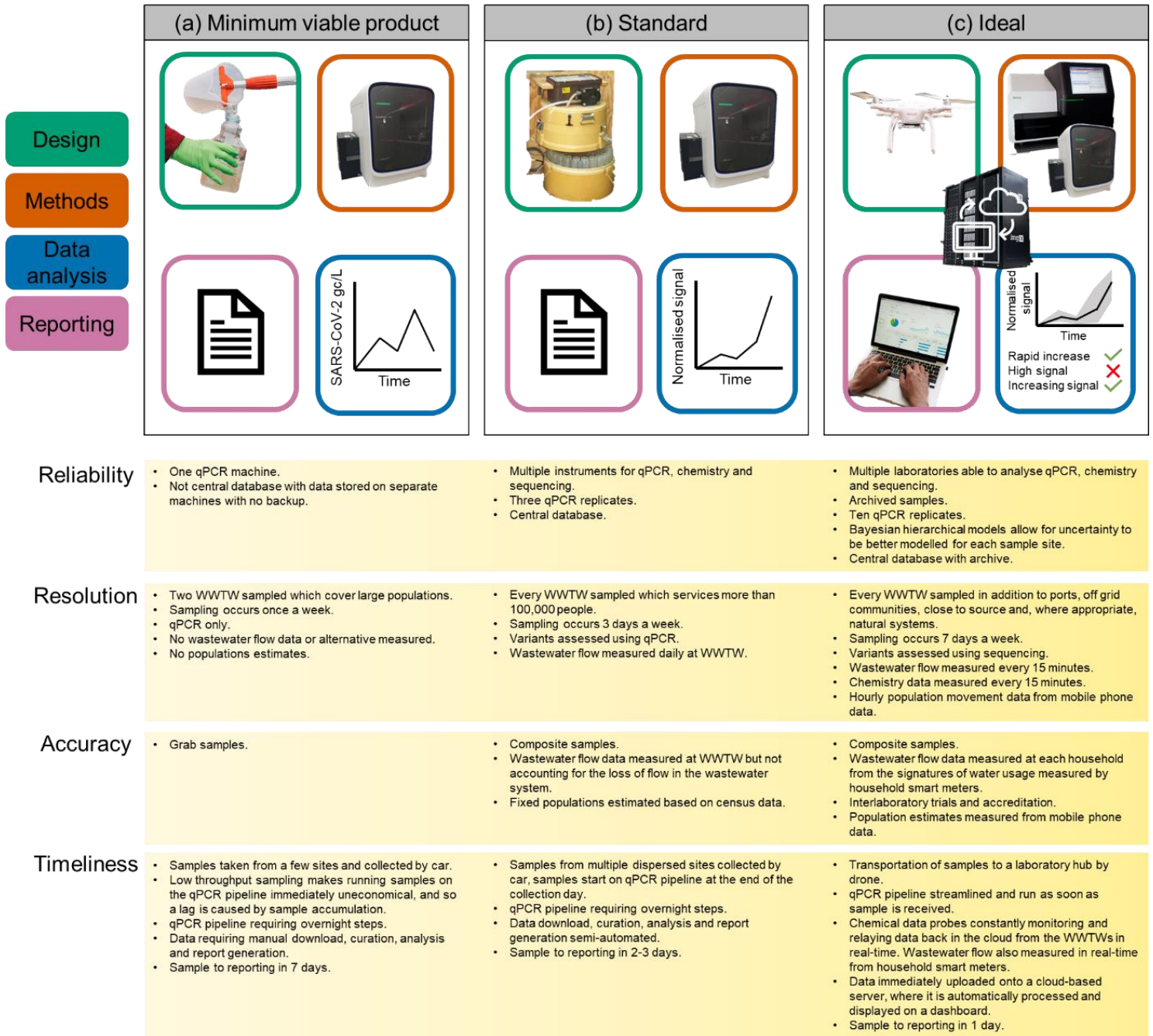
#### 676 **5. The ideal: towards a wastewater surveillance fit for the 21<sup>st</sup> century**

