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Catching a resurgence: Increase in SARS-CoV-2 viral RNA identified in wastewater 48 h before COVID-19 clinical tests and 96 h before hospitalizations

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HIGHLIGHTS
• SARS-CoV-2 WW viral signal increases 48 h prior to increase in new COVID cases.
• SARS-CoV-2 WW viral signal increases 96 h prior to increase in COVID hospitalisations.
• Time-lagged correlations confirm WBE predates new COVID cases and hospitalisations.
• WBE may be better than clinical testing at detecting surges in COVID cases.
• RNA epidemiological metric may become a valuable public health metric.

GRAPHICAL ABSTRACT

ABSTRACT
Curtailing the Spring 2020 COVID-19 surge required sweeping and stringent interventions by governments across the world. Wastewater-based COVID-19 epidemiology programs have been initiated in many countries to provide public health agencies with a complementary disease tracking metric and non-discriminating surveillance tool. However, their efficacy in prospectively capturing resurgences following a period of low prevalence is unclear. In this study, the SARS-CoV-2 viral signal was measured in primary clarified sludge harvested every two days at the City of Ottawa’s water resource recovery facility during the summer of 2020, when clinical testing recorded daily percent positivity below 1%. In late July, increases of >400% in normalized SARS-CoV-2 RNA signal in wastewater were identified 48 h prior to reported >300% increases in positive cases that were retrospectively attributed to community-acquired infections. During this resurgence period, SARS-CoV-2 RNA signal in wastewater
1. Introduction

Wastewater has been used as a window in the habits and health of populations. A relatively novel epidemiological technique called wastewater-based epidemiology (WBE) has been successfully used by public health units and authorities for decades to track trends in the general population for numerous diverse substances: illicit and recreational drugs (Been et al., 2014; Feng et al., 2018; González-Mariño et al., 2016; Gushgari et al., 2018), toxic and harmful chemicals and pesticides (Du et al., 2018; Markosian and Mirzoyan, 2019; Roussis et al., 2017), as well as prevalence of disease and illness (Nordgren et al., 2009; Yang et al., 2017). An intense interest has developed in looking to sewage to seek out biomarkers, pathogens and viruses to further the understanding of the prevalence of diseases, diets and habits in the general population (Bowes and Halden, 2019; Lorenzo and Picó, 2019). The earliest occurrence of using sewage as a source of information was approximately 25 years ago, where scientists discovered that polio was in fact found and transmissible in sewage and human stool (Metcalf et al., 1995). One of the major advantages of WBE is that with well-designed sampling/testing regimes, it can be utilized to gain valuable, and sometimes otherwise unobtainable information for subsets of population, as small as a single building or neighborhood, or as large as a city or county (Choi et al., 2018; Tscharkie et al., 2019).

Most recently, WBE has been applied to the quantification of COVID-19 disease signal. The COVID-19 global pandemic has had profound impacts on day-to-day life through sickness, death and the economic and social effects of lockdowns, closures and curfews put in place by local governments to control and limit the community spread of the disease. At the beginning of the pandemic, COVID-19 surveillance approaches invoked by nations around the world included case ascertainment of individual patients via nucleic acid-based tests, serological tests and contact tracing. As such, WBE has been investigated as a potential surveillance tool to obtain a rapid measure of the relative incidence rate of COVID-19 infections in the population at large, generating an entirely new epidemiological metric for public health units (PHU) to utilize. Fecal shedding of SARS-CoV-2 viral particles has been shown to occur before, during and after active COVID-19 infections, for periods ranging from a few days to several weeks (Chen et al., 2020; Cheung et al., 2020; Xing et al., 2020; Xu et al., 2020), with current literature suggesting that children may shed viral particles longer than adults (Li et al., 2020b; Ma et al., 2020). Additionally, fecal shedding of SARS-CoV-2 viral particles in stool occurs in patients infected with COVID-19 and appears to be independent of the presence of SARS-CoV-2 viral particles in the upper respiratory tract.

