

1 **Bias of library preparation for virome characterization in untreated and treated**
2 **wastewaters**

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1 **Abstract**

2 The use of metagenomics for virome characterization and its implementation for
3 wastewater analyses, including wastewater-based epidemiology, has increased in the
4 last years. However, the lack of standardized methods can led to highly different results.
5 The aim of this work was to analyze virome profiles in upstream and downstream
6 wastewater samples collected from four wastewater treatment plants (WWTPs) using
7 two different library preparation kits. Viral particles were enriched from wastewater
8 concentrates using a filtration and nuclease digestion procedure prior to total nucleic
9 acid (NA) extraction. Sequencing was performed using the ScriptSeq v2 RNA-Seq (LS)
10 and the NEBNext® Ultra™ II RNA (NB) library preparation kits. Cleaned reads and
11 contigs were annotated using a curated *in-house* database composed by reads assigned
12 to viruses at NCBI. Significant differences in viral families and in the ratio of detection
13 were shown between the two library kits used. The use of LS library showed
14 *Virgaviridae*, *Microviridae* and *Siphoviridae* as the most abundant families; while
15 *Ackermannviridae* and *Helleviridae* were highly represented within the NB library.
16 Additionally, the two sequencing libraries produced outcomes that differed in the
17 detection of viral indicators. These results highlighted the importance of library
18 selection for studying viruses in untreated and treated wastewater. Our results underline
19 the need for further studies to elucidate the influence of sequencing procedures in
20 virome profiles in wastewater matrices in order to improve the knowledge of the virome
21 in the water environment.

22

23 **Keywords:** Wastewater, Metagenomics, Enteric viruses, viability RT-qPCR.

24

1 **1. Introduction**

2 The reuse of water, including for irrigation, cooling, and other non-potable applications
3 is an emerging topic due to climate change and water scarcity. Treatment and
4 regeneration of household sewage water in urban regions are usually performed by
5 wastewater treatment plants (WWTPs); however, they are not always able to completely
6 eliminate the microbiological risks present in treated wastewaters (Chalmers et al.,
7 2010; Randazzo et al., 2019; Sano et al., 2016). Fecal bacteria have traditionally been
8 used as indicators for the presence of pathogenic microorganisms even though they fail
9 to detect the presence of human pathogenic enteric viruses (Eslamian, 2016; Gerba et
10 al., 2013; Kitajima et al., 2014). Thus, several viruses (i.e. crAssphage, Pepper mild
11 mottle virus, adenovirus, polyomavirus, ...) have been proposed as indicators because
12 of their similarity to pathogenic viruses in terms of environmental stability and
13 resistance to wastewater sanitation treatments (Farkas et al., 2020). The presence of
14 human enteric viruses in treated wastewaters has been well documented (Gerba et al.,
15 2018; Randazzo et al., 2019; Sano et al., 2016), posing public health risk-related
16 concerns also because of their stability into the environment. Thus far, nearly one
17 hundred different types of human enteric viruses are known, which cause a variety of
18 illnesses and diseases in humans (Fong and Lipp, 2005), primarily gastroenteritis and
19 hepatitis, and new pathogenic strains and species continue to be discovered. Among
20 others, the viruses most commonly detected in untreated and treated wastewaters
21 include human norovirus, adenovirus (AdV), enterovirus (EV), sapovirus (SaV),
22 astrovirus (HAstV), rotavirus A (RV), and hepatitis A and E viruses (HAV and HEV)
23 (Haramoto et al., 2018). Surveillance of human enteric viruses in untreated and treated
24 wastewaters is performed by molecular procedures (e.g., real time PCR (qPCR) or
25 digital PCR (dPCR)) (Haramoto et al., 2018). Currently, a wastewater-based

26 epidemiology surveillance has been globally implemented to monitor COVID-19
27 disease, caused by severe acute respiratory syndrome coronavirus 2 (SARS-CoV-2)
28 with notable implications for public health response in local settings (Bivins et al.,
29 2020; Polo et al., 2020). These approaches require reference sequences for primer and
30 probe design which limit the number and variety of viruses to be analyzed.
31 Alternatively, recent shotgun or untargeted metagenomic approaches enable the
32 simultaneous identification of viral sequences from a sample, referred to as ‘virome’,
33 which is a diverse community of mainly eukaryotic RNA and DNA eukaryotic viruses
34 and bacteriophages. Virome characterization in wastewater provides a potential solution
35 to the challenges associated with the traditional surveillance of viruses in sewage
36 (Nieuwenhuijse and Koopmans, 2017).

