A catalog of 48,425 nonredundant viruses from oral metagenomes expands the horizon of the human oral virome

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A catalog of 48,425 nonredundant viruses from oral metagenomes expands the horizon of the human oral virome
A catalog of 48,425 nonredundant viruses from oral metagenomes expands the horizon of the human oral virome

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SUMMARY
The human oral cavity is a hotspot of numerous, mostly unexplored, viruses that are important for maintaining oral health and microbiome homeostasis. Here, we analyzed 2,792 publicly available oral metagenomes and proposed the Oral Virus Database (OVD) comprising 48,425 nonredundant viral genomes (≥ 5 kbp). The OVD catalog substantially expanded the known phylogenetic diversity and host specificity of oral viruses, allowing for enhanced delineation of some under-represented groups such as the predicted Saccharibacteria phages and jumbo viruses. Comparisons of the viral diversity and abundance of different oral cavity habitats suggested strong niche specialization of viromes within individuals. The virome variations in relation to host geography and properties were further uncovered, especially the age-dependent viral compositional signatures in saliva. Overall, the viral genome catalog describes the architecture and variability of the human oral virome, while offering new resources and insights for current and future studies.

INTRODUCTION
Viruses are perhaps the most abundant and diverse organisms on the Earth (Brüssow and Hendrix, 2002; Dion et al., 2020; Paez-Espino et al., 2016) and can contribute to human health and diseases (Jakobsen et al., 2020; Nakatsu et al., 2016; Norman et al., 2015). With the development of next-generation sequencing technologies and virus identification tools, previous studies suggested that the types of viruses inhabiting in humans were far more than the known viruses provided by the Reference Sequence (RefSeq) collection (Clooney et al., 2019; Gregory et al., 2020; Shkoporov et al., 2019), which had limited the exploration of the association between the virus and human health. Thus, expanding the genome resources of human-associated viruses is a key issue for virologists. For example, Soto-Perez et al. (2019) established the Human Virome Database (HuVirDB) via integrating virome samples from multiple body sites (including blood, stool, lung, and skin) and Tisza et al. (Tisza and Buck, 2021) developed the Cenote Human Virome Database (CHVD) from public metagenomic samples spanning gut, mouth, nose, skin, and vagina. Three recent studies had performed large-scale virus identification from public available fecal metagenomes and constructed the Gut Virome Database (GVD) (Gregory et al., 2020), Gut Phage Database (GPD) (Camarillo-Guerrero et al., 2021), and Metagenomic Gut Virus (MGV) catalog (Nayfach et al., 2021). These resources revealed the massive viral diversity in humans and would facilitate the exploration of the viral characteristics and host-virus interaction.

The importance of human oral virome has attracted increasing attention over the past few decades. Some ubiquitous viruses such as the Epstein-Barr virus (EBV) that infect in the oral cavity can probably contribute to oral cancers (Guidry et al., 2018), and the alterations of the oral viral community have been linked to a wide range of oral diseases such as periodontal disease (Ly et al., 2014), lichen planus (Carrozzi, 2008), and hand, foot, and mouth disease (Ho et al., 2021). Also, our previous study had shown that the diversity and structure of saliva and dental plaque viromes were distinctly changed in patients with rheumatoid arthritis and were correlated with their therapeutic plans (Guo et al., 2021b). In addition, longitudinal analysis had revealed that the oral virome of healthy subjects was highly diverse, individually specific, and temporally stable (Abeles et al., 2014; Pride et al., 2012). Because the understanding of oral virome is increasing rapidly, however, only a few studies explored the genome resources of the oral viruses. The...
CHVD database contained over 45,000 viral genomes from nearly 6,000 metagenomic samples from various body sites (Tisza and Buck, 2021), including 13,565 viruses that were assembled from 1,287 oral samples, which developed into the current largest viral genome reference of the human oral cavity. Meanwhile, a recent study analyzed four saliva samples using deep short-read and long-read metagenomic sequencing and generated hundreds of viral genomes, including high proportions of Streptococcus phages and jumbo phages (Yahara et al., 2021). Collectively, these efforts raised the requirement for expanding the genome resources and provided a systematic insight into the oral virome.

In this study, we analyzed 2,792 publicly available human oral metagenomic samples covering 12 oral cavity sites and nine countries to build a complementary genome catalog of oral virome. The catalog was named oral virome database (OVD) including 48,425 nonredundant viruses (de-replicated from 121,054 viruses with >95% nucleotide similarity) with a majority never found in existing databases. Using OVD, we gained preliminary insights into the taxonomic and host ranges, spatial diversity, and geographic and individual heterogeneity of the human oral virome.