WBE has been applied at wastewater resource recovery facilities (WRRFs), with early studies demonstrating that viral signal is detected from both influent and primary clarified sludge, with recent studies identifying that at WRRFs, primary clarified sludge may provide more sensitive measurements than in influent (D’Aoust et al., 2021; Graham et al., 2020; Peccia et al., 2020). Additionally, recent studies outlined that viral particles of SARS-CoV-2 in wastewater could in fact be predominantly solid particle-bound (Canadian Water Network (CWN), 2020; Graham et al., 2020) due to the virus’ lipid bilayer, which causes it to be lipophilic (Schoeman and Fielding, 2019). Furthermore, it was observed in several studies that SARS-CoV-2 viral signal in wastewater could be utilized as a tool to track epidemiological metric trends (Ahmed et al., 2020; Medema et al., 2020a; Nemudriy et al., 2020; Thompson et al., 2020). One of the recent developments to the field of WBE is the renewed interest in investigating biomarkers to normalize the SARS-CoV-2 viral signal. Previous WBE work identified several possible normalization biomarkers in wastewater such as 5-Hydroxyindoleacetic acid (Chen et al., 2014), caffeine and cotinine (Rico et al., 2017). However, due to the higher incidence of SARS-CoV-2 viral particles in feces as compared to urine (Collinvignarelli et al., 2020; Jeong et al., 2020; Lodder and de Roda Husman, 2020; Zhurakivska et al., 2020), other biomarkers present in wastewater solids, such as pepper mild mottle virus (PMMoV) (Jafferli et al., 2020; Kitajima et al., 2018; Rosario et al., 2009; Symonds et al., 2014) and cAsspHage (Crank et al., 2020; Polo et al., 2020) saw are being utilized for the normalization of SARS-CoV-2. Some recent studies have demonstrated that biomarker normalized SARS-CoV-2 viral signal may in fact track be capable of tracking the incidence of COVID-19 infections in the community (Gerlity et al., 2021; Kitamura et al., 2021) more effectively than non-biomarker-normalized viral signal (Jafferli et al., 2020; Wu et al., 2020), even if accounting for corrections due to flow or solids mass flux (D’Aoust et al., 2021).

To appropriately utilize epidemiological data obtained by measuring SARS-CoV-2 viral signal in wastewater, PHUs require an understanding of the predictive ability of this epidemiological metric to appropriately action obtained data (Hart and Halden, 2020; Hill et al., 2020; Thompson et al., 2020). Recent studies appear to demonstrate that WBE was able to identify the onset of community infection prior to clinical testing during the first wave, (D’Aoust et al., 2021; Kaplan et al., 2020; La Rosa et al., 2020; Medema et al., 2020b; Nemudriy et al., 2020; Peccia et al., 2020; Sherchan et al., 2020; Vallejo et al., 2020; Wu et al., 2020) however, there isn’t a clear consensus on how early WBE can detect resurgences of the disease. Due to reporting limitations or absence of widespread clinical PCR or serological testing at the onset of the first wave of the COVID-19 pandemic, the predictive qualities of SARS-CoV-2 viral signal tracking in wastewater may not have been fully assessed, creating a critical and fundamental lack of understanding of intrinsic predictive qualities of wastewater-based SARS-CoV-2 surveillance, particularly for subsequent resurgences of COVID-19 infections in the community. As such, it remains unclear whether WBE can have a role in identifying COVID-19 resurgences before other, more traditional epidemiological metrics, such as clinical and serological testing results. As a result, the objective of this study is to establish a relationship between SARS-CoV-2 RNA signal in wastewater during a resurgence in a community with established clinical testing and (i) increases in SARS-CoV-2 RNA signal in wastewater, (ii) increases in the number of new COVID-19 positive patient cases in the community, and (iii) increases in the number of new hospitalizations of COVID-19 positive patients in the community.