37 In this pilot study, we have used metagenomics analyses using two different library
38 preparation kits for metagenomic sequencing to characterize the virome composition in
39 influent and effluent samples from four different WWTPs. Thus, the objectives of this
40 study were to: 1) evaluate different sequencing libraries for virome characterization; and
41 2) investigate virome distribution and diversity in influent and effluent samples.

42 **2. Materials and Methods**

43 **2.1. Sample processing**

44 Five-hundred mL of influent (IW) and effluent (EW) grab samples from four different
45 WWTPs were collected in November 2018 in Valencia (Spain). Treatment plants
46 differed in the number of equivalent population, the volume of treated wastewater and
47 the disinfection treatments (Table S1). *Escherichia coli* counts, expressed as Most
48 Probably Number (MPN), were performed using the Colilert[®] kit (IDDEX Laboratories,
49 Spain) following the ISO 9308-2:2012 standard on the same sampling day. Samples
50 were kept for further analyses at -20°C and thawed for 12 h at approximately 20 °C

51 before processing. After thawing, 200 mL of each sample were inoculated with 7 log
52 PCRU/L of mengovirus (MgV) vMC₀ (CECT 100000), used as a process control.
53 Samples were processed using the aluminum-based precipitation protocol described
54 elsewhere (AAVV, 2018; Randazzo et al., 2019). Briefly, 200 mL of sample was
55 adjusted to pH 6.0. The Al(OH)₃ precipitate was performed mixing 1 part of AlCl₃ 0.9N
56 per 100 parts of sample and the solution was mixed at 150 rpm for 15 min. Then,
57 samples were centrifuged at 1,700 x g for 20 min and the pellet was resuspended in 10
58 mL of 3% beef extract (pH 7.4) and shaken at room temperature (RT) for 10 min at 150
59 rpm. Finally, samples were centrifuged for 30 min at 1,900 x g and the resulting pellet
60 was resuspended in 1 mL phosphate saline buffer (PBS, pH 7.4) and stored at -80°C.

61 **2.2. Sample processing for metagenomics**

62 Viral particles were enriched from sample concentrates (n=8) following the NetoVIR
63 protocol, which includes both filtration and nuclease digestion steps (Conceição-Neto et
64 al., 2015). In brief, 500 µL of concentrates were homogenized using the MP FastPrep24
65 5G (MP Biomedicals, Spain) for 40 seconds at a speed of 6.0. The homogenate was
66 centrifuged at 16,000 × g for 3 min and 200 µL of the supernatant was filtered through a
67 0.8 µm PES filter (Sartorius, UK) to remove large particles. The filtrate was incubated
68 with benzonase (Millipore, Spain) and micrococcal nuclease (New England Biolabs,
69 USA) enzymes at 37°C for 2 h to degrade free nucleic acids. Capsid protected viral
70 nucleic acids were extracted with the NucleoSpin[®]RNA virus kit (Macherey-Nagel
71 GmbH & Co., Germany), according to the manufacturer's instructions, without adding
72 carrier RNA. Thus, both DNA and RNA viral nucleic acids were concomitantly
73 extracted. Nucleic acids were eluted in 50 µL RNase-free water. Libraries were
74 generated from 1 to 50 ng of a DNA-RNA sample using two different library
75 preparation kits. The first library preparation kit was the ScriptSeq v2 RNA-Seq Library

76 Preparation Kit (Illumina, USA), referenced as LS, with slight modifications. An initial
77 denaturation step (95 °C for 5 min) was added to the protocol, and PCR cycles were
78 increased to 20 to obtain enough library concentration to sequence. Additionally, the RT
79 enzyme from the original library preparation kit was substituted by the AMV Reverse
80 Transcriptase (Promega, Spain). The second library preparation kit was the NEBNext®
81 Ultra™ II RNA Library Prep Kit (New England BioLabs Inc., Ipswich, UK) (referenced
82 as NB) following manufacturer's instructions. The two libraries compared in this study
83 differ in terms of fragmentation times, enzymes, cDNA synthesis conditions, primers
84 used in the PCR, as well as the conditions for aforesaid amplification (Table S2).
85 Libraries were normalized, pooled, and sequenced using the NextSeq™ 500 platform
86 (Illumina), following the manufacturer's protocol, with a configuration of 150 cycles
87 paired-end reads. Sequencing was performed by Lifesequencing S.L. (Valencia, Spain).