677 There is no doubt that adequate funding and experience of wastewater surveillance provides a solid  
678 advantage in implementing an efficient programme for monitoring emerging diseases. Previous  
679 knowledge, infrastructure and collaborations between stakeholders in São Paulo state have clearly  
680 provided their programme a head-start, and the well-funded Welsh programme is testament to the  
681 insights that can be gained if funding is made available. As the pandemic progresses to an endemic  
682 state, and as potentially new pandemics may emerge, the swift response of the São Paulo state

683 programme to the COVID-19 pandemic is evidence of the value of experience. Given that many  
684 countries, such as Wales, now have experience in wastewater surveillance due to the COVID-19  
685 pandemic, we outline what an ideal wastewater surveillance system may look like and how to progress  
686 from minimal viable products (Fig. 4a) and standard wastewater surveillance (Fig. 4b) to an ideal system  
687 (Fig. 4c). To do so, we draw on the experiences of both the programmes in Wales and in São Paulo,  
688 relevant for any disease with biomarkers in wastewater, not just SARS-CoV-2.

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690

691 **Figure 4 Progress of wastewater surveillance systems, from a (a) minimum viable product, to the**  
 692 **(b) standard and then finally, to the (c) ideal system, with developments in design, laboratory**  
 693 **methods, data analysis and reporting. In addition to this, there are bullet points containing**  
 694 **features of (a) to (c), split into fundamental elements of an effective wastewater monitoring**  
 695 **programme, including reliability, resolution, accuracy and timeliness which filter through to all**  
 696 **elements of the programme, including design, laboratory methods, data analysis and reporting.**

## 697 **5.1. The ideal: design**

698 One of the most fundamental questions in designing a wastewater surveillance programme is what scale  
699 the sampling should be conducted on, both in space and time. Taking multiple samples per week has  
700 been highlighted as a key frequency of temporal sampling (Harris-Lovett et al., 2021) and the Centers  
701 for Disease Control and Prevention (CDC) recommends a minimum of three samples within a trend  
702 period of 15 days (CDC, 2022). The greater the number of samples over a week, the more robust the  
703 trends, and therefore, a frequency of 3-7 sample points a week should be aimed for, all of which would  
704 need to be taken with composite refrigerated samplers over a 24-hour period (Fig. 4b-c). Preferably the  
705 composite sampler should take samples every 10 mins to capture the variability in SARS-CoV-2  
706 concentration. The latter is particularly important when used as an early warning system for pathogen  
707 emergence where the levels of SARS-CoV-2 RNA are expected to be highly temporally variable. This  
708 sampling frequency is not feasible in the real-world for most countries as it incurs high cost and complex  
709 logistics, and often fewer samples can be sufficient in answering the relevant public health questions.  
710 Preventing degradation at this stage is also vital, ensuring that samples are kept at 4°C and transported  
711 to the laboratory within 24 hours of it being taken. It is likely in the future that preservatives may be  
712 added to the samples to prevent the loss of genetic material (S. H. Bell et al., 2023).

713 For spatial sampling, this will depend on the make-up of the sewersheds and the wastewater  
714 infrastructure, but it is beneficial to target large urban populations to that you are able to quickly cover  
715 a large percentage of a region's population. However, to get a representative view of a region, rural or  
716 off-grid populations must not be neglected. Many of the more nuanced decisions based on geographic  
717 sampling may also be politically motivated. Therefore, before designing a surveillance programme, an  
718 initial demographic mapping exercise should take place, whereby the surveillance planners are given  
719 information on aspects of populations served by sewersheds such as age, population density, deprivation  
720 levels, healthcare provision, migration rates (e.g. commuting, national/international tourism). An ideal  
721 system would also include the sampling of communities not connected to sewersheds through the  
722 sampling of open sewers, the natural environment and septic tanks. This type of sampling would take  
723 greater resource and logistics, but it may be important to monitor communities most at risk.

724 Environmental surveillance should also extend to regions where sewage is discharged without sufficient  
725 treatment, to monitor waterborne diseases and other hazardous sewage contaminants that pose a risk to  
726 other water users (e.g. recreation and irrigation). There may also be strategic sampling for, example, an  
727 early warning system, which would prioritise entry points into a region, such as international borders,  
728 airports and seaports, with sampling of the onboard wastewater storage facilities of planes and ferries  
729 (Farkas Id et al., 2023).