2. Materials and methods

2.1. Characteristics of the City of Ottawa’s water resource recovery facility

Primary clarified sludge samples were harvested from the City of Ottawa’s Robert O. Pickard Environmental Centre, in Ottawa, ON, which services approximately 1 M residents of the national capital region. The hydraulic residence time of the Ottawa sewershed ranges from 2 h to 35 h, with an average residence time of approximately 12 h. The Ottawa water resource recovery facility (WRRF) has an average daily flow of 435,000 m³ per day. The facility is comprised of preliminary treatment consisting of coarse and fine screening and grit chambers for the removal of larger particles. The primary treatment units consist of rectangular primary clarifiers, where samples from...
Evaluation of viral recovery efficiency was also performed as per D’Aoust et al. (2021); with the data in this study not being corrected for recovery efficiency. Dilution tests were periodically performed to determine if inhibition was present in the samples. The method’s limit of detection for the N1 and N2 gene regions was accessed by determining the concentration at which a detection rate of ≥95% (<5% false negatives) was obtained, as per the MIQE recommendations (Bustin et al., 2009). Additionally, samples were discarded if they did not meet the following requirements: 1) standard curves are linear (R2 ≥ 0.95), 2) the copies/well are found in the linear range of the standard curve, and 3) the primer efficiency was between 90%–130%. Additionally, all samples were analyzed as technical triplicates and samples with values greater than 2 standard deviations away from the mean were discarded.

2.3. COVID-19 epidemiological data

During the study period, there were three main methods of COVID-19 surveillance used to compare to the wastewater RNA signal of SARS-CoV-2, with the metrics being commonly communicated to the public. The first method was to observe the daily number of new confirmed COVID-19 infections, determined via laboratory-confirmed SARS-CoV-2 tests using reverse transcriptase-quantitative polymerase chain reaction (RT-qPCR) that targeted the RNA-Dependent RNA polymerase for multiple genes. Tests were classified as positive for SARS-CoV-2 infection if at least one of the two gene “targets” were detected (Government of Ontario, 2020). There was a notable increase in COVID-19 testing in Ottawa during early June with an average daily testing increase from approximately 270 tests per day to approximately 800 tests per day. As a result of increased testing resources, additional testing was performed at dedicated community clinics, physician offices and in hospitals, retirement homes and long-term care facilities, ultimately improving the quality of clinical epidemiological data being generated daily.

A second method used for surveillance of COVID-19 infection is the test positivity rate of clinical testing within the health unit boundaries. COVID-19 test positivity is the proportion of positive COVID-19 cases of Ottawa residents reported as a percent of all individuals tested for COVID-19 on a specific date Test positivity is reported in this study on a daily basis and also as a seven-day mid-point average.

Fig. 1. Bird’s-eye view of the Ottawa WRWF, with the primary clarifiers identified.
The third method used for COVID-19 surveillance was the daily hospital census counts, defined as the number of patients residing in Ottawa with confirmed COVID-19 infections as of midnight that day that were admitted to any of Ottawa’s five acute care hospitals. The average length of stay during the study period was 19.5 days, interquartile range 22 days.

2.4. Statistical analysis

In order to test for significance and for strength of correlation between SARS-CoV-2 RNA signal and epidemiological metrics, a student’s t-test and Pearson’s correlation analyses were performed, with a p-value of 0.05 or lower signifying significance.

To evaluate if a lag existed between the appearance of increased SARS-CoV-2 RNA signal in wastewater and epidemiological metrics, time-step analyses were also performed where the correlations between viral RNA signal and epidemiological metrics were offset by a period of 1 to 7 days. The time-step analysis was performed across three discrete time-periods: i) pre-resurgence and resurgence, ending after the resurgence (June 21st to July 21st), ii) pre-resurgence and post-resurgence, extending to the end of the data set (June 21st to July 25th) and iii) full data period (June 21st to August 4th).

