1	Supplementary Material
2	
3	Supplementary Methods
4	
5	Soil viral DNA extraction
6	Viral DNA was extracted from 50 g of fresh soil per sample using a previously reported protocol
7	[1] with minor modifications. For each sample, two 50 mL conical tubes were filled with 25 g of
8	soil and 37.5 mL of 0.02 μm filtered AKC' extraction buffer (per liter: 10% PBS, 10g K Citrate,
9	1.44 g Na ₂ HPO ₄ , 0.24g KH ₂ PO ₄ , and 36.97g MgSO ₄). Resulting slurries were vortexed briefly
10	until homogenized, shaken on an orbital shaker for 15 minutes at 400 RPM, vortexed for 3
11	minutes, and centrifuged at 4,700 x g for 15 minutes. Supernatant was filtered through a 0.22 μm
12	filter to remove most cells, and supernatants from tubes corresponding to the same sample were
13	combined into one 70 mL polycarbonate ultracentrifuge tube. Supernatants were centrifuged
14	using an Optima LE-80K ultracentrifuge (Beckman-Coulter Life Sciences, Indianapolis, IN, USA)
15	with a 45 Ti rotor at 32,000 x g for 3 hours at 4 °C.
16	Supernatant was decanted and the resulting pellets (the viral fraction), were resuspended in 200

Supernatant was decanted and the resulting pellets (the viral fraction), were resuspended in 200 µl of ultrapure water. Free DNA was removed from the resuspended pellets by treatment with 30 units of RQ1 RNase-free DNase and 30 µl of 10X DNase buffer (Promega Corp., Madison, WI, USA) for two hours at room temperature (22 °C). The reaction was quenched with 30 µl of DNase stop solution (Promega Corp., Madison, WI, USA).

DNA was extracted from the viral fraction using the DNeasy PowerSoil kit (Qiagen, Hilden, Germany), following the manufacturer's protocol, except the bead-beating step was replaced by a 10-minute 70 °C incubation, a 5-second vortex, and a 5-minute 70 °C incubation. Extracted DNA was quantified by an Invitrogen Qubit 4 Fluorometer using a 1x High Sensitivity DNA assay (Thermo Fisher Scientific, Inc., Waltham, MA, USA).

27 Soil total DNA extraction

Total DNA from soil was extracted from 0.5 g of soil using the DNeasy PowerSoil kit (Qiagen,
Hilden, Germany), following the manufacturer's instructions. Extracted DNA was quantified by an
Invitrogen Qubit 4 Fluorometer using a 1x High Sensitivity DNA assay (Thermo Fisher Scientific,
Inc., Waltham, MA, USA).

32

33 Read processing and assembly

Trimmomatic [2] was used to remove library adapters and quality-trim raw reads (minimum qscore of 30 evaluated on 4-base sliding windows; minimum read length of 50). BBDuk [3] was then used to remove PhiX sequences. *De novo* assembly on individual libraries was performed with MEGAHIT [4] in meta-large mode with a contig cutoff size of 2,000 bp. Assembly statistics were generated using the BBTools stats.sh script [3]. To remove redundant contigs across assemblies, we used the PSI-CD-HIT [5] implementation of BLASTN to cluster contigs at a global identity threshold of 0.95.

41

42 Detection and classification of viral contigs

43 We used VirSorter [6] and DeepVirFinder [7] to identify putative viral contigs. For VirSorter, only 44 contigs assigned to categories with the most confident (categories 1 and 4) or likely (categories 45 2 and 5) predictions were retained; for DeepVirFinder, only sequences with a score \geq 0.9 and pvalue < 0.05 were retained. The union of contigs identified by both methods was used for 46 47 downstream analyses. Viral contig identification was first performed on individual assemblies to 48 measure the enrichment of viral signal in the set of contigs found in each library. Viral contig 49 identification was also performed on the subset of clustered contigs with lengths \geq 10 Kbp to 50 generate a database of non-redundant sequences representing viral operational taxonomic units 51 (vOTUs). The length threshold and 95% sequence identity from prior clustering were based on 52 previous benchmarking and definitions of vOTUs [8, 9].