730 It is important for wastewater monitoring to be coordinated at national level, implemented at state and  
731 municipal level, and have the support of all stakeholders (sanitation companies, laboratories,  
732 government) to generate reliable data for decision makers, all the while working with standardised  
733 protocols for sample collection, sample processing, data processing and reporting. These initiatives  
734 have already been established in other countries such as South Africa, Turkey and the US (Sutton et al.,  
735 2022; Tlhagale et al., 2022), as well as within Europe (Izquierdo-Lara et al., 2021). In addition, an  
736 International Organization for Standardization standard is currently being developed for SARS-CoV-2  
737 quantification in wastewater. Once established, these wastewater surveillance programmes are not only  
738 important for dealing with COVID-19, but can also be applied to other infectious diseases, as well as  
739 chemical compounds (e.g. disease biomarkers, pharmaceuticals, microplastics, illicit drugs). Once the  
740 sampling regime, infrastructure and stakeholder networks have been put in place, sample processing is  
741 the only module which needs to change to tackle a wealth of other problems. For this to be feasible,  
742 however, there must be guaranteed funding over the medium and long term to maintain this  
743 infrastructure, taking a programme beyond just research. Post pandemic funding is required to stop  
744 networks, equipment and expertise from dissipating before the start of the next event which requires  
745 wastewater surveillance, be that a critical public health issue (e.g. new pandemics, endemic,  
746 antimicrobial resistance (AMR)) or other social issue (e.g. illicit drug taking). A break in funding was  
747 seen in the Welsh programme in July 2023, where Welsh Government cuts to healthcare meant that the  
748 entire programme was abruptly cancelled. The period of cancellation only lasted a month, after which  
749 the need for insights into SARS-CoV-2 levels and variants was so great, there was a reversal to policy  
750 made by Welsh Government. Instead, a minimal viable product was put together, cutting sites from 47

751 to 11 WWTPs and sampling frequency from five to three days a week, along with no hospital sampling,  
752 no dashboard and removal of AMR monitoring. However, the disruption resulted in the loss of four  
753 months of wastewater surveillance, a loss of trust between stakeholders and a loss of expertise.

754 Further extension of the current stakeholder networks would also be beneficial, to better integrate  
755 wastewater surveillance results into public health decision making. In the São Paulo state programme  
756 this would involve cross collaborations between Execution State and Municipal Spheres in  
757 collaboration Sanitation Agencies (sampling and laboratory), Universities and Research Institutions,  
758 Health Epidemiological and Sanitary Offices, Environmental and Health Laboratories. In Wales,  
759 stronger connections could be made with current partners, Public Health Wales and local authorities, to  
760 utilise wastewater data on a more localised basis, while also designing surveys to answer public health  
761 questions faced by different health boards in Wales. Better connection with international networks  
762 would also be beneficial, strengthening those that are already in place, such as with São Paulo state and  
763 Wales, but also incorporating other nations, as has been seen with the Pan-American network for  
764 Environmental Epidemiology (PANACEA) network, whose work spans 15 countries.

## 765 **5.2 The ideal: laboratory methods**

766 The trade-off between the choice of different methodologies is largely between cost of the consumables,  
767 processing time and the methodology which consistently gives the greatest yield. In an ideal system,  
768 the methodology with the highest yield would be preferentially favoured, but for results to be delivered  
769 to policy makers in time for them to be useful, processing time is also very important, as is the  
770 responsible expenditure of laboratory resources from publicly funded projects. Therefore, an ideal  
771 system must be flexible to the constraints in which it is born into and will differ between programmes.  
772 In addition to cost, time and yield, accuracy is also important, with good accuracy being dependent on  
773 sufficient technical replication. An ideal system would have samples processed in a minimum of  
774 triplicate from the concentration phase to qPCR, but preferably more (Fig. 4c). When at the qPCR stage,  
775 it would also be beneficial to use a minimum of three probes for the biological marker you are trying to  
776 detect. For example, with SARS CoV-2, using a combination of probes that target different regions of