88 **2.3. Data analyses**

89 Obtained reads were cleaned for adaptor removal using cutadapt software (Martin,
90 2011) with a minimum overlap of 5 nucleotides between read and adaptor and a
91 maximum error rate of 0.1. Reads were cleaned with the *reformat.sh* script from
92 BBDMap software (sourceforge.net/projects/bbmap/) in order to remove nucleotides from
93 both ends with Phred scores lower than 20 and reads shorter than 50 bp. Cleaned reads
94 were merged in to single reads with FLASH v1.2.11 (Magoč and Salzberg, 2011)
95 allowing outies. Additionally, cleaned reads were assembled with Ray 2.3.1. (Boisvert
96 et al., 2012) using 31-mers.

97 Merged reads and contigs were taxonomically annotated using BLASTn algorithm
98 (Boratyn et al., 2013) with a manually curated *in-house* database constructed with all
99 the viral sequences (NCBI:txid10239; release May, 5 2020) available at GenBank
100 (<https://www.ncbi.nlm.nih.gov/nuccore/?term=viruses%5Borganism%5D>). For the

101 BLASTn analysis of viral reads against this curated *in-house* database, a cut off of 70%
102 of query sequence coverage and 80% of identity was used, respectively. Rarefaction
103 curves and diversity indexes Shannon and Simpson were calculated with R package
104 *vegan* v2.5-6.

105

106 **2.4. Virus quantification**

107 For virus quantification an optimized viability RT-qPCR was applied as previously
108 described (Randazzo et al., 2019). In brief, 150 µL sample concentrates were added to
109 50 µM PMAxx (Biotium, USA) and 0.5% Triton 100-X (Thermo Fisher Scientific,
110 Spain) and incubated in the dark at RT for 10 min at 150 rpm. Then, samples were
111 exposed to photo-activation using a photo-activation system (Led-Active Blue, GenIUL,
112 Spain) for 15 min. RNA was extracted using the NucleoSpin® RNA virus kit
113 (Macherey-Nagel GmbH & Co.) according to the manufacturer's instructions including
114 the Plant RNA Isolation Aid (Ambion, Spain) pretreatment. Primers, probes and RT-
115 qPCR conditions for norovirus GI, norovirus GII, RV, HAV, HEV, mengovirus and
116 HAstV quantification have been previously reported (Randazzo et al., 2019, Cuevas-
117 Ferrando et al., 2020).

118 For crAssphage quantification by qPCR, the primer set CPQ_064 described by Stachler
119 et al. (2017) was used. PCR conditions were an initial denaturation step of 30 seconds at
120 95°C followed by 45 cycles of 5s at 95°C and 30s at 60°C. The Premix Ex Taq master
121 mix for probe-based real-time PCR kit (Takara, France) was used for the reaction. For
122 the crAssphage quantification, the standard curve was performed with a customized
123 gBlock® fragment (Integrated DNA Technologies, Spain) of 228 bp that contained the
124 crAssphage sequences used for amplification.

125 Limit of quantification, qPCR efficiency and standard curve R^2 values for all the tested
126 genes are displayed in Table S3. For all RT-qPCR assays, undiluted and ten-fold diluted
127 RNA was tested to check for RT-qPCR inhibitors.

128 **2.5. Correlation and similarity analyses**

129 Correlation analyses were carried out between data sets obtained by both libraries at
130 family level, and between metagenomics and RT-qPCR results using the R package
131 *Hmisc* v4.2-0 (<https://CRAN.R-project.org/package=Hmisc>) and applying the Spearman
132 method (ρ). Significance was set at 0.05. Representation of correlation matrix values
133 was performed with the R library *corrplot* v0.84 ([https://CRAN.R-](https://CRAN.R-project.org/package=corrplot)
134 [project.org/package=corrplot](https://CRAN.R-project.org/package=corrplot)).