To assign taxonomic classifications to the recovered vOTUs, we first predicted protein content using Prodigal in metagenome mode [10]. The generated amino acid file was then used to build a gene-sharing network using vConTACT2[11] with the following parameters: NCBI RefSeq database of bacterial and archaeal viral genomes (v85) was used as a reference, Diamond [12] was used to calculate protein alignment, and MCL [13] and ClusterOne [14] algorithms were used to calculate protein and genome clusters, respectively.

59

60 Read mapping

61 Read mapping against the database of non-redundant vOTUs was performed with BBMap [3] at 62 a minimum sequence identity of 90%. Resulting SAM files were converted to BAM files and 63 indexed using SAMtools [15]. The parse function of BamM [16] was then used to generate two 64 vOTU tables: one displaying the trimmed pileup coverage (tpmean mode) and the other one 65 displaying the absolute number of mapped reads (counts mode). Finally, we calculated the per-66 sample horizontal coverage for each vOTU using the genomecov function in BEDtools [17]. We 67 then identified instances in which vOTUs displayed < 75% coverage over the length of the contig 68 and filtered them out of the vOTU tables using an in-house R script 69 (github.com/cmsantosm/SpatioTemporalViromes/blob/master/Processing/Scripts/votu filtering. 70 Rmd)

71

72 Detection and classification of 16S rRNA gene fragments

Reads containing 16S rRNA gene sequences were recovered using SortMeRNA [18] by comparing quality-filtered reads against representative versions of the bacterial and archaeal SILVA databases [19]. The RDP classifier [20] was then used to assign taxonomy. The output hierarchical file was further parsed using the hier2phyloseq function implemented in the RDPutils package [21] to generate a count table.

79 K-mer profiling

Sourmash [22, 23] was used to compute k-mer signatures for each library using a compression
ratio of 1,000 and k-mer size of 31.

82

83 Data analysis and visualization

84 All statistical analyses were conducted using R version 3.6.3. Unless otherwise stated, analyses 85 were performed using the trimmed pileup coverage vOTU table. The vegan package [24] was 86 used for the following analyses: accumulation curves were calculated using the specaccum 87 function, Bray-Curtis dissimilarity matrices were calculated on Hellinger-transformed relative 88 abundances using the vegdist function, permutational multivariate analyses of variance 89 (PERMANOVA) were performed with the adonis function, and Mantel tests were performed with 90 the mantel function. We performed two sets of PERMANOVA: one set testing the individual effects 91 of single factors (STable 4) and another set testing the effect of either biochar or nitrogen 92 treatments while accounting for differences due to collection time point and/or spatial structuring 93 along the W-E gradient (STable 5). In the PERMANOVA models, W-E and S-N positions were 94 coded as continuous variables. The function pcoa from the package ape [25] was used to perform 95 principal coordinate analyses. Hierarchical clustering on z-transformed values was performed 96 with the hclust function. Differential abundance analyses were performed with DESeg2 [26] using 97 count tables as input. The factors included in the design formula were those that showed a 98 significant effect in the overall composition analyses: collection time point, W-E gradient, and 99 biochar treatment to model vOTU abundances; and collection time point to model 16S OTU 100 abundances. All plots were generated with the ggplot2 [27], ggdendro [28], GGally [29], and 101 eulerr [30] packages. All scripts and intermediate files are available at 102 github.com/cmsantosm/SpatioTemporalViromes/

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170 Supplementary Figure 1

(A) Left panel: aerial view of the agricultural field; the highlighted area indicates the experimental 171 block used for this study. Picture was taken on July 10th, 2018, in the middle of the tomato growing 172 season (see B). Middle panel: diagram depicting the spatial distribution of plots within the block: 173 174 dots indicate the eight sampled plots with color and shape corresponding to the biochar treatment; 175 nitrogen fertilization treatment is indicated by the color of the individual plots (low - light gray, high - dark gray). Fertilization began between the two collection time points (see B). Right panel: 176 diagram depicting an individual plot. Dotted black lines represent the approximate area where 177 178 biochar was buried along each bed. Brown dots indicate the approximate location of the eight soil 179 cores harvested for each sample. (B) Pictures on the left show how biochar additions (up) and 180 August sample collection (down) were performed. Right: timeline of the growing season, biochar 181 and nitrogen amendments, and collection time points. Note that the distance between time points is not to scale. (C) Library construction workflows used to generate the total metagenomes (MG) 182 183 and viromes. The blank space represents a virome sample that failed at the library construction step, and the asterisk highlights the virome sample omitted from compositional analyses due to 184 185 suboptimal sequencing throughput and read mapping (see Figures 1A and D and Supplementary 186 Figure 3).