777 the SARS CoV-2 genome. Multiple probes build redundancy and has increased confidence in positive  
778 and negative signals in other wastewater surveillance programmes (Huang et al., 2021), in a similar  
779 way to increased replication, with the CDC initially recommending the use of three sets of probes for  
780 clinical testing (Yaniv et al., 2021), with the N1, N2 and E gene probes showing the greatest sensitivity  
781 and correlation with cases (Hong et al., 2021; Huang et al., 2021). This also minimises the potential for  
782 false-negative results arising from gene dropouts (Isabel et al., 2022; Wollschläger et al., 2021).

783 Many of the underlying techniques for RNA extraction and concentration have not changed for decades,  
784 however, staying at the cutting edge of technological advances regarding measurement of RNA and  
785 other nucleic acids is important. For example, the emergence of digital qPCR can allow for improved  
786 limits of detection and quantification, which could be important in some contexts, such as early  
787 detection, or in highly dilute samples (Ahmed, Bivins, et al., 2022; Tiwari et al., 2022). The  
788 characterisation of variants by high throughput sequencing is another area in which emerging  
789 sequencing technologies could improve on elements of accuracy, sample processing time and cost.  
790 Implementation of nanopore technologies in replacement of sequencing by synthesis technologies, such  
791 as Illumina platforms, could eliminate PCR steps which currently makes up part of the EasySeq™ RC-  
792 PCR SARS CoV-2 used in Wales, while also reducing sample processing time and cost; although it  
793 must also be highlighted that the sequencing accuracy of these technologies is still behind those of the  
794 Illumina platform (Barbé et al., 2022; Rios et al., 2021). This highlights another component, additional  
795 to cost, processing time and accuracy, and that is the uncertainty of running new technologies, which  
796 can pose a significant risk to the stakeholder trust placed in the data being produced by a surveillance  
797 programme, especially when it is being used for policy making decisions.

798 Going forward, design of wastewater surveillance programmes should also incorporate other biological  
799 markers for qPCR outside of those relevant to pandemics, as well as regularly sequencing samples (Fig.  
800 4c). Biological markers of multiple pathogens are being regularly monitored in São Paulo state and  
801 Wales, such as poliovirus and *Vibrio cholerae* in São Paulo state and as of September 2022, influenza,  
802 enterovirus, respiratory syncytial virus, poliovirus, and norovirus in Wales. However, a truly cutting-  
803 edge wastewater surveillance programme would go beyond monitoring bacterial and viral pathogens

804 and would use the established infrastructure to investigate other pressing societal issues which lack  
805 data. In the case of biological markers, this could include monitoring cancers such as prostate cancer,  
806 hormones or, like in the case of the programme in Wales, monitoring of AMR. One tool which is being  
807 harnessed by the Welsh programme is metagenomics, which can produce large databases that are not  
808 marker specific (Adriaenssens et al., 2021). In Wales, the datasets produced for monitoring of AMR  
809 can also be mined for a plethora of prokaryotic and eukaryotic pathogens. Additionally, surveillance  
810 programmes do not need to focus solely on biological material and having multiple laboratory process  
811 which can process an array of chemical markers (e.g. illicit drugs, pharmaceuticals, emerging  
812 contaminants or microplastics) would be value added for relatively little cost, benefiting from the  
813 sampling design and analysis already in place. An example of this value added can be seen in the Welsh  
814 programme, where after cutting the programme's budget and thus the number of sites (from 47 to 11),  
815 the cost per WWTP sampled increased from £7,571 per site per month to £8,065 per site per month,  
816 while also losing two days of sampling, a week AMR monitoring, monitoring at hospital sites and a  
817 data dashboard.