135 For each individual sample, the Jaccard index was used to analyze the similarity among
136 results obtained with both libraries. Calculations were performed using R package
137 *betapart* v1.5.2 (Baselga, 2010) taking into account the beta.JAC values representing
138 the overall beta diversity for each sample pair.

139 **3. Results**

140 **3.1. Overview of bias due to library preparation**

141 Each concentrated sample was sequenced using two sequencing libraries: the ScriptSeq
142 v2 RNA-Seq Library Preparation Kit (LS) and the NEBNext® Ultra™ II RNA Library
143 Prep Kit (NB). The average number of reads was 3.2 and 11.5 million for LS and NB
144 libraries, respectively. Rarefaction analyses showed that 5 out of 8 samples sequenced
145 by the LS library reached the plateau, while 2 out of 8 samples sequenced by NB library
146 reached it. Despite that, remaining samples were close to stabilization with both
147 libraries (Fig. S1). Merged viral reads were annotated through a BLASTn comparison
148 with the curated *in-house* database that comprised all the viral sequences (CDS and

149 complete genomes) available at GenBank. For the LS library, the percentage of viral
150 reads ranged from 0.6% to 2.4% in influent and from 0.4% to 4.4% in effluent samples.
151 For NB library, the BLASTn analysis showed a high number of sequences ascribed to
152 the same taxon, suggesting an overrepresentation due to sequencing bias, representing
153 between 33 and 60% of the total viral reads. For that reason, the relative calculations of
154 subsequent analyses were made also taking into account this overrepresentation. These
155 corrected calculations will be called NB-corrected. For the NB library, viral reads
156 ranged from 38% to 58% in influent and from 14% to 24% in effluent samples. Taking
157 into account the results of NB-corrected, these percentages ranged from 9% to 12% and
158 from 7% to 12% in influent and effluent samples, respectively. Shannon and Simpson
159 diversity indexes were calculated for each type of sample (influent or effluent) and for
160 each library (LS or NB) (Fig. 1). Shannon indexes were higher in influent samples
161 sequenced with LS library (mean values of 3.85 ± 0.33 for LS and 1.30 ± 0.12 for NB);
162 however, for effluent samples both indexes showed similar means (1.81 ± 1.21 for LS
163 and 1.90 ± 0.2 for NB), being effluent samples sequenced with LS library more variable
164 ($0.36-3.07$). Similar results were obtained for Simpson index, even though the mean
165 values for influent samples were highly different (0.93 ± 0.02 for LS and 0.62 ± 0.05 for
166 NB).

167 Raw data was deposited at SRA under the Bioproject PRJNA67378 with the following
168 accession numbers: SAMN16633937-SAMN16633944 for ScriptSeq v2 RNA-Seq
169 Library Preparation Kit samples and SAMN16634071-SAMN16634078 for NEBNext®
170 Ultra™ II RNA Library Prep Kit samples.

171 **3.2. Mengovirus recovery**

172 Mengovirus (MgV) was used as a process control to analyze the performance of each
173 library to recover reads and the entire genome of MgV. Its recovery, represented as the

174 percentage of viral reads and the percentage of MgV isolate M genome (L22089.1)
175 obtained for each sample with each library, was different depending on the library used.
176 For LS library, the percentage of viral reads of MgV ranged from 0.05% to 0.79% in
177 influent and from 0.35% to 3.68% in effluent samples. For NB libraries, these values
178 ranged from 0.01% to 0.16% in influent samples, and from 0.63% to 5.77% in effluent
179 samples. However, the percentage of MgV reads with the NB-corrected values were
180 higher in effluent samples, ranging from 1.38% to 11.38%. For the analysis of the
181 recovery of MgV genome, assembled contigs belonging to this species were compared
182 with the genome of Mengovirus isolate M (L22089.1). LS library genome recovery
183 ranges from 6.0% to 95.1%. The highest recovery was obtained in the sample IW3. On
184 the other hand, the coverage of this genome by the NB library ranged from 98.4% to
185 100% (EW3).