189 Supplementary Figure 2

(A) Relative abundances of microbial phyla in each library, based on recovered 16S rRNA gene 191 192 fragments with an assigned taxonomy. Only the top 10 most abundant phyla are shown. "Low 193 abundance" represents the summed relative abundances of all OTUs not assigned to the 10 most 194 abundant phyla. (B) Microbial phyla with significantly different relative abundances between total 195 metagenomes (MGs) and viromes (adjusted P < 0.05; paired Wilcoxon test). Red and blue bars 196 indicate phyla significantly enriched in total metagenomes and viromes, respectively. Higher 197 values along the x-axis indicate more significant enrichment. (C) Rarefaction curves of 16S rRNA 198 gene OTU profiles derived from total metagenomes. Each line represents one metagenome.



199

Number of Mapped Reads

200 Supplementary Figure 3

201 (A) Percent of high-quality reads mapped to the set of vOTU contigs identified in our dataset. Green colors highlight read mapping below (light green) or above (dark green) the threshold of 202 203 75% coverage over the length of the vOTU sequence required for detection. (B) Distribution of 204 the total number of reads mapped to vOTUs (x-axis) and the total number of vOTUs detected in 205 each sample (y-axis). Note different y-axis maxima between graphs. The circled virome sample 206 performed suboptimally (>2 standard deviations below the mean number of mapped reads and 207 mean richness across viromes) and was therefore discarded for downstream compositional 208 analyses. Total MG = total metagenome.



209

210 Supplementary Figure 4

211 Ranked relative abundances of individual vOTUs across profiling methods. Facets display

212 individual sets of paired total metagenomes (TMGs) and viromes (Vir) derived from the same

soil sample. The position of each vOTU along the y-axis indicates its rank within a profile, and a

214 link between two profiles indicates that the particular vOTU was found in both the total

215 metagenome and the virome from the same sample. Color denotes the detection category for

216 each vOTU, considering all viromes and total metagenomes (as in the Euler diagram in Figure

217 2C): blue indicates vOTUs exclusively detected in viromes, gray indicates vOTUs exclusively

218 detected in total metagenomes, and red indicates vOTUs found in both datasets.



219 Mean Rel. Abund. (log 220 Supplementary Figure 5

221 Abundance-occupancy curves of vOTUs in viromes (A), 16S OTUs in viromes (B), vOTUs in total 222 metagenomes (MGs) (C), and 16S rRNA gene OTUs in total metagenomes (D). In all panels, the 223 bottom left scatter plots represent the mean relative abundance (x-axis) and occupancy (percent 224 of samples in which a given vOTU was detected, y-axis) that individual vOTUs or 16S OTUs 225 displayed within a collection time point (April or August). Thus, vOTUs or 16S OTUs detected in both time points are represented twice. The top left density curves show the distribution of relative 226 227 abundances for all vOTUs or 16S OTUs. The bottom right bar plots display the percent of vOTUs 228 or 16S OTUs (x-axis) found at each occupancy level (y-axis).



230 Supplementary Figure 6

(A) Virus family and (B) predicted host phylum of vOTUs that were significantly enriched according
 to collection time point and that could be taxonomically classified (n = 92 vOTUs) (Fig 3). (C)
 Phylum or Proteobacteria class of the 16S rRNA gene OTUs that were significantly differentially
 abundant between collection time points. In all plots, color indicates the collection time point in
 which vOTUs or 16S rRNA gene OTUs were enriched.



Supplementary Figure 7.