RESULTS
Construction of a global human oral virus catalog
To address the material for oral virome research, we collected a total of 2,792 publicly metagenomic data originating from the human mouth and established the largest oral metagenomic dataset so far, spanning 16 studies from the USA, China, 5 European countries (France, Germany, Italy, Luxembourg, and Spain), and two Oceanian countries (Australia and Fiji) (Figure 1A; Table S1). This dataset contained data of samples from 12 oral cavity sites, including dental plaque (895 samples), saliva (696 samples), tongue (439 samples), buccal mucosa (356 samples), and others (406 samples). These samples represented 9.3 Tb high-quality non-human metagenomic data after being processed with a unified pipeline, and further generated a total of 5.7 million long contigs (≥5 kb; total length 73.3 Gb) via metagenomic assembly for each sample (Table S2). Using an integrated homology- and feature-based pipeline (see STAR Methods), approximately 2.1% (n = 121,054) of the contigs were detected as highly credible viral sequences. These viruses were finally clustered at >95% nucleotide similarity (Gregory et al., 2019) to generate a nonredundant oral virome database (OVD) with 48,425 viral operational taxonomic units (vOTUs). The length of vOTUs ranged from 5,004 bp to 433,840 bp, with an average length of 29,800 bp and an N50 length of 37,664 bp (Figure S1). We evaluated the completeness and contamination of the OVD catalog using CheckV (Nayfach et al., 2020), revealing 7.0% complete, 15.5% high-, 21.0% medium-, and 56.4% low-completeness viruses, and 99.5% of all these vOTUs were low contaminated (Figure 1B). 22.6% (n = 10,931) of the vOTUs with estimated high completeness and low contamination were classified as high-quality viruses (Table S3).

An accumulation curve for vOTUs was not yet reaching a plateau, indicating that the oral virome had not been fully captured by OVD (Figure 1C). We clustered the vOTUs into 3,572 approximately genus-level groups and 529 approximately family-level groups based on their protein similarity and gene sharing proportion following a previously reported method (Nayfach et al., 2021). Accumulation curve analysis showed that the OVD catalog appears to be approaching an asymptote at the approximate genus and family ranks.

We compared the vOTUs of OVD with several existing virus catalogs, including three human gut virus catalogs (i.e., GVD, GPD, and MGV), the oral viruses from CHVD, and the available viral genomes in the RefSeq database. Viral quality estimation based on CheckV as well as three other tools suggested that the confidence of vOTUs in OVD was comparable with the other catalogs (Figure S2). Remarkably, 74.6% of the CHVD oral viruses were found in OVD, which accounted for 20.9% of all OVD vOTUs (Figure 1D). Conversely, despite that the gut virus catalogs were constructed from a huge number of fecal metagenomes and viromes and contained more nonredundant viruses than the OVD, only less than 11% of OVD viruses were overlapped with them, suggesting a considerable habitat specificity of the human-associated virome. Moreover, the RefSeq viruses rarely overlapped with both gut and oral catalogs. These findings yielded that approximately 70% (33,608/48,425) of oral vOTUs were novel and thus highlighted the considerable novelty of human oral virome.

Taxonomic landscape and host range of oral viruses
Referring to a gene-sharing pipeline (Bin Jang et al., 2019) with updated reference, 45.8% (22,194/48,425) of vOTUs of the OVD catalog were taxonomically assigned into the viral families. These vOTUs were dominated by three viral families that belonged to the Caudovirales order, including Siphoviridae, Myoviridae,
and Podoviridae (Figure S3A), which were also the most dominant taxa in the human gut virome catalogs (Zuo et al., 2020). And members of the crAss-like phages (189 vOTUs), Autographiviridae (137 vOTUs), Quimbyviridae (120 vOTUs) (a putative viral family defined by Benler et al.’s study (Benler et al., 2021)), and Microviridae (23 vOTUs) frequently appeared in the remaining vOTUs. We performed host prediction of the vOTUs based on their homology of genome sequences (≥90% nucleotide identity and ≥30% viral
coverage) or CRISPR spacers (bit-score ≥45) to a large-scale oral prokaryotic genome collection (representing 3,569 species from over 50,000 metagenome-assembled genomes) (Zhu et al., 2021). This process predicted the prokaryotic hosts of 67.3% (32,570/48,425) of the vOTUs in OVD (Figure S3B). At the phylum level, the predicted hosts of oral viruses were dominated by Firmicutes, Actinobacteria, Proteobacteria, and Bacteroidota, followed by Fusobacteria, Saccharibacteria, and Campylobacterota, which were previously reported to be enriched in the human oral cavity (Human Microbiome Project, 2012; Zhu et al., 2021).