186 **3.3. Virome comparison**

187 Regarding the virome composition for each library at family levels, results showed high
188 differences between the two approaches (Fig. 2). While the most represented families
189 with the LS library were *Virgaviridae*, *Microviridae* and *Siphoviridae*; the most
190 abundant families with the NB library were *Ackermannviridae* and *Helleviridae*. These
191 differences allowed the detection of some viral families depending on the library used.
192 For example, families as *Rhabdoviridae*, *Pospiviroidae* or *Mitoviridae* were only
193 detected when the NB library was used for sequencing. Also the taxa uncultured human
194 fecal virus (NCBI:txid239364) and uncultured marine virus (NCBI:txid186617) were
195 only detected with the NB approach. Regarding the families detected with both
196 libraries, only *Genomoviridae* showed total correlation ($\rho=1$) between values obtained
197 with both libraries in influent and effluent samples. For influent wastewater samples,
198 families *Nairoviridae* and *Virgaviridae* showed total correlations; however, this

199 correlation was only observed for *Parvoviridae* in effluent samples. Other families
200 showed high correlations ($\rho=0.8$) in influent wastewaters, as *Peribunyaviridae* and
201 *Picornaviridae*; while *Podoviridae*, *Poxviridae*, *Reoviridae* and *Virgaviridae* families
202 showed high correlations ($\rho=0.8$) in effluent samples. Jaccard indexes showed
203 similarities between the same sample sequenced with each library that ranged from 0.76
204 (IW1) to 0.91 (IW2), with mean values of 0.83 ± 0.09 for IWs and 0.81 ± 0.04 for EWs.

205 **3.4. Analyses of viral fecal indicators by NGS and correlation with enteric** 206 **viruses detected by RT-qPCR**

207 Each of the libraries used in this study showed different power for detecting fecal
208 indicators. Similarly, influent and effluent samples showed different detections rates
209 (Fig. 3A). For example, LS library detected CrAssphage with read percentages higher
210 than 1% but these percentages decreased to less than 0.01% with NB library. Most
211 importantly, LS library was unable to detect the fecal indicator adenovirus in effluent
212 wastewaters, while the NB library detected adenoviruses in percentages between 0.1-
213 1%. The same scenario was observed for the Picobirnavirus indicator. However,
214 indicators families as *Inoviridae*, *Microviridae*, *Myoviridae* and *Podoviridae* showed a
215 better detection with the LS library. *Siphoviridae* family detection did not show
216 differences in its detection capacity between the two different tested libraries, with the
217 exception of sample EW3 (Fig. 3A).

218 Correlation between the number of reads of proposed viral indicators obtained with both
219 libraries and the quantifications obtained by RT-qPCR for enteric viruses along with the
220 *E. coli* counts were calculated. Figure 3B shows the Spearman values of correlation (ρ)
221 calculated with 95% confidence. Norovirus GI and GII showed high correlation values
222 ($\rho>0.8$) with the indicators crAssphage, Picobirnavirus and *Inoviridae*. The highest
223 correlation values ($\rho=0.7$) between RV and indicators reads were crAssphage,

224 *Inoviridae* and *Microviridae*. For HAstrV, high correlations ($\rho > 0.8$) only occurred with
225 crAssphage and *Myoviridae*. HAV and HEV did not correlate with any of the indicators.
226 Interestingly, the proposed indicator AdV showed negative correlations with all the
227 enteric viruses analyzed ($\rho > -0.7$), with the exception of HAV and HEV. Results
228 obtained by NGS for the pepper mild mottle virus (PMMoV), also proposed as viral
229 indicator, showed no correlation with any of the enteric viruses.

