LazypipeX: Customizable Virome Analysis Pipeline Enabling Fast and

Sensitive Virus Discovery from NGS data

Ilya Weinstein[1,2*], Olli Vapalahti[1,2,3], Ravi Kant[1,2,4#] and Teemu Smura[2,3#]

[1] Department of Veterinary Biosciences, University of Helsinki, 00014 Helsinki, Finland
[2] Department of Virology, University of Helsinki, 00014 Helsinki, Finland
[3] Department of Virology and Immunology, University of Helsinki and Helsinki University Hospital, 00014
Helsinki, Finland
[4] Department of Tropical Parasitology, Institute of Maritime and Tropical Medicine, Medical University
of Gdansk, 81-519 Gdynia, Poland

* To whom correspondence should be addressed. Email: Ilja.Weinstein@helsinki.fi

# Equal contributions

13 ABSTRACT

Metagenomic next-generation sequencing (mNGS) is pivotal for detecting known and novel viruses in
clinical, animal, and wildlife samples, but its efficacy relies on robust bioinformatics pipelines. We
present LazypipeX with multiple advanced features including flexible annotation strategies that can be
adjusted to match different datasets and applications. Using a synthetic benchmark, we show that a
simple two-round annotation strategy designed for speed and low false positive rate required in virus
diagnostics, can reduce execution time to a fraction of BLASTN search without any loss in accuracy.
Additionally, using real data from mosquito samples we show that annotation strategies based on
combinations of fast-to-accurate homology searches reduce execution time from 5 to 20-fold compared
to BLASTN/BLASTP baseline. Using one of these "annotation chains" we were able to discover and
characterise multiple novel whole-genome viral sequences that were missed by Lazypipe1, Lazypipe2
and CZ ID analysis. In LazypipeX we also implemented and benchmarked an improved background
filtering and Integrative Genome Viewer reports for visualising coverage and placement of contigs
relative to reference genomes. LazypipeX represents a highly efficient and versatile tool for virome
analysis, offering customizable and transparent workflows that can facilitate virus discovery and
identification in diverse mNGS applications.

Key words: bioinformatics pipeline; metagenomics; mNGS; virus discovery; virome

30 1 INTRODUCTION

31 Rapid development in metagenomic next generation sequencing (mNGS) technology has enabled
32 virologists to assess the true diversity of viruses in various samples, included human and animal clinical

samples, wildlife and environment (Bassi *et al.* 2022). This is particularly true for viruses that do not propagate in cell culture and novel or emerging viruses that fail to be detected with conventional PCR- or antigen-based methods. mNGS is a high-throughput, unbiased technology that can identify both known and novel pathogens and has great potential for virus discovery, surveillance and broad-spectrum clinical diagnostics (Alawi *et al.* 2019; Kalantar *et al.* 2020; Bassi *et al.* 2022).