818 Irrespective of the type of methodology, making them replicable is key to an ideal system. If the  
819 methodologies are run in separate laboratories, detailed standard operating procedures are vital, as even  
820 small deviations in the methodology can render results incomparable. Variables such as the supplier of  
821 laboratory consumables and equipment, and even down to batch number of the consumables used can  
822 significantly impact replication, sometimes due to errors by the manufacturer. Ideally, consumables  
823 from the same batch would be used between laboratories to alleviate this problem. Another important  
824 step for replicability is the use of ring trials, so that samples and standards are consistently being  
825 compared between laboratories to ensure that the methodologies are still aligned. One way in which  
826 replication can be assured is to run all samples through the same laboratory, however, this then  
827 introduces another problem: having a single source of truth which is not challenged. Not only this, but  
828 logistical problems with one laboratory could cripple a national programme, and therefore, a hybrid  
829 approach would be beneficial.

830 Finally, an ideal system would maximise the use of automation and robotics. The greatest potential  
831 application of robotics in the processing of wastewater samples is in the liquid handling stages. Not  
832 only do robotics allow the high throughput processing of samples, and therefore broader coverage in  
833 time and space, but they also eliminate variation introduced by different laboratory members and their  
834 approach to processing samples (Hayase et al., 2023). Although, it must be stated that making sure that  
835 the robotics between laboratories is identical in terms of manufacturer, programming and mechanics is  
836 key, or otherwise these components have the potential of introducing the same, if not more, variation  
837 between samples. In addition, the future may also see automated in-sewer detection systems (Ou et al.,  
838 2023). One issue of potential major concern for future pandemics is the nature of the organism being  
839 tracked. In the case of SARS-CoV-2, no conclusive evidence was presented to suggest that it was  
840 infectious when present in wastewater (Jones et al., 2020). This allowed samples to be collected and  
841 processed under Biological Safety Containment Level 2 (BSL2) conditions. However, subsequent  
842 pandemics might be associated with organisms that remain highly infectious in faeces and urine and  
843 thus require sampling handling under BSL3 containment. Based on the COVID-19 pandemic, in this  
844 situation it is likely that the BSL3 facilities will be prioritised for clinical surveillance rather than  
845 wastewater surveillance, due to the low numbers of BSL3 laboratories outside healthcare settings. A  
846 plan of action is therefore needed for this scenario, akin to those that exist for other BSL3 organisms  
847 (e.g. Ebola (Jelden et al., 2016)).

### 848 **5.3 The ideal: data analysis**

849 Much of the data analysis required in wastewater surveillance is accounting for variation introduced by  
850 the source (e.g. shedding rates of SARS CoV-2 between variants and age of hosts), sewage network  
851 (e.g. dilution, degradation, transport times) and laboratory processes (e.g. qPCR efficiencies, limits of  
852 detection). In an ideal system, sources of variation should be reduced or explained and incorporated  
853 into the analyses. The goal of design and methods of an ideal surveillance system is therefore accuracy  
854 (Fig. 4), which will reduce variation in the measurements being taken, as well as accounting for as much  
855 of this variation as possible. For example, population and dilution normalisation of a signal in  
856 wastewater would take place in an ideal system but would be further improved by data streams that

857 were more accurate and representative. In the case of population normalisation, instead of static figures  
858 like those from a census, dynamic population estimates from anonymized call detail records (mobile  
859 phones) would be used (Lai et al., 2019), which could account for daily fluctuations in populations  
860 caused by phenomenon such as commuting and tourism. Dilution normalisation would also be improved  
861 with the use of chemical markers in addition to the wastewater flow measurements taken at wastewater  
862 treatment works (Wilde et al., 2022), as is currently implemented in the Welsh programme, which would  
863 rely on the measurement of a suite of chemical markers (e.g. ammoniacal-N, phosphate, electrical  
864 conductivity, caffeine).

865 Another way of better representing variation in the wastewater signal is by providing error margins.  
866 Bayesian hierarchical modelling can be a particularly powerful tool in doing so (Medema et al., 2020),  
867 as not only could you incorporate existing knowns (priors) such as limits of qPCR quantification, qPCR  
868 efficiency and COVID-19 cases, the hierarchical element would allow you to produce separate estimates  
869 of wastewater signal, as well as a margin of error, per sample site.