Strikingly, 12.4% of the vOTUs were predicted to infect hosts belonged to more than one prokaryotic phylum, which was considerably larger than the proportion (6.7%) in gut viruses (Fisher’s exact test, p < 0.001; Figure S3B), suggesting a relatively broad host range of most oral viruses.

According to genome-wide similarity at the protein level (Nishimura et al., 2017), we constructed a proteomic tree of 10,931 high-quality vOTUs, of which 56.3% could be assigned to a known viral family and 78.0% could be predicted to known prokaryotic hosts (Tables S3 and S4). The tree roughly showed that the viruses tend to cluster by both family-level taxonomies and potential host affiliations (Figure 2A), and a similar finding was also observed in multivariate analysis of the protein profiles of vOTUs (Figure S4). These findings suggested that the host adaptation had an important driving effect on genomic evolution of oral viruses, to a great extent in agreement with previous reports (Remold et al., 2008; Simmonds et al., 2019).

The vOTUs of the largest family, Siphoviridae, were predicted to infect some of the most dominant bacteria in the human oral microbiota (Human Microbiome Project, 2012), including Streptococcaceae and Actinomycetaceae (Figures 2B and S5A). The Myoviridae viruses tended to infect Proteobacteria members, especially the Pasteurellaceae and Neisseriaceae species (Figure S5B), and the Podoviridae viruses were mostly grouped into a clade in the proteomic tree and part of them were predicted to infect the species of Streptococcaceae and Actinomycetaceae. For other viruses, the Bacteroidaceae species were dominant hosts of viruses belonging to crAss-like viruses, Microviridae, Flandersviridae, Gratiairidae, and Quimbyviridae. Strikingly, we noticed that the vOTUs predicted to infect Saccharibacteria (formerly known as TM7 which is ubiquitously distributed in the human oral cavity (Ferrari et al., 2014; He et al., 2015)) were mostly clustered in a single clade in the proteomic tree (Figure 2A). These predicted Saccharibacteria phages (n = 344; Table S5) were moderate in genome size (average 35.5 kbp) but highly prevalent in oral metagenomes across all samples (average prevalence rate, 25.9%) (Figures S6A and S6B). The taxa of these predicted Saccharibacteria phages were currently unknown, but the aforementioned clustering analysis could group them into two major family-level groups and seven major genus-level groups with remarkable stratification of bacterial hosts at the species level (Figure 3A). In particular, the seven genus-level groups showed large differences in their abundances (represented as the “abundance of reads”, see STAR Methods) in samples from different countries and oral cavity sites (Figures S6C and S6D), highlighting a high geographic and spatial heterogeneity of them. Collectively, these results were important for understanding the host-virus interactions in the oral microbiome and will promote future studies in evolution and/or epidemiology scopes.

Recent studies have shown that the human oral virome is rich in jumbo viruses (Carr et al., 2020), a type of tailed bacteriophages with genome sizes more than 200 kbp. The OVD catalog contained 391 high-quality jumbo vOTUs with a high prevalence across all regions and oral sites (Figures S7A and S7B, Table S6). Similar to the Saccharibacteria phages, no jumbo vOTUs could be assigned into known viral families based on our current method/database. Instead, clustering analysis grouped the jumbo viruses into eight major approximately family-level groups that obviously differed in host predictions (Figures 3B and S8). Comparison of the annotatable functions revealed that, as expected, the jumbo viruses encoded a significantly higher proportion of auxiliary metabolic genes (AMGs) than the non-jumbo viruses (Fisher’s exact test, p = 0.04; Figures 3C and S7C), suggesting a high metabolic potential of these viruses. Besides, 91.6% of jumbo vOTUs were lytic viruses, which is remarkably higher than that of non-jumbo viruses (Fisher’s exact test, p < 1e-10; Figure 3D).

**Spatial diversity, structure, and function of the oral virome**

The oral cavity is a heterogeneous ecosystem containing distinct niches with remarkably different microbial communities (Xu et al., 2015). To explore the spatial specificity of oral virome, we undertook a compositional comparison of the vOTU profiles of four major oral cavity sites (i.e., buccal mucosa, dental plaque, saliva, and tongue), spanning a total of 1,909 deeply sequenced metagenomes (>5 million clean reads) for analysis. In terms of diversity, the salivary virome had the highest viral richness and diversity than other sites, followed by the tongue virome, buccal mucosa, and dental plaque (Wilcoxon rank-sum test, p < 0.01;...
**Figure 2.** Phylogenomic analysis of high-quality vOTUs in OVD

(A) A proteomic tree of 10,931 high-quality vOTUs. The tree was generated using ViPTreeGen (Nishimura et al., 2017). Outer rings display metadata for each vOTU: innermost ring, viral family-level taxonomic assignments; ring 2, phylum-level host assignments; ring 3: lysogenic or lytic types; and outermost ring, the sequence length of the vOTUs. See Table S3 for details.