Virome identification with mNGS is extensively dependent on bioinformatic analysis and it is increasingly recognised that applied bioinformatic methods will significantly influence the results (Kalantar *et al.* 2020; de Vries *et al.* 2021). The bioinformatics analysis of mNGS data is an actively evolving field, with a wide range of tools available for different steps of the analysis — including read preprocessing, assembly, and annotation — as summarised by Bassi et al. and Ibañez-Lligoña et al. (Bassi *et al.* 2022; Ibañez-Lligoña *et al.* 2023). Since application of stepwise tools require specific knowledge in bioinformatics, pipelines supporting complete workflows have been developed. These can be divided into three types: pipelines for characterising viral diversity and abundance, pipelines for virus diagnostics or identification and pipelines for virus discovery. Here we will focus on pipelines for virus diagnostics and discovery. Pipelines developed for, or applicable to, viral diagnostics focus on high sensitivity and specificity since these will translate into accurate clinical diagnoses and treatment. Open source pipelines that have been applied in virus diagnostics include metaMix (Morfopoulou and Plagnol 2015), RIEMS (Scheuch, Höper and Beer 2015), Kaiju (Menzel, Ng and Krogh 2016), VIP (Li *et al.* 2016), DAMIAN (Alawi *et al.* 2019), Jovian (Zwagemaker *et al.* 2021) and CZ ID (Kalantar *et al.* 2020) with an excellent comparative evaluation presented by de Vries et al. (de Vries *et al.* 2021). Pipelines for virus discovery focus on sensitivity and, as suggested by several authors, benefit from integrating protein alignments and/or protein domain annotation in order to identify highly divergent emerging and novel viruses although at a cost of heavier computation (Zhao *et al.* 2017; Alawi *et al.* 2019; Kalantar *et al.* 2020; Pérot *et al.* 2022). Based on this notion most pipelines for virus diagnostics that include protein level alignments, or protein domain annotation, are also applicable to virus discovery. There is also a number of open-source pipelines that were specifically designed to support discovery of novel pathogens from mNGS data such as VirusSeeker (Zhao *et al.* 2017), DAMIAN (Alawi *et al.* 2019), CZ ID (Kalantar *et al.* 2020) and Microseek (Pérot *et al.* 2022). We have previously published mNGS pipelines, Lazypipe and Lazypipe2, primarily designed for virus discovery (Plyusnin *et al.* 2023), which have been successfully applied to characterize viromes and identify novel viruses in clinical samples (Lehikoinen *et al.* 2024), ticks (Kuivanen *et al.* 2019; Zakham *et al.* 2021; Alfaro-Alarcón *et al.* 2022), mosquitos (Truong Nguyen *et al.* 2022, 2022; Korhonen *et al.* 2024), snakes (Alfaro-Alarcón *et al.* 2022), farm animals (Virtanen *et al.* 2021, 2023) and wildlife (Forbes *et al.* 2019; Mull *et al.* 2023; Alburkat *et al.* 2024; Erdin *et al.* 2024; Ogola *et al.* 2024).

Here we describe LazypipeX with multiple new features including a novel concept of annotation strategies that can incorporate multiple nucleotide, protein and hidden Markov model (HMM) based searches into contig annotations, and that can be adjusted by the user depending on the data and analysis goals. Furthermore, we benchmarked several annotation strategies on synthetic and real data.

With these benchmarks, we showed that a conceptually simple two-round strategy — where the first
round uses a fast Minimap2 search (Li 2018) against a virus-only database, followed by a second round
using BLASTN (Altschul *et al.* 1990) against a larger databases to filter false positives — can reduce
execution time to just 12% of that required by a one-round BLASTN search, without any loss in
accuracy. We also showed that combining varying fast-to-accurate homology searches for analysis of
real mosquito dataset reduced execution time from 5- to 20-fold compared to the BLASTN/BLASTP
baseline. In real terms this translated into average time savings of 19 out of 20 hours for a collection of
~75Mbp sequence libraries analysed on a high-performance cluster. Examining results from one of
these chained annotations we were able to recover and characterise multiple novel whole-genome viral
sequences that were not recovered by Lazypipe1 and manual BLASTX analysis (Truong Nguyen *et al.*
2022). Additionally, we demonstrate how combining HMMscan (Eddy 2011) with other searches
allowed us to identify and characterize a very divergent rhabdo-like genome from a horse fly sample.
In LazypipeX we also implemented (and benchmarked) an improved background filtering and
Integrative Genome Viewer reports for visualising coverage and placement of contigs relative to
reference genomes.
86
2   MATERIAL AND METHODS

2.1   Improved background filtering

LazypipeX supports application of multiple background filters to remove reads and contigs from non-
targeted host or contaminating eukaryotic species. Removing eukaryotic backgrounds allows fast
annotation of background reads and a significantly smaller reference database for contig annotation
since eukaryotic sequences can be excluded from the reference. Filters are applied to both reads prior
to assembling and to contigs after assembling. Multiple background filters can be applied in tandem
when analysing pooled samples (e.g. arthropods) or samples contaminated by multiple sources. As
many samples are likely to contain human nucleic acids introduced during sample collection and library
preparation, it is recommended to filter *Homo sapiens* background in addition to (possibly other) host
background.
The pipeline is supported with several precompiled background filters covering approximately 20
species including human, mosquitos, ticks, bats, domestic and other mammals. These are available
from a public server and can be downloaded on demand. Non-default background filters can be added
by specifying a filter name and a fasta file in the configuration file.
LazypipeX background filtering is implemented using *bwa mem* (Li 2013) alignment of reads to genomic
and genomic mRNA sequences. Resulting sam alignments are then filtered by alignment score (AS,
default) and/or mapping quality (mapq) with Sambamba (Tarasov *et al.* 2015) and SeqKit (Shen *et al.*
2016).