230 3.5. Virome comparison between influent and effluent wastewaters

231 Clean reads obtained from each library and each sample were assembled and contigs
232 longer than 200 bp were used for taxonomical classification by using BLASTn
233 algorithm and the *in-house* database. Due to the different results observed at reads level
234 for each library, contigs classification obtained with both libraries for each sample were
235 merged for results representation and virome analysis. The relative abundances of
236 different taxa are shown in Fig. 4. As observed in the heatmap graphic, the most
237 abundant viruses were bacteriophages, as Dickeya phage or Listeria phage WIL-3, even
238 with higher percentages in effluent samples. The higher detection of these phages in
239 treated samples, as occurred with other species (i.e. cucumber green mottle mosaic
240 virus, EBPR podovirus 2, PMMoV, Stealth virus 1, or Tobacco and tomato mosaic
241 viruses), can be due to the decrease of other viruses after wastewater treatment that
242 allows the its detection. Similarly, this effect could be the responsible of the detection of
243 some viruses in effluent samples that were not detected in influent samples, as the case
244 of human adenovirus and human gammaherpesvirus. Some viruses were only found in
245 high percentages in influent samples, as the indicator crAssphage, some *Aeromonas*
246 phages, *Escherichia* phages or viruses belonging to the *Microviridae* family.
247 Wastewater treatments could be the factor that produce this decrease; however, more

248 studies along time from the same WWTPs must be performed in order to ensure the
249 effect of performed treatments.

250 **4. Discussion**

251 The virome of wastewaters have been previously characterized from samples collected
252 around the world (Adriaenssens et al., 2018; Aw et al., 2014; Cantalupo et al., 2011;
253 Fernandez-Cassi et al., 2018; Furtak et al., 2016; Nieuwenhuijse et al., 2020; Rusiñol et
254 al., 2020; Strubbia et al., 2019a, 2019b; Wang et al., 2018); however, much less is
255 known about the virome of effluent samples as only one study has analyzed two effluent
256 samples collected in the UK (Adriaenssens et al., 2018). As far as we know, this is the
257 first study that concomitantly analyzes the RNA and DNA viruses present in influent
258 and effluent samples besides providing a comparison of viruses profiles detected using
259 different sequencing library kits.

260 Results obtained in our study showed high differences regarding not only viruses, but
261 also the power of detection of viral fecal indicators. Both aspects are important for the
262 use of random metagenomics as tool for specific detection. Our results evidenced the
263 influence of the library used for virome studies together with their variability.
264 Additionally, by using MgV as process control for both metagenomic and RT-qPCR
265 analyses, we further assessed the sensitivity of each library, being higher when using
266 NB library. Recoveries of MgV complete genome were between 6.0 and 95.1% for LS
267 library and between 98.4 and 100% for NB library. In contrast, in a recent study, MgV
268 reads were not recovered from spiked water and sediment samples (Adriaenssens et al.,
269 2020). According to the authors, this was likely due to an inclusion of an inactivation
270 step of the DNase at 75°C, which potentially exacerbated the effect of the RNase step
271 (Adriaenssens et al., 2018). The use of models of a virus of interest when comparing
272 sequencing libraries can be an excellent tool for the library selection.

273 For the analysis of the virome of influent and effluent wastewaters, results obtained by
274 both libraries were merged. Phages as crAssphage, *Aeromonas* phages, *Escherichia*
275 phages or viruses belonging to the *Microviridae* family were found in high percentages
276 in influent wastewaters. The absence of these viruses in effluent samples can be due to
277 the sanitation treatments applied in WWTPs, even though further analysis that includes
278 a wider sampling design needs to be performed. These results are in line with previous
279 studies showing a high abundance of bacteriophages families (Aw et al., 2014;
280 Cantalupo et al., 2011; Fernandez-Cassi et al., 2018; Rusiñol et al., 2020; Wang et al.,
281 2018) in influent sewage samples. Nevertheless, other studies showed *Virgaviridae* as
282 the most represented viruses (Furtak et al., 2016). Differences in virome profiling with
283 other studies might be due to the influence of library sequencing and the intrinsic
284 characteristics of the virome related to the sample itself and the area of study. On the
285 other hand, the higher presence of some viruses or even its detection only in effluent
286 samples could be produced by the decrease of other viruses that allowed its detection.