(B) Distribution of prokaryotic hosts of high-quality vOTUs. The vOTUs are grouped at the family level, and the host taxa are shown at the family level. The number of vOTUs that had more than one predicted host is labeled by light blue color.

**Figure 4A.** Analysis at the family level showed that several dominant viral families, such as *Siphoviridae* and *Myoviridae*, are universally present among all sites with similar distribution patterns (Figure S9), although almost all (20/21) families had statistically differences among four sites (Kruskal-Wallis rank-sum test, adjusted p < 0.01; Table S6). In addition, principal coordinates analysis (PCoA) based on the Bray-Curtis distance of vOTU profiles revealed a clear separation of viromes among different sites (Figure 4B).

Comparison at the vOTU level revealed that the majority of (57.7%) of 48,425 vOTUs were statistically enriched in the virome of one of the four oral cavity sites (Wilcoxon rank-sum test, adjusted p < 0.01; Table S8), suggesting...
strong niche specialization of viromes within individuals. The buccal mucosa-enriched vOTUs were more frequently concentrated in Siphoviridae compared with other sites, while the saliva-enriched vOTUs had more jumbo viruses, crAss-like viruses, and Quimbyviridae compared with other sites (Figure 4C). Of the enriched vOTUs with a predicted host, most of the buccal mucosa-enriched and tongue-enriched vOTUs were predicted to infect members of Firmicutes, while the dental plaque-enriched vOTUs had the highest proportions of predicted Bacteroidota, Fusobacteriota, and Saccharibacteria phages compared with other sites. Moreover, we compared the functional capacities of the site-associated vOTUs to elucidate the spatial specificity of oral virome functions. Although some basic viral functions (e.g., integrase, phage terminase, and ssDNA-binding protein) were ubiquitous across all oral cavity sites, there were numerous important functions that significantly differed in frequency among the enriched viruses of four sites (Figures 4D, S10; Table S9). For example, the saliva-enriched vOTUs had a higher frequency of two enzymes, dihydrofolate reductase (K00287) and thymidylate synthase (K00560), that were related to folate metabolism, the dental plaque-enriched vOTUs had a higher frequency for DNA (cytosine-5)-methyltransferase 1 (K00558, involving DNA methylation) and phosphoadenosine phosphosulfate reductase (K00390, involving sulfur metabolism), and two lysozymes (K01185 and K07273) were more frequent in the buccal mucosa-enriched vOTUs.

Geography and host properties describe the salivary and dental viromes

Finally, to investigate the variations of the oral virome in geography and host properties (i.e., gender, age, and body mass index [BMI]), we performed comparative analyses of vOTU profiles in two oral cavity sites.
with the largest sample size, saliva (containing 490 samples from six counties), and dental plaque (containing 772 samples from four counties). For the salivary virome, multivariate analyses revealed that three factors, including geography, age, and BMI, had considerably impacted the holistic viral profiles (permutational multivariate analysis of variance [PERMANOVA] \( p < 0.001 \) for all), whereas the effect size of gender was comparably slight (PERMANOVA \( p = 0.023 \); Figures 5A and 5C). The salivary viral compositions of American and European individuals were close in tendency and largely separated from that of Fiji individuals (Figures 5A and 5B). These geography-dependent variations may partly be explained by study bias but also potentially connected to the population differences in genetic background, lifestyle, or urbanization, as observed in the gut virome studies (Yan et al., 2021; Zuo et al., 2020). The viral richness and diversity significantly increase from children to adults and decrease from adults to elders, and the highest viral richness and diversity occurred at the ages between 20 and 30 years (Figure 5D). Several viral families such as Myoviridae, Podoviridae, and Retroviridae were enriched in the salivary virome of children compared with those of adults and elders, whereas the crAss-like viruses and Quimbyviridae were