2.2 Flexible and transparent annotation strategies

LazypipeX introduces the concept of "annotation strategy" and allows the user to select different strategies depending on the analysis goals and data types. Annotation strategy can be constructed on demand using several supported homology searches that can be arranged in chains to improve sensitivity or in rounds to improve precision and speed. The supported search engines include nucleotide (Minimap2, BLASTN), protein (SANSparallel, BLASTP/BLASTX and DIAMOND blastp/blastx (Altschul *et al.* 1997; Buchfink, Xie and Huson 2015; Somervuo and Holm 2015)) and HMMscan (Eddy 2011) searches against default or custom reference databases.

Default databases for Minimap2, BLASTN, BLASTP and DIAMOND aligners were constructed using custom Perl scripts. Specifically, databases for Minimap2 and BLASTN were created by downloading RefSeq (release 228) and NT (version 2024/01/01) sequence and accession to taxid data. Taxid lists for archaea, bacteria and viruses were created with *get_species_taxids.sh* from NCBI BLAST+ 2.16.0 package (Camacho *et al.* 2009) and mapped to lists of sequence ids. Archaeal, bacterial and virus sequences were selected to virus (Vi) and archaea-bacteria-virus (ABV) databases using SeqKit (Shen *et al.* 2016). BLASTN indexes were created with NCBI BLAST+ 2.16.0 makeblastdb (Camacho *et al.* 2009). Minimap2 databases were implemented as fasta files combined with accession-to-taxid tsv files. Databases for BLASTP and DIAMOND were created by downloading the latest UniRef100 sequence and accession to taxid data (version 2024/11/27) and by using seed sequences for the cluster-to-taxid mapping. Archaeal, bacterial and virus sequences were selected to Vi and ABV databases using NCBI BLAST+ get_species_taxids.sh (Camacho *et al.* 2009) and SeqKit (Shen *et al.* 2016). BLASTP index was compiled with NCBI BLAST+ 2.16.0 makeblastdb (Camacho *et al.* 2009). DIAMOND index was compiled with DIAMOND version 2.1.6 (Buchfink, Xie and Huson 2015). HMMscan databases were compiled by downloading all viral HMM profiles from PFAM (Mistry *et al.* 2021) and the NeoRdRp2.1 database (Sakaguchi, Nakano and Nakagawa 2024). HMM databases were indexed with hmmpress from HMMER package v3.4 (Eddy 2011). SANSparallel was run on the UniProtKB database via the online server (Somervuo and Holm 2015). All databases described in this section were made freely available for download with stable links and *install_db.pl* utility included with LazypipeX distribution.

For each of the applied database search we used bitscore and/or e-value cut-offs specified in the main config.yaml file. Minimap2 was run with options "-x asm20 –secondary yes –cs -s 200" followed by selection of top hits without ties. BLASTN was run with options "-evalue 10 -max_target_seqs 5" followed by selection of tophits that exceeded bitscore cutoff of 200. BLASTP was run with options "-evalue 10 -max_target_seqs 5" followed by selection of tophits that exceeded bitscore cutoff of 200. DIAMOND blastp was run with options "--evalue 10 --max_target_seqs 5 --very-sensitive" followed by selection of tophits that exceeded bitscore cutoff of 200. SANSparallel server was accessed using PANNZER v3 (Törönen and Holm 2022) with options "-m SANStopHtaxid --SANS_H 5 -R" followed by selection of tophits that exceeded bitscore cutoff of 200. For BLASTP, DIAMOND and SANSparallel ORFs were extracted using MGA (Noguchi, Taniguchi and Itoh 2008) with default options and excluding ORFs shorter than 72 nucleotides. DIAMOND blastx was run with options "--evalue 10 --min_orf 72 --

query-gencode 1 --max-target-seqs 5 --very-sensitive" followed by selection of tophits without ties that

exceeded bitscore cutoff of 200. HMMscan was run with "-E 0.01" followed by selection of tophits with

ties.