870 Invariably, even in an ideal system, due to the nature of wastewater being a complex matrix, and the  
871 molecular pipelines for the quantification of viral particles being so complex, there will be outliers  
872 which skew the overall trend. Therefore, another element of an ideal data workflow would include an  
873 automated outlier detection system for recognising extreme values in a timeseries. Proposed outlier  
874 removal methods included Generalized Extreme Studentized Deviate (GESD) test, which have been  
875 used in other viral epidemiology contexts (Wiemken et al., 2020). The difficulty with applying such  
876 methods is that outbreaks of a disease such as COVID-19, and thus the viral particles in the wastewater,  
877 can escalate rapidly, meaning that it can be difficult to separate outliers from trends if you are sampling  
878 on a weekly basis. This reaffirms the need for fine scale temporal sampling, which allows for better  
879 discrimination between a real rapid increase of viral particles in the wastewater and outliers introduced  
880 through the sampling and laboratory pipelines.

881 A crucial element to data analysis of wastewater surveillance data is being able to communicate trends  
882 in the data to policy makers and public health bodies. Not only does this mean an ideal system must



883 have excellent data visualisation, it also impacts the way in which the trends are delivered to the end  
884 user. Many regions around the world are moving away from reporting wastewater surveillance results  
885 in a static report, and instead moving to dashboards, whereby live results can be presented in an  
886 interactive format that is suited to different audiences, which is part of a wider popularity of displaying  
887 data in dashboards. Not only does this allow for a more engaging way to showcase and distribute data  
888 insights, it also reduces the timeline between data generation and availability to end users, allowing for  
889 quicker dissemination and data-driven action.

890 Finally, and most fundamentally, for data analysis to be undertaken, no matter how simple, the data  
891 must be readily available, which makes effective data sharing and databasing another crucial element  
892 of an ideal system. The more complex the analysis, the more data streams are required. For example,  
893 multiple molecular laboratories may be providing the SARS-CoV-2, each of which should be able to  
894 freely upload results to a central database with ease without the hinderance of firewalls. In the ideal  
895 system, those measurements would be easily combined with data sources from the chemistry laboratory  
896 providing physiochemical measurements on the sample, the water utility providing wastewater flow at  
897 their treatment works, the public health body providing case data, or the weather service providing  
898 rainfall data for monitored areas. Collating data from such a broad range of institutions requires  
899 extensive data sharing agreements, which in some cases must be purchased, but in many other cases  
900 rely on effective partnerships born out of years of collaboration, or the power of government support.  
901 An example from the programme in Wales of the prior would be the 2022 strategic partnership between  
902 Cardiff University and Dŵr Cymru Welsh Water for future collaboration and research which has  
903 facilitated the exchange of expertise and data sharing.

## 904 **6. Conclusion**

905 Despite the paucity of readily available historical information on wastewater systems in Wales  
906 compared to São Paulo, wastewater infrastructure, governance and history were seen to be key drivers  
907 in shaping wastewater surveillance programmes. Wastewater is predominantly collected through  
908 networked systems in both countries, bringing waste to centralised WWTPs, but significant populations

909 also remain off-grid. While the off-grid populations in São Paulo state are primarily in urban irregular  
910 settlements, the Welsh off-grid populations are in mostly in upland rural settings, and in both cases, this  
911 usually concerns the most deprived demographics. Ease of access to wastewater samples that capture  
912 significant, and representative, portions of the population is clearly a determinant of the cost-benefit of  
913 a wastewater surveillance programme. Wales has an advantage when compared to São Paulo state due  
914 to the high WWTP density over a small geographic area and population, allowing samples to be taken  
915 easily from smaller populations. Besides access challenges, another novel finding is that stakeholder  
916 engagement constrains programme design, laboratory methods and data analysis. Despite challenges,  
917 achieving a near ideal programme, that delivers reliable, high-resolution, and accurate data in near-real-  
918 time, is in the domain of the possible. Costs involved in implementing such a programme will depend  
919 on existing infrastructure as well as the level of data and expertise sharing between stakeholders. As the  
920 world shifts from pandemic to endemic SARS-CoV-2 levels, surveillance programme funding has come  
921 under intense scrutiny. Our work indicates that there are clear benefits in maintaining wastewater  
922 surveillance programmes and their network of stakeholders over the long-term for pandemic  
923 preparedness, but also as they can broaden their applications to other health indicators.

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