3. Results, discussion and implications

3.1. Trend and peak of wastewater SARS-CoV-2 RNA viral level in wastewater compared to other clinical COVID-19 surveillance metrics

This study compares PMMoV normalized SARS-CoV-2 viral copies, a metric outlined in previous SARS-CoV-2 WBE work (D’Aoust et al., 2021; Kitamura et al., 2021; Wu et al., 2020) with clinical epidemiological metrics. This WBE-based metric normalizes SARS-CoV-2 viral copies to the number of copies of PMMoV in wastewater to account for variations inherent to wastewater samples; in particular variations related to the wastewaters water quality, solids and fecal matter concentrations. In addition, PMMoV normalizes for variations associated with sampling, storage, and testing. PMMoV is ubiquitous in both the liquid and solid fraction of wastewater and is effective due to its low temporal variability in several locales due to the metrics’ ability to correct for the true quantity of fecal matter in wastewater (D’Aoust et al., 2021; Wu et al., 2020). It is hypothesized that due to the fact PMMoV signal in wastewater is likely diet based (due to consumption peppers, some tomatoes, or related foodstuff infected with PMMoV) (Jarret et al., 2008; Peng et al., 2015; Zhou et al., 2020), this approach will work best when performing a survey of a larger group of individuals or geographical area, where enough survey participants with varied enough diets will average out the signal.

Fig. 2. Ct values of the SARS-CoV-2 N1 and N2 gene regions, and the PMMoV normalization gene (1:10 dilution shown), outlining the low variability of PMMoV, justifying its use as a normalization gene. The 10th and 90th percentile are displayed, along with the median (dotted line inside shapes). The variance of the different targets is also shown at the top.

Positive correlations were observed (Fig. 3c, d and e) between the normalized viral RNA signal and both the number of new daily positive COVID-19 cases (N1: \( R = 0.673, p < 0.001 \); N2: \( R = 0.648, p < 0.001 \)) and clinical testing percent positivity (N1: \( R = 0.468, p < 0.001 \); N2: \( R = 0.404, p < 0.001 \)). These findings are in agreement with reported observations from previous studies (D’Aoust et al., 2021; Nemudryi et al., 2020; Peccia et al., 2020). This study, as opposed to an earlier study in the same region (D’Aoust et al., 2021), demonstrates that strong correlations do in fact exist between viral RNA signal and the...
epidemiological metric of clinical daily new COVID-19 cases, as supported by several other studies (Hart and Halden, 2020; Michael-Kordatou et al., 2020; Trottier et al., 2020; Wu et al., 2020). The moderate correlations between viral RNA signal and clinical daily new COVID-19 cases in previous work was largely attributed to inadequate resources to achieve a developed clinical testing framework early in the pandemic (lower daily # of tests). In this study, a strong testing regiment was deployed in Ottawa and hence produced a more reliable daily new COVID-19 cases metric across this study period. A weak positive correlation exists between N1 and N2 SARS-CoV-2 PMMoV-normalized RNA signal and COVID-19-caused hospitalizations ($R = 0.347$, $p < 0.001$ and $R = 0.464$, $p < 0.001$ for N1 and N2, respectively).

Fig. 3. Epidemiological metrics and SARS-CoV-2 viral concentrations over the study period: a) COVID-19-caused hospitalizations, b) percent test positivity (7-day mid-point floating average), c) percent test positivity and d) number of new daily cases, along with e) wastewater SARS-CoV-2 N1 and N2 gene copies/PMMoV gene copies. Open data points signify points which were below the limit of quantification. The dotted line in the background represents the number of daily clinical tests/100 K pop. Performed.
respectively; Fig. 3a and e). This supports previous findings that WBE efforts to track the rate of incidence of COVID-19 infections in the community provides useful and confirmatory information.

3.3. Temporal association between surveillance metrics

A visual comparison of clinical testing percent positivity and the number of new daily positive COVID-19 cases compared to the PMMoV-normalized RNA signal shows that the viral RNA signal precedes both the clinical testing percent positivity and new daily positive COVID-19 cases epidemiological data.