Here, we tested several annotation strategies dividing them into one- and two-round strategies for virus

identification, and chained strategies for virus discovery. All benchmarked strategies listed here are

predefined in the pipeline configuration file and can be easily accessed from the command line.

2.2.1   Fast one-round strategies

Minimap.NT$_{Vi}$ and Minimap.RefSeq$_{Vi}$: Minimap2 maps all contigs against NCBI NT or RefSeq viruses

(databases *Minimap.NT$_{Vi}$* and *Minimap.RefSeq$_{Vi}$* in Table1). The goal is to provide very fast annotations,

e.g. for preliminary analysis, at the expense of possibly higher false positive rates. For one-round and

two-round strategies we use a naming convention with database search (e.g. Minimap2) followed by

reference database name (e.g. NT) and database segment subscript (e.g. Vi for viruses).

2.2.2   Two-round strategies for virus identification

In two-round annotations the pipeline first runs a fast search against a virus-only database, e.g.

Minimap2 on NCBI NT viruses and then prunes any false positives in the second round with a slower

BLASTN search on a more extensive database. The two-round strategy aims to speed up the pipeline

by reducing the search space covered by the slower search, while keeping false positive rate in check.

To explain the rationale behind this approach, let's assume that execution time scales linearly to the

size of the query and reference data. Then, execution time for annotation of all contigs with the slower

and more sensitive search against the complete database can be expressed as:

1) $T_1 \sim S(Q_{all})\,S(R_{all})\,V_{slow}$

And execution time for annotating only virus contigs in two-rounds can be expressed as:

2) $T_2 \sim S(Q_{all})\,S(R_{vi})\,V_{fast} + S(Q_{vi})\,S(R_{all})\,V_{slow}$

Where, *S(Q$_{all}$)*, is the size of the assembly, S(*Q$_{vi}$*), is the size of virus part of the assembly, *S(R$_{all}$)*, is the

size of the reference database, *S(R$_{vi}$)*, is the size of the virus part of the reference, *V$_{slow}$* and *V$_{fast}$*, are

the speed of the slower and the faster search.

Finally, assuming both searches are executed in a similar environment, the ration of execution times

can be approximated by:

3) $\frac{T_2}{T_1} \approx \frac{S(Q_{all})\,S(R_{vi})\,V_{fast} + S(Q_{vi})\,S(R_{all})\,V_{slow}}{S(Q_{all})\,S(R_{all})\,V_{slow}} = \frac{S(R_{vi})\,V_{fast}}{S(R_{all})\,V_{slow}} + \frac{S(Q_{vi})}{S(Q_{all})}$

For example, if V$_{fast}$/V$_{slow}$ is ~1/10, and if viruses make up ~10% of the query data and 15% of the

reference data, then T2/T1 ≈ 0.115. In this case the two-round strategy is expected to yield a ~9-fold

speedup. We argue that these assumptions are realistic. For example, in NT released 2024/01/01 virus

sequences constituted 15.3% and the proportion of assembled viral data in most metagenomics
datasets is below 10% (e.g. in (Erdin *et al.* 2025) library from rRNA depleted tissue RNA this was
1.68%), and e.g. Minimap2, although not specifically evaluated against BLASTN, is reported to be tens
of times faster than long read genomic mappers.

Keeping this rationale in mind we tested two two-round strategies: *Minimap.NT$_{Vi}$ -Blastn.NT$_{ABV}$* and
Minimap.RefSeq$_{Vi}$-Blastn.RefSeq$_{ABV}$.

Minimap.NT$_{Vi}$ -Blastn.NT$_{ABV}$: In the 1st round map all contigs with Minimap2 against NT viruses (NT$_{Vi}$,
database descriptions are given in Table1). In the 2nd round annotated pre-labelled viral contigs with
BLASTN against NT archaea, bacteria and viruses (NT$_{ABV}$). The strategy includes background filtering
and assumes that most eukaryotic sequences are removed before the annotation step. This allows to
reduce the database for the 2nd round to NT$_{ABV}$.