Figure 4. Comparison of viromes among four oral cavity sites
(A) Boxplots show the viral richness (the number of observed vOTUs, bottom panel) and diversity (Shannon’s index, upper panel) of four sites. Wilcoxon rank-sum test was implemented between two groups.
(B) Principal coordinates analysis (PCoA) based on the Bray-Curtis distance of vOTU profiles. Samples are shown at the first and second principal coordinates (PC1 and PC2), and the ratio of variance contributed by these two PCs is shown. Ellipsoids represent an 80% confidence interval surrounding each group. (C) Barplots show the proportions of site-enriched viruses grouping by their family-level taxa (left panel) and phylum-level host predictions (right panel). See Table S8 for details.
(D) Heatmap shows the occurrence rates of functions that exhibited as the top 50 functions in the vOTUs that significantly enriched in four oral sites. “+” represents the function with the highest occurrence rate in the corresponding vOTU groups. Occurrence rate represents the ratio of number of site-enriched viruses with corresponding KO to the total number of site-enriched viruses. Bold font shows several enzymes that are described as examples in Results. The comparison result of these functions is shown in Table S9.
depleted in the children (Figure S11). Parallely, we found that the viruses that infect Fusobacteriota, Campylobacterota, and Euryarchaeota decreased continuously with age, as well as the Proteobacteria phages in subjects under 40, whereas the Firmicutes phages increased continuously in subjects under 40 (Figure S12). Phages of Actinobacteriota and Saccharibacteria were enriched in adults aged 40 to 60. Besides, the individuals with BMI <18 appeared to have a markedly lower level of viral richness and diversity compared with the others (Figure 5E), but this phenomenon was not significant after adjusting for age and geography (p = 0.09).

For the dental plaque virome, multivariate analyses showed that the geography, gender, and age factors significantly impacted the variations of viral compositions (PERMANOVA p < 0.001 for geography and age and p = 0.002 for gender; Figure 6A). The viral richness and diversity of China and Australian individuals were remarkably higher than those of American and Italian individuals (Figure 6B), although the overall dental virome of American and Australian individuals was close. Females had a higher level of viral richness and diversity in their dental plaque than males, and middle-aged individuals had higher viral richness (not diversity) than the children and elders (Figures 6C and 6D). Besides, both viral composition and diversity seemed not to be correlated with the host’s BMI (Figures 6A and 6E).

DISCUSSION
In this study, we reported a comprehensive catalog of 48,425 nonredundant viral sequences deriving from 2,792 public oral metagenomic samples across nine countries. The high coverage of the resulting viral clades at the approximate genus and family levels demonstrated the value of viral identification from the oral metagenomes. More importantly, nearly 70% of the viruses in the OVD catalog were newly found
compared with the existing human virus catalogs or RefSeq viral genome reference, highlighting its value in exploring the previously uncharacterized “dark matter” of oral virus.

Taxonomic and host annotations of OVD described the phyllogenetic view of the oral virome. At the viral family level, the known oral vOTUs were dominated by members of Siphoviridae and Myoviridae, and Podoviridae, highly similar with the composition of gut virome (Zuo et al., 2020). Several recently discovered viral taxa that are widespread in the human gut, such as crAss-like viruses (Edwards et al., 2019; Shkoporov et al., 2018) and Quimbyviridae (Benler et al., 2021), were also frequently present in OVD. In addition, OVD contained some important viral groups that are underrepresented in the gut microbiota, including the predicted phages of Fusobacteriota and Saccharibacteria. Fusobacteria are well-established opportunistic pathogens of oral or colorectal cancers (Abed et al., 2020; Al-Hebshi et al., 2017; Harrandah et al., 2020; Kostic et al., 2012), and its phages might be also potential participants in the etiology of colorectal cancer (Nakatsu et al., 2018) (manuscript in preparation). Saccharibacteria are ubiquitous in oral microbiota and have been linked to multiple diseases (Bor et al., 2019; Kuehbacher et al., 2008), but their phages are uninvestigated. Here, we identified 176 high-quality vOTUs predicted to infect Fusobacteriota and 344 high-quality vOTUs predicted to infect Saccharibacteria, largely expanding the genome contents of these viruses. Besides, we identified 391 high-quality jumbo vOTUs from oral metagenomes and found that they were enriched in metabolism-associated genes and lytic viruses, probably linking to their nutrient availability conditions and pathogen elimination functions in the human body (Silveira and Rohwer, 2016).

We profiled the viral compositions of metagenomic samples and identified tremendous signatures in relation to spatial distribution, geographic, and individual heterogeneity of the oral virome; these results will be useful for experimental design (e.g., choosing of sampling sites) and mechanistic research in the future. Noticeably, we found that both salivary and dental plaque viromes were shaped by the individuals’ age,
with rapidly increasing viral richness from child to adulthood and then slow decrease in old age. This age-dependent trajectory was similar to the observations in human gut virome (Gregory et al., 2020). Although the effect of virome dynamics needs further exploration, our results suggested that, in addition to the oral bacteria (Huang et al., 2020; Lira-Junior et al., 2018), the oral viruses may also correlate with maturation and aging of human beings.