Unconstrained analysis of principal coordinates based on vOTU Bray-Curtis dissimilarities calculated on Hellinger-transformed relative abundances across virome samples. Shapes represent the collection time point, and colors indicate (A) biochar treatment or (B) nitrogen fertilization treatment (note that April samples were collected before any nitrogen was added to the field)



243244 Supplementary Figure 8.

(A) Unconstrained analyses of principal coordinates based on Bray-Curtis (B-C) dissimilarities
 calculated on Hellinger-transformed relative abundances across 16S rRNA gene OTU profiles
 derived from total metagenomes (MGs). (B) Correlation between spatial distance across the west east axis (in meters between plots) and Bray-Curtis dissimilarities calculated across 16S rRNA
 gene OTU profiles derived from total metagenomes. Inset values display the Mantel r statistic and
 associated P-value. Bray-Curtis dissimilarities were calculated on Hellinger-transformed relative
 abundances.



252 253 Supplementary Figure 9.

254 (A) Partial canonical analysis of principal coordinates (CAP) performed on Bray-Curtis 255 dissimilarities calculated on vOTU profiles from viromes. The effects of collection time point and 256 W-E position were removed. Colors indicate biochar treatment, and shape indicates sampling 257 time point (legends are to the right of panel B). (B) Hierarchical clustering of biochar treatments 258 based on the relative abundances of vOTUs significantly affected by biochar amendments. The 259 heatmap shows the mean relative abundance (z-transformed) of each vOTU (rows) across biochar treatments (columns). (C) Partial CAP performed on Bray-Curtis dissimilarities calculated 260 261 on 16S rRNA gene OTU profiles from total metagenomes. For A and B, Bray-Curtis dissimilarities 262 were calculated on Hellinger-transformed relative abundances. The effect of collection time point 263 was removed.

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- 266 267

Supplementary Table Legends

268 269 Supplementary Table 1

- 270 Properties of biochar amended to agricultural field
- 271

272 Supplementary Table 2

273 Sequencing depths (before and after quality filtering) and sample metadata for all libraries 274 reported in this study.

275

276 Supplementary Table 3

277 Viral cluster assignment for RefSeq genomes and vOTU contigs278

279 Supplementary Table 4

Permutational multivariate analyses of variance testing the effect of individual variables on community composition. Analyses were performed on Bray-Curtis dissimilarities calculated on Hellinger-transformed relative abundances across vOTU profiles in viromes and 16S rRNA gene OTU profiles in total metagenomes (Total MG). The effect of nitrogen concentration was only tested in the August subset of samples, after the fertigation had occurred. The last column indicates the formula used to run the test.

287 Supplementary Table 5

Permutational multivariate analyses of variance testing the effect of biochar or nitrogen treatments while controlling for the variation due to collection time point and/or W-E spatial gradient. Analyses were performed on Bray-Curtis dissimilarities calculated on Hellinger-transformed relative abundances across vOTU profiles in viromes and 16S rRNA gene OTU profiles in total metagenomes (Total MG). The effect of nitrogen concentration was only tested in the August subset of samples, after the fertigation had occurred. The last column indicates the formula used to run the test.

295

296 Supplementary Table 6

Set of vOTUs differentially abundant across collection time points (Wald test, adjusted P-val <
0.05). The first 7 columns are the default output from DESeq2. Column "VC" indicates the viral
cluster assigned to each vOTU by vConTACT2.

300

301 Supplementary Table 7

302 Set of 16S rRNA gene OTUs differentially abundant across collection time points (Wald test, 303 adjusted P-val < 0.05). The first 7 columns are the default output from DESeq2.

304

305 Supplementary Table 8

ANOVA test of the effects of collection time point, biochar, nitrogen amendment concentration,
 W-E position, and S-N position on the measured chemical properties of soil. The effect of nitrogen
 amendment concentration was only tested for the August samples.

309

310 Supplementary Table 9

Set of vOTUs significantly affected by plot position along the west-east axis of the sampled field (Wald test, adjusted P-val < 0.05). The first 7 columns are the default output from DESeq2.

- Column "VC" indicates the viral cluster assigned to each vOTU by vConTACT2.
- 314

315 Supplementary Table 10

- 317 Set of vOTUs differentially abundant across biochar treatments (likelihood ratio test, adjusted P-val < 0.05). Column "VC" indicates the viral cluster assigned to each vOTU by vConTACT2.