Minimap.RefSeq$_{Vi}$-Blastn.RefSeq$_{ABV}$: In the 1st round map all contigs with Minimap2 against RefSeq
virus (RefSeq$_{Vi}$). In the 2nd round annotated pre-labelled viral contigs with BLASTN against RefSeq
archaea, bacteria and viruses (RefSeq$_{ABV}$). The strategy includes background filtering.

These two strategies were designed for virus identification focusing on speed and low false positive
rate. These are suited for samples with clearly defined background sequences such as tissue, swab,
plasma and arthropod samples.

Minimap.NT-Blastn.NT: In the 1st round map all contigs with Minimap2 against complete NT. In the 2nd
round annotated pre-labelled viral contigs with BLAST against complete NT. This strategy has no
background filtering and is suited for samples with complex backgrounds (e.g. faeces or environment
samples).

Table 1. List of reference databases supported by LazypipeX.

Database	Description
Minimap.NT$_{ABV}$, Minimap.NT$_{Vi}$	Fasta and accession-to-taxid files for annotations with Minimap2. NCBI NT archaea-bacteria-viruses and NT viruses.
Blastn.NT$_{ABV}$ and Blastn.NT$_{Vi}$	BLASTN index for NT archaea-bacteria-viruses and NT viruses.
Minimap.RefSeq$_{ABV}$ and Minimap.RefSeq$_{Vi}$	Fasta and accession-to-taxid files for annotations with Minimap2. RefSeq archaea-bacteria-viruses and RefSeq viruses.
Blastn.RefSeq$_{ABV}$ and Blastn.RefSeq$_{Vi}$	BLASTN index for RefSeq archaea-bacteria-viruses and RefSeq viruses
Blastp.UniRef100$_{ABV}$ and Blastp.UniRef100$_{Vi}$	BLASTP index for UniRef100 archaea-bacteria-viruses and viruses
Diamond.UniRef100$_{ABV}$ and Diamond.UniRef100$_{Vi}$	Diamond index for UniRef100 archaea-bacteria-viruses and viruses
HMMscan.PFAM$_{Vi}$	Viral HMM profiles from PFAM database (Mistry et al., 2021)
HMMscan.RdRp	NeoRdRp2 database (Sakaguchi et al., 2024)

2.2.3   Chained strategies for virus discovery

This section describes chained annotation strategies designed for virus discovery with focus on
reducing false negative rate while keeping the execution time feasible. In the chained annotation the
pipeline starts with a fast nucleotide search such as Minimap2. Contigs that have no hits in the first
search are channelled to the second search that is expected to yield higher sensitivity either due to the
search engine or the database. The process is then iterated to searches that have increasingly higher
sensitivity, and in most cases lower precision and lower speed. By chaining several searches this
strategy aims to improve the overall sensitivity. Strategies *Vi.chain1* to *Vi.chain4* include background
filtering and assume that most eukaryotic reads/contigs are removed prior to annotation. This allows to
limit applied reference databases to archaea-bacteria-virus segments (NT_{ABV} and $UniRef100_{ABV}$). These
are suited for samples with clearly defined eukaryotic background sequences such as tissue, swab,
plasma and arthropod samples. Strategy Vi.chain3.env does not filter background and annotates
against complete databases. This is suited for samples with complex backgrounds (e.g. faecal or
environment samples).

Vi.chain0: This is a baseline strategy using BLASTN and BLASTP, which represent the more reliable
tools for identification of viral sequences (Tangherlini *et al.*, 2016). Contigs are annotated with BLASTN
against NT_{ABV} (Table 1), then contigs with no hits are annotated with BLASTP against $UniRef100_{ABV}$.

Vi.chain1: In this strategy Minimap2 is introduced in front of BLASTN. Contigs are annotated with
Minimap2 against NT_{ABV}, then contigs with no hits are annotated with BLASTN against NT_{ABV}, then
contigs with no hits with BLASTP against $UniRef100_{ABV}$.

Vi.chain2: In this strategy BLASTN is omitted from the chain. Contigs are annotated with Minimap2
against NT_{ABV}, then contigs with no hits with BLASTP against $UniRef100_{ABV}$.