287 The presence of pathogenic viruses is an important aspect for defining the final use of
288 treated waters as it may be the case of irrigation. Due to their high environmental
289 resistance, the presence of human enteric viruses has been reported in treated
290 wastewaters (Adriaenssens et al., 2018). However, some of these pathogenic viruses are
291 not always detected by metagenomics analyses. For instance, in the study by Fernández-
292 Cassi et al. (2018), human adenoviruses (HAdV) reads were not detected in samples
293 concentrated from 10 liters of wastewater. *Adenoviridae* was also not detected in the
294 study of Adriaenssens et al. (2018), in which the sample was concentrated from 1 liter of
295 wastewater. In our study, concentrating 200 mL of effluent samples, we were able to
296 detect HAdV in percentages between 0.16% and 0.35%. In contrast, percentages of
297 HAdV in influent wastewaters were lower than 0.01%. Overall, the majority of the

298 annotated virome belonged to bacteriophages. This indicates that metagenomics is poor
299 in sensitivity when used to detect a low abundance of viral pathogens against a large
300 background of bacteriophages, as occurred for the enteric viruses detected by viability
301 RT-qPCR. For example, in the present study, norovirus genomes could not be retrieved
302 from the reads as reported elsewhere (Adriaenssens et al., 2018; Fernández-Cassi et al.,
303 2018; Strubbia et al., 2019b). In the current study, the number of generated paired reads
304 per sample was 3.2 and 11.5 million for LS and NB, respectively; while Adriaenssens et
305 al., (2018) reported between 10 and 110 million, increasing significantly the probability
306 to retrieve full or partial viral genomes. Alternatively, methods to detect and
307 characterize specific viruses have been described and rely on the selection of target
308 RNA prior to library preparation through a capture using VirCapSeq-VERT target
309 enrichment, as reported for norovirus (Strubbia et al., 2019b).

310 **5. Conclusion**

311 The use of metagenomics for virome characterization and its implementation for
312 wastewater analyses has increased in the last years (Nieuwenhuijse and Koopmans,
313 2017). However, the major problem of this approach is the lack of standardized
314 procedures and the substantial differences among studies; thus, available data must be
315 interpreted with caution. The present study showed a procedure that allows the detection
316 and the characterization of viral populations in untreated and treated wastewater
317 samples. Overall, this study sheds light on the diversity of the viral communities in
318 untreated and treated wastewaters providing valuable information also in terms of viral
319 fecal indicators. The study also evidences the bias on virome profiles obtained by tested
320 sequencing libraries. Our results underline the need for further studies to elucidate the
321 influence of sequencing procedures in virome profiles in wastewater matrices in order to
322 improve the knowledge of the virome in the environment.

323 **Declaration of competing interest**

324 The authors declare that they have no known competing financial interests or personal
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326 Names of specific vendors, manufacturers, or products are included for informational
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338 **Contributions**

339 GS designed the work. AP-C and EC-F processed the samples. AP-C performed the
340 bioinformatic work and data analysis. AP-C, EC-F, WR and GS wrote the paper. All
341 authors have read and agreed to the published version of the manuscript.

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Figure legends

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509 Figure 1: Shannon and Simpson diversity indexes for viral species in influent (IW) and
510 effluent (EW) samples processed by using ScriptSeq v2 RNA-Seq Library Preparation
511 kit (LS) and NEBNext® Ultra™ II RNA Library Prep Kit libraries (NB) for
512 metagenomics characterization.

513 Figure 2. Relative abundance at family level of the viral population detected in influent
514 and effluent samples from four different WWTPs by metagenomics with ScriptSeq v2
515 RNA-Seq Library Preparation kit (LS, green) and NEBNext® Ultra™ II RNA Library
516 Prep Kit (NB, orange).

517 Figure 3. Viral indicators analysis in influent (IW) and effluent (EW) samples. Panel A,
518 Detection of viral indicators with ScriptSeq v2 RNA-Seq Library Preparation kit (LS)
519 and NEBNext® Ultra™ II RNA Library Prep Kit (NB and NB-corrected). Panel B,
520 Correlation matrix between the reads of viral indicators obtained by NGS and the load
521 of enteric viruses (RT-qPCR) and E. coli counts.

522 Figure 4. Heatmap showing the virome composition at species level obtained by
523 merging the results of ScriptSeq v2 RNA-Seq Library Preparation kit (LS) and
524 NEBNext® Ultra™ II RNA Library Prep Kit (NB). Only species with percentages
525 higher than 1% are shown.

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