In conclusion, the OVD catalog of over 48,000 viral genomes will largely improve further exploration of human oral virome both in depth and breadth.

Limitations of the study
A major limitation of the current OVD is the challenge of viral completeness (77.4% of viruses had <90% completeness) and the recovery of low-abundance viruses. State-of-the-art virome analysis strategies such as long-read metagenomic sequencing or virus-like particle (VLP) sequencing have been successfully used in uncovering the viral genome of fecal samples (Wang et al., 2021; Yahara et al., 2021) and are promising to obtain a more comprehensive landscape of the oral virome. More samples are still in need to address the representation of regions and oral cavity sites. Moreover, although we had provided preliminary information about the correlations between oral virome and host properties, the alteration of oral virome and its association with disease still need to be clarified.

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Supplemental information can be found online at https://doi.org/10.1016/j.isci.2022.104418.

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AUTHOR CONTRIBUTIONS
S.L., R.G., and Q.Y. conceived and directed the study. Y.Z. performed data collection and investigation. S.L., R.G., Y.Z., F.C., Q.L., and G.W. carried out data processing and analyses. S.L. and R.G. drafted the manuscript. Z.J. and H.J. participated in design and coordination. P.L., X.W., and J.L. revised the manuscript. All authors read and approved the final manuscript.

DECLARATION OF INTERESTS
The authors declare no competing interests.


## STAR METHODS

### KEY RESOURCES TABLE

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<td>Shi et al., 2020 sequencing reads</td>
<td>(Shi et al., 2020)</td>
<td>NCBI BioSample - see Table S2 for details</td>
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<tr>
<td>Velsko et al., 2019 sequencing reads</td>
<td>(Velsko et al., 2019)</td>
<td>NCBI BioSample - see Table S2 for details</td>
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<td>Wang et al., 2019 sequencing reads</td>
<td>(Wang et al., 2019)</td>
<td>NCBI BioSample - see Table S2 for details</td>
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<td>Wei et al., 2021 sequencing reads</td>
<td>(Wei et al., 2021)</td>
<td>NCBI BioSample - see Table S2 for details</td>
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<td>Zeller et al., 2014 sequencing reads</td>
<td>(Zeller et al., 2014)</td>
<td>NCBI BioSample - see Table S2 for details</td>
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<tr>
<td>Zhang et al., 2015 sequencing reads</td>
<td>(Zhang et al., 2015)</td>
<td>NCBI BioSample - see Table S2 for details</td>
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</table>

### Software and algorithms

| fastp v0.20.1 | (Chen et al., 2018) | http://opengene.org/fastp/fastp.v0.20.1 |
| bowtie2 v2.4.1 | (Langmead and Salzberg, 2012) | https://github.com/BenLangmead/bowtie2 |

(Continued on next page)
RESOURCE AVAILABILITY

Lead contact
Further information and requests for resources and reagents should be directed to and will be fulfilled by the Lead Contact, Shenghui Li (lsh2@qq.com).

Materials availability
This study did not generate new unique reagents.

Data and code availability
- The viral sequences and annotation files of OVD have been deposited in the GitHub website with URL: https://github.com/RChGO/OVD/.
- The original codes used in the paper are provided in the following link: https://github.com/RChGO/virusDetect.
- The metadata of all 2,792 oral metagenomic samples, including NCBI BioSample ID, oral cavity site, host properties (geography, gender, age, and BMI), and data and assembly information, are available in Table S2. All other data reported in this paper will be shared by the lead contact upon request.

EXPERIMENTAL MODEL AND SUBJECT DETAILS

Human oral samples
We reviewed multiple studies based on whole-metagenome sequencing of oral samples and collected a total of 2,792 oral metagenomic samples publicly available in the NCBI (https://www.ncbi.nlm.nih.gov/) database until May 2021. The metadata of all selected samples were summarized and listed in Tables S1 and S2.