A times-step correlation analysis suggests that increases in SARS-CoV-2 signal in wastewater precedes increases in new daily positive COVID-19 cases and clinical testing percent positivity by two days (N1: $R = 0.703$, $p < 0.001$; N2: $R = 0.721$, $p < 0.001$; N1: $R = 0.703$, $p < 0.001$; N2: $R = 0.714$, $p < 0.001$) (Table 1). Furthermore, the same time-step correlation analysis also suggests that increases in SARS-CoV-2 signal in wastewater precedes increases in the number of hospitalized cases by four days (N1: $R = 0.741$, $p < 0.001$; N2: $R = 0.767$, $p < 0.001$). This is further confirmed visually (Fig. 4). The precedence of viral RNA signal vs. other epidemiological metrics has been reported elsewhere (Chavarria-Miró et al., 2020; Kumar et al., 2020; Peccia et al., 2020).

Recent studies proposed that wastewater viral SARS-CoV-2 RNA signal may in fact precede hospital admissions by 1 to 5 days in primary sludge (Fig. 4) (Kaplan et al., 2020; Peccia et al., 2020), and 6–14 days in raw wastewater (Kumar et al., 2020). This supports findings by Peccia et al. (2020), identifying that hospitalizations may begin peaking four days after an increased viral RNA signal, with the caveat that hospitalizations rates may be significantly higher when viral RNA signal is the result from shedding of an at-risk population group (young infants, elderly, immunocompromised, etc.). In addition, the RNA signal post-resurgence indicates that the resurgence events in viral RNA signal serves to weaken the correlation of subsequent peaks with clinical testing metrics. It is hypothesized that early and later infections and overlapping shedding events of duration and intensity convolute peaks within a period of time following the initial new peaks/surges in COVID-19 infections. Nonetheless, the RNA epidemiological metric of copies/copies PMMoV may become extremely valuable to public health units in their preparation for the next COVID-19 resurgences.

Earlier studies investigating the predictive capabilities of SARS-CoV-2 WBE (Kumar et al., 2020; Peccia et al., 2020) may have been performed at times where clinical metrics were dampened due to limited clinical testing resources, which may in turn have impacted the conclusions regarding predictive ability. This present study was performed during a period of relatively higher testing intensity in the community (avg. 80.9 tests per/100 K pop. performed) and confirms previous findings of early detection of SARS-CoV-2 WBE as compared to clinical metrics such as hospitalizations, new daily positives cases or percent positivity.

4. Conclusions

This study demonstrates that due to the general stability of SARS-CoV-2 RNA measurements in PCS, this fraction of wastewater can reliably be utilized to predict rapid increases or resurgences in COVID-19 cases in the community at large (Alpaslan-Kocamemi et al., 2020; D’Aoust et al., 2021; Peccia et al., 2020), and the results strongly suggest that PCS can be utilized as part of sentinel-type wastewater based epidemiology endeavors at WRRFs.

Advantages of this approach include: i) the analyses in water resource recovery facilities are high-enrollment surveys, which capture a majority of the urban population, regardless of individual’s willingness to get tested clinically; ii) tests are anonymous and do not pose ethical challenges as opposed to mass clinical testing or targeted sewershed monitoring efforts; iii) data can be used as a confirmation to clinical tests, increases the effectiveness of local public health units and data remains independent of official testing strategies or media sentiment, and iv) provides an exit strategy/path forward for public health units as