Vi.chain3: In this strategy, BLASTP is substituted with a faster protein search, DIAMOND blastp
(Buchfink, Xie and Huson 2015). Contigs are annotated with Minimap2 against NT_{ABV}, then contigs with
no hits with DIAMOND blastp in very-sensitive mode against $UniRef100_{ABV}$. We also tested DIAMOND
blastp in ultra-sensitive mode with no change in sensitivity and DIAMOND blastx in both modes with
clearly lower sensitivity (data not shown). ORFs for DIAMOND search are extracted using MGA
(Noguchi, Taniguchi and Itoh 2008) with default options and excluding ORFs shorter than 72
nucleotides.

Vi.chain4: In this strategy BLASTP is substituted with SANSparallel. Contigs are annotated with
Minimap2 against NT_{ABV}, then contigs with no hits with SANSparallel against UniProtKB (via
SANSparallel online server). ORF extraction is identical to *Vi.chain3*.

Vi.chain3.env: Performs no background filtering and annotates against complete databases. Contigs
are annotated with Minimap against NT, then contigs with no hits with Diamond blastp against
UniRef100.

2.2.4   Supported reference databases

LazypipeX is supported with several precompiled and updated reference databases available from a

public server (Table 1). These can be installed to local volumes with *install_db.pl* utility that is included

in the LazypipeX distribution. The default nucleotide database is the NCBI NT, which was also the most

common reference database employed by 8 out 13 pipelines examined by de Vries et al (de Vries *et al.*

2021) and by CZ ID (Kalantar *et al.* 2020). For the default protein database we chose UniRef100,

similarly to GenomeDetective which uses UniRef90 (Vilsker *et al.* 2019). Other options would be NCBI

NR, e.g. DAMIAN (Alawi *et al.* 2019) and CZ ID (Kalantar *et al.* 2020), and RefSeq proteins, e.g.

metaMix (Morfopoulou and Plagnol 2015), but we choose UniRef100 as in our view it is both

representative and at the same time an order of magnitude smaller than NR.

244 **2.3   Other new features**

LazypipeX adds *host.source* and *genome.composition* information from International Committee on

Taxonomy of Viruses (ICTV) Virus Metadata Resource (VMR) to contig annotation tables whenever

such information is available for the identified genus or family.

Other new features include support for annotations with DIAMOND blastp/blastx (Buchfink, Xie and

Huson 2015) and HMMscan (Eddy 2011). Diamond indices for archaea-bacteria-viruses and viruses in

UniRef100, as well as viral HMM profiles from PFAM database (Mistry *et al.* 2021) and NeoRdRp2

database (Sakaguchi, Nakano and Nakagawa 2024) were added to the list of supported databases.

LazypipeX can optionally output Integrative Genome Viewer (IGV) reports (Robinson *et al.* 2023) for

identified viral contigs. These reports are produced by first binning contigs by the assigned species. For

each species top hits from the annotation table are then retrieved and aligned to contigs with Minimap2.

BAM files are then converted to an interactive IGV reports visualising genome coverage and variants.

These reports are convenient for identifying contig overlaps in fragmented assemblies as well as

sequence similarity and variants. For well annotated reference sequences these reports can also

provide information on contig coverage of key genes such as polymerases. Example IGV reports for

viruses identified in 10 Finnish mosquito pools (accessions SRR19859349-SRR19859358) with

Vi.chain1 annotation strategy are given in Supp. File 1.

261 **2.4   Benchmarking**

2.4.1   Benchmarking background filtering

We evaluated LazypipeX background filtering on two datasets. For the first one we used the human

simulated metagenome from the MetaShot project (Fosso *et al.* 2017). Reads were labelled as human

reads or reads from other taxa based on sequence accessions included in the id field. Thresholds for

alignment score (AS) and mapping quality (mapq) were optimized by stepwise iteration over a range of

values. Background reads were filtered with BWA MEM (Li 2013) and SeqKit (Shen *et al.* 2016) using

human reference genome (acc. GCF_000001405.40). For the second dataset we used Illumina paired-

end 2 x 150 bp libraries sequenced from two liver samples collected from two Finnish brown rats (*Rattus*

norvegicus) of which *Rocahepevirus ratti* was previously detected (Erdin *et al.* 2025). For rat samples

background was filtered with *Rattus norvegicus* reference genome (acc. GCF_036323735.1) and

background filtering was evaluated against read binning by LazypipeX with BLASTN against NT.