METHOD DETAILS

Preprocessing and assembly
Raw reads were qualified via fastp v0.20.1 (Chen et al., 2018) with the options ‘-u 30 -q 20 -l 90 -y -trim_poly_g’, and human reads were further removed by matching quality-filtered reads against the human genome GRCh38 with bowtie2 v2.4.1 (Langmead and Salzberg, 2012). The remaining clean reads of each sample were assembled into contigs using SPAdes v3.14.1 (Nurk et al., 2017) with the options ‘-k 21,33,55’ (for samples with read length <100 bp or less) or ‘-k 21,33,55,77’ (for samples with read length >100 bp). Total 5,670,086 contigs with ≥5 kbp sequence length (average length 12,934 bp, total length 73.3 Gbp) were used to further analyze.
**Viral identification and decontamination**

All assembled contigs (n = 5,670,086, length ≥ 5 kbp) were firstly assessed by CheckV v0.7.0 (Nayfach et al., 2020), and then contigs would be removed if their microbial genes are no less than 10 and five times the number of viral genes, resulting in a total of 310,627 contigs for further analyses. We identified potential viral sequences from the remaining contigs based on any of the following criteria: 1) contig whose viral genes was more than the number of microbial genes in CheckV (contigs with length <10 kbp and estimated as low-quality or undetermined contigs were removed); 2) contig with p-value <0.01 and score >0.90 in DeepVirFinder v1.0; 3) contig identified by VIBRANT v1.2.1 (Kieft et al., 2020) with default options. A total of 127,298 contigs were recognized as proposed viral sequences, including 107,424 identified by CheckV, 40,796 identified by DeepVirFinder, and 88,620 identified by VIBRANT. To decontaminate the viral sequences, according to the previous study (Gregory et al., 2020), we searched the bacterial universal single-copy orthologs (BUSCO) (Manni et al., 2021) within viral sequence using hmmsearch (Eddy, 2011) with default options and calculated the ratio of the number of BUSCO to the total number of genes in each viral sequence (BUSCO ratio). Then we removed high-contaminated viral sequences with ≥ 5% BUSCO ratio, and the remaining contigs (n = 121,054) were considered as the final viral sequences for each sample.

**Viral clustering and gene calling**

The viral sequences were de-replicated based on the following steps: 1) all viral sequences (n = 121,054) were aligned in pairs using BLASTn v2.9.0 (Camacho et al., 2009) with the options ‘-evalue 1e-10 -word_size 20 -num_alignments 99999’. 2) viral sequences which shared 95% nucleotide identity across 75% of the sequence were clustered into a viral operational taxonomic unit (vOTU) using the custom scripts (https://github.com/RChGO/virusDectect). 3) For each vOTU, the longest viral sequence was considered as the representative sequence and used for further analyses. Total 48,425 vOTU sequences clustered from all oral samples were integrated into the oral virome database (OVD). In addition, the shared vOTUs between different virome databases were identified based on the same steps as above.

We further clustered 48,425 vOTUs of OVD into approximately genus-level and family-level groups respectively. Firstly, totaling 1,846,359 putative protein sequences in vOTUs were called using Prodigal v2.6.3 (Hyatt et al., 2010) with options ‘-p meta’, and pairwise protein sequence alignments were performed using DIAMOND v2.0.6.144 (Buchfink et al., 2015) with the options ‘-e 1e-5 –max-target-seqs 99999’. Then we calculated the percentage of shared gene and average amino acid identity (AAI) between each pair of vOTUs. According to the previous study (Nayfach et al., 2021), at the family level, we kept the connections between vOTUs with >20% AAI and >10% genes shared. At the genus level, we kept the connections between vOTUs with >50% AAI and >20% genes shared. Finally, clustering was performed based on connections between vOTUs using MCL with the option ‘-I 1.2’ for the family level or ‘-I 2’ for the genus level. All vOTUs in OVD represented approximately 529 family- and 3,572 genus-level groups.

**Viral taxonomy**

Taxonomic annotation of viral sequences was performed based on protein sequence alignment to the combined database derived from Virus-Host DB (Mihara et al., 2016) downloaded in May 2021, crAss-like protein sequences from Guerin’s study (Guerin et al., 2018) and viral protein sequences from Benler’s study (Benler et al., 2021). To implement accurate family-level taxonomic classification, we firstly aligned proteins of viral sequences from NCBI RefSeq against the combined database using DIAMOND with the options ‘--query-cover 50 --subject-cover 50 --id 30 --min-score 50 --max-target-seqs 10’. A viral sequence was annotated to the viral family level when over a quarter of its proteins were matched to the same family.