| Table 1 | Time-step analyses of correlations (Pearson’s R) between normalized SARS-CoV-2 viral RNA signal (copies/copies PMMoV) and 7-day rolling average percent positivity, test percent positivity and daily new cases epidemiological metrics. |
|---------|--------------------------------------------------------------------------------|
| **Hospitalized cases** | **7-day rolling average test percent positivity** | **Test percent positivity** | **Daily new cases** |
| **Daily offset** | N1 | N2 | N1 | N2 | N1 | N2 | N1 | N2 |
| **June 21 - July 21** | | | | | | | | |
| 0 | 0.488 | 0.502 | 0.675 | 0.662 | 0.551 | 0.635 | 0.680 | 0.754 |
| 1 | 0.599 | 0.578 | 0.622 | 0.647 | 0.634 | 0.723 | 0.642 | 0.692 |
| 2 | 0.532 | 0.487 | 0.482 | 0.497 | 0.485 | 0.492 | 0.485 | 0.492 |
| 3 | 0.747 | 0.754 | 0.558 | 0.593 | 0.191 | 0.208 | 0.384 | 0.310 |
| 4 | 0.741 | 0.767 | 0.415 | 0.457 | 0.338 | 0.332 | 0.265 | 0.237 |
| 5 | 0.512 | 0.487 | 0.301 | 0.341 | 0.144 | 0.140 | 0.375 | 0.405 |
| 6 | 0.534 | 0.490 | 0.104 | 0.167 | -0.031 | 0.023 | 0.107 | 0.190 |
| 7 | 0.375 | 0.361 | -0.012 | 0.077 | -0.040 | 0.070 | 0.078 | 0.178 |
| **June 21 - July 25** | | | | | | | | |
| 0 | 0.539 | 0.551 | 0.639 | 0.666 | 0.505 | 0.612 | 0.707 | 0.712 |
| 1 | 0.689 | 0.562 | 0.631 | 0.653 | 0.657 | 0.672 | 0.662 | 0.709 |
| 2 | 0.634 | 0.489 | 0.623 | 0.667 | 0.724 | 0.761 | 0.699 | 0.763 |
| 3 | 0.768 | 0.686 | 0.488 | 0.576 | 0.128 | 0.203 | 0.277 | 0.346 |
| 4 | 0.790 | 0.690 | 0.324 | 0.437 | 0.264 | 0.277 | 0.245 | 0.192 |
| 5 | 0.632 | 0.483 | 0.188 | 0.303 | 0.114 | 0.108 | 0.378 | 0.354 |
| 6 | 0.645 | 0.482 | 0.038 | 0.118 | -0.073 | 0.008 | 0.102 | 0.202 |
| 7 | 0.531 | 0.409 | -0.074 | 0.015 | -0.157 | 0.041 | -0.032 | 0.163 |
| **Whole data set** | | | | | | | | |
| 0 | 0.347 | 0.470 | 0.644 | 0.626 | 0.469 | 0.527 | 0.673 | 0.648 |
| 1 | 0.453 | 0.484 | 0.606 | 0.602 | 0.634 | 0.668 | 0.654 | 0.634 |
| 2 | 0.429 | 0.424 | 0.593 | 0.603 | 0.703 | 0.714 | 0.703 | 0.721 |
| 3 | 0.566 | 0.565 | 0.448 | 0.491 | 0.027 | 0.081 | 0.182 | 0.375 |
| 4 | 0.803 | 0.886 | 0.307 | 0.360 | 0.233 | 0.278 | 0.223 | 0.158 |
| 5 | 0.497 | 0.477 | 0.165 | 0.234 | 0.079 | 0.034 | 0.306 | 0.206 |
| 6 | 0.546 | 0.500 | 0.020 | 0.064 | -0.001 | 0.021 | 0.074 | 0.195 |
| 7 | 0.488 | 0.443 | -0.080 | -0.119 | -0.074 | 0.005 | -0.025 | 0.178 |
something to eventually transition to once immunization is available, to cut down on clinical testing costs, while maintaining broad surveillance capabilities. However, limitations to this WBE approach includes the following: i) data obtained only gives high-altitude view of the situation in the community, and it is currently not possible to correlate these data to an actual number of cases in the community; ii) testing protocols must be optimized and retain sensitivity at very-low levels of disease incidence to remain relevant; iii) widespread adoption of WBE requires scale-up of public health units’ current capabilities and/or partnership with private laboratories or research institutions, and iv) increases in resolution (more localized, upstream of water resource recovery facilities, for example) of tests in watersheds may lead to ethical questions for public health unit if it wishes to act upon the data, due to the potential risk for identifying or singling out small subgroups of a population. It is noted however that these issues are not dissimilar to other ethical issues currently existing with other COVID-19 public health endeavors, such as COVID-19 potential exposure notification smartphone applications.

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**CRediT authorship contribution statement**

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**Declaration of competing interest**

The authors declare that no known competing financial interests or personal relationships could appear to influence the work reported in this manuscript.

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**Appendix A. Supplementary data**

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