2.4.2 Benchmarking strategies for virus identification on simulated data

We evaluated LazypipeX with one-round and two-round strategies designed for virus identification on

the human simulated metagenome (Fosso *et al.* 2017). This dataset is a simulated 20.5 M PE 2 × 150

Illumina library covering approximately 80 viral and 70 bacterial pathogens imbedded in a background

of human reads (Fosso *et al.* 2017). Specifically, we compared two-round strategy *Minimap.NT$_{Vi}$-*

Blastn.NT$_{ABV}$ against a one-round *Blastn.NT$_{ABV}$*, two-round strategy *Minimap.NT$_{Vi}$-Blastn.NT* against

one-round *Blastn.NT (*for definitions see sections 2.2.1 and 2.2.2). The corresponding two-round and

one-round strategies were compared to investigate expected time savings and possible reduction in

accuracy for the two-round approach. We also run *Minimap.RefSeq$_{Vi}$-Blastn.RefSeq$_{ABV}$* to test

annotations on a smaller RefSeq database, and *Minimap.NT$_{Vi}$* and *Minimap.RefSeq$_{Vi}$* to see how a

single round of Minimap annotations would compare to other strategies. In addition, we included results

with DIAMOND blastp against complete NR (labelled as *Diamondp.NR*) as this represents a common

approach in viral metagenomics. DIAMOND blastp was run as an annotation strategy within LazypipeX

as described in section 2.2.

To compare against older releases of Lazypipe we included results with Lazypipe version 2.1 with

background filtering against human reference assembly (accession GCF_000001405.40) and Minimap

annotation (--ann minimap) against NT. Lazypipe2 with Minimap against NT had best performance in

previous evaluations (Plyusnin *et al.* 2023). Finally, we included results from CZ ID web pipeline, version

Illumina v8.3 with NCBI index dated 2024/02/06 and human host subtraction (Kalantar *et al.* 2020). For

all local pipeline runs we used RefSeq release 228, NT dated 2024/01/01, UniRef100 dated 2024/11/27

and NR dated 2025/08/07, and for all local runs, except *Minimap.NT$_{Vi}$-Blastn.NT* and *Blastn.NT,* we

filtered background with human reference assembly (accession GCF_000001405.40).

To properly perform benchmarking sequence accessions present in human simulated metagenome

(Fosso *et al.* 2017) were linked to NCBI taxonomy dated 2025/09/10. Also, for all compared pipelines

reported taxids were linked to NCBI taxonomy dated 2025/09/10 using taxonkit (Shen and Ren 2021).

For CZ ID read counts were selected from NT count field, when this was non-zero, and from NR count

field when NT count was zero. For all pipelines taxa with less then 10 reads and taxa labelled as

bacteriophages were ignored. All benchmarking was performed on a Linux/Unix CPU supercluster with

30 cores each running at 2.1 GHz.

2.4.3 Benchmarking virus discovery on real data

To test the idea of combining searches for virus discovery we run chained annotation strategies

Vi.chain0, Vi.chain1, Vi.chain2, Vi.chain3, Vi.chain4 and Vi.chain3.env on Illumina MiSeq dataset from

90 pools of *Ochlerotatus mosquitoes* (NCBI BioProject PRJNA852425) that was previously analysed
with Lazypipe1 and downstream BLASTX (Truong Nguyen *et al.* 2022). Background was filtered with
four mosquito chromosome level assemblies publicly available at the moment of analysis (*Aedes*
aegypti acc. GCF_002204515.2, *Aedes albopictus* acc. GCF_035046485.1, *Culex pipiens pallens* acc.
GCF_016801865.2, *Culex laticinctus* acc. CA_964187845.1) and the human reference assembly
(GCF_000001405.40). Other pipeline options were kept at default. To compare against previous
pipeline releases the same data was analysed with Lazypipe2 --ann minimap and identical background
filtering.