**Host prediction**

The oral microbial genome catalogue was derived from 3,569 prokaryotic species (containing over 50,000 metagenome-assembled genomes) from a previous study (Zhu et al., 2021). Based on the catalogue, the virus-host prediction was performed using two bioinformatic methods that included CRISPR-spacer matches and prophage blasts. For CRISPR-spacer matches, we firstly predicted CRISPR spacer sequences from the oral prokaryotic genome catalogue using MinCED v0.4.2 (Bland et al., 2007) with the option ‘-minNR 2’, and then assigned a host to the virus if host CRISPR spacer sequence was
matched to the viral genome (bit-score \geq 45) using BLASTn with options `-evalue 1e-5 -word_size 8 -num_alignments 99999` (Gregory et al., 2020). For prophage blasts, the viral sequence was blasted against host genome sequences, and assigned a host if the viral sequence was exactly matched to the host genome at \geq 90% nucleotide identity and \geq 30% viral coverage (Gregory et al., 2020). For 48,425 vOTUs in OVD, 57.0% (27,622/48,425) could be predicted to hosts by CRISPR-spacer matches, and 32.5% (15,737/48,425) could be predicted by prophage blasts, while 22.3% (10,789/48,425) were predicted by both two approaches.

**Functional annotation**

Functional annotation of viral proteins was performed based on the KEGG (Kyoto Encyclopedia of Genes and Genomes) database (downloaded in December 2020) (Kanehisa et al., 2016) using DIAMOND with the options `--query-cover 50 --subject-cover 50 -e 1e-5 --min-score 50 --max-target-seqs 50`. Each protein was assigned a KEGG orthologue (KO) on the basis of the best-hit gene in the database.

Prediction of viral lifestyle (i.e., lytic or lysogenic phage) was implemented based on VIBRANT v1.2.1 (Kieft et al., 2020).

**Phylogenetic analysis**

Phylogenetic analysis was implemented based on high-quality vOTUs with \geq 90% completeness and <10% contamination assessed by the CheckV algorithm. Firstly, by referring to the phylogenetic approach described by Low’s study (Low et al., 2019), we used single-copy protein markers to try to construct a phylogenetic tree based on 4,019 Caudovirales genomes from NCBI RefSeq downloaded in January 2021. Single-copy protein markers were defined using the script provided by Nayfach’s study (https://github.com/snayfach/MGV/blob/master/marker_gene_tree/marker_tree.py) (Nayfach et al., 2021). However, there were only 4 single markers with \geq 10% prevalence among Caudovirales viruses, which displayed the very limited phylogenetic signal. A similar situation occurred in phylogenetic analysis of high-quality vOTUs in OVD (3 single markers), which made it difficult to construct a phylogenetic tree using single-copy protein markers of vOTUs. To perform phylogenetic analysis, we used another phylogenetic approach that was based on genome-wide similarities. We generated a viral proteomic tree of high-quality vOTUs in OVD using ViPTreeGen v1.1.2 (Nishimura et al., 2017) that provided the appropriate choice for understanding previously unknown viral genomes. The proteomic tree was then visualized using iTOL v6.3.2 (Letunic and Bork, 2021).

**The composition of human oral virome**

To explore the spatial specificity of oral virome, all vOTUs in OVD were profiled in oral metagenomes from 4 major oral cavity sites (i.e., buccal mucosa, dental plaque, saliva, and tongue), spanning a total of 1,909 deeply sequenced metagenomes (>5 million clean reads; Table S2). Clean reads in each metagenomic sample were mapped to vOTUs in OVD using bowtie2 with the options `--end-to-end --fast --no-unal -u 5000000 --no-sq`. The abundance profile of vOTUs in each sample was generated by aggregating the number of reads mapped to each vOTU, resulting in “abundance of reads” for each vOTU. The relative abundance of vOTUs was its abundance divided by the number of total mapped reads in each sample. The relative abundance profile at the family level was generated by aggregating the relative abundance of vOTUs assigned to the same family.

**QUANTIFICATION AND STATISTICAL ANALYSIS**

All statistical analyses were performed in R v4.0.3. Alpha diversities were estimated based on the relative abundance of vOTUs: 1) Shannon diversity index was calculated using the function `diversity` with the argument index = shannon; 2) The number of observed vOTUs was counts of unique vOTUs in each sample. Principal coordinates analysis (PCoA) was performed based on the Bray-Curtis distance of vOTU profiles using the function `pcoa`. Permutational multivariate analysis of variance (PERMANOVA) was performed using the function `adonis`, and ADONIS R² was adjusted using the function `RsquareAdj`. When using PERMANOVA, each host property was analyzed after adjusting the other properties. Statistical significance was verified using the function `wilcox.test` and `kruskal.test`. P-values were adjusted using the function `p.adjust` with the argument method = BH, and an adjusted P-value < 0.01 was considered statistical significance.
Data visualizations except for proteomic tree were also carried out using the function `ggplot2` in R. The map of the world was created using the function `geom_polygon` that could load map data provided by the function `map_data`. For the analyses about the effect of individual age and BMI on viral diversities and abundance, regression line in scatter plots was added using the function `geom_smooth` with the argument `method = loess`. 