Additionally, we analysed a previously unpublished Illumina MiSeq library sequenced from a Finnish
horse fly (*Tabanidae sp.*) to test how HMMscan can augment annotation chains for virus discovery.
Background was filtered with a chromosome level assembly from a related species (*Chrysops*
caecutiens, acc. GCA_963971475.1) and the human reference assembly (GCF_000001405.40). For
annotation we applied a strategy that is equivalent to *Vi.chain1* with added HMMscan steps. Specifically,
we used Minimap2 against NT_{ABV}, then BLASTN against NT_{ABV}, then BLASTP against $UniRef100_{ABV}$,
then HMMscan against viral HMM profiles collected from Pfam (Mistry *et al.* 2021) and NeoRdRp v2.1
(Sakaguchi, Nakano and Nakagawa 2024). Other pipeline options were kept at default.

We also compared LazypipeX virus discovery against web CZ ID on three samples with novel findings
by LazypipeX: the horse fly sample and two mosquito samples (accessions SRR19859382 and
SRR19859383). These were analysed with CZ ID version Illumina v8.3, NCBI index 2024/02/06 and
"mosquito" host subtraction for the mosquito samples.

All Lazypipe runs were performed on a Linux/Unix CPU supercluster with 30 cores each running at 2.1
GHz.

2.5   Characterization of viral findings

Annotated array plots visualising ORF structure and closest homologs were constructed for selected
viral findings as follows. Open reading frames (ORFs) were identified with orfipy (Singh and Wurtele
2021), setting minimal ORF length to 300, excluding partial 5' ORFs, including partial 3' ORFs and using
the standard genetic code. Smaller ORFs residing completely inside larger ORFs regardless of the
strand were excluded. Extracted ORFs were annotated with the top hit from DIAMOND blastp
v2.1.10.164 (Buchfink, Reuter and Drost 2021) against UniRef100 viral segment (database retrieved
27/11/2024). ORFs were classified into groups using reciprocal best BLASTP links for pairs of contigs
and manual editing. These groups were then labelled either by gene or protein name most commonly
appearing in the assigned ORF annotations. ORF structure was visualized as an array plot using
geneviewer (Velden 2025).

Boostrapped maximum likelihood (ML) trees were constructed for selected viral findings and
corresponding references as follows. For the "*Inari merhavirus*" refenreces included all ICTV
Alpharhabdovirinae isolates, *Rhabdoviridae* sequences previously identified from this dataset (Truong

Nguyen *et al.* 2022), and BLASTX homologs in clustered NR (with bitscore cut-off set to 400). For "*Primrhavirus kuusamo*" and *"Stangrhavirus lieksa"* references included all ICTV *Deltarhabdovirinae* isolates, *Rhabdoviridae* sequences previously identified from this dataset (Truong Nguyen *et al.* 2022), and BLASTX homologs in clustered NR (with bitscore cut-off set to 400). From these, L-gene protein sequences were extracted, excluding short sequences (<1500 aa) and sequences with no BLASTP homology to the new viruses. Sequences were aligned with MAFFT v7.505 (Katoh and Standley 2013) with options *--retree 2 –maxiter 100* and trimmed with trimAl 1.2rev59 (Capella-Gutiérrez, Silla-Martínez and Gabaldón 2009) with *-automated1* option. For negevirus-like contigs references were collected by mapping BLASTX hits to at most 100 nucleotide accessions and complementing these with exemplar isolates of *Negevirus* member species as listed in (Nunes *et al.* 2017). From these protein sequences corresponding to Methyltransferace_RNA_helicase_RdRp gene were extracted excluding <1000 nt sequences and sequences with no BLASTP homology, aligned and trimmed following the same protocol as for the rhabdo-like contigs. Previously published negevirus- sequences (Truong Nguyen *et al.* 2022) were not included due to poor alignment with the new and reference sequences. For artivirus contigs references were collected by mapping BLASTX hits to at most 100 nucleotide accessions, adding all *Artitivirus* isolates from ICTV and adding all *Totiviridae* sequences published previously (Truong Nguyen *et al.* 2022). From these protein sequences corresponding to both ORFs were extracted excluding <1000 aa ORF1 sequences and sequences with no BLASTP homology, aligned and trimmed as for the rhabdovirus contigs. Alignments for ORF1 and ORF2 were concatenated. For flavivirus contigs references were collected by mapping BLASTX hits with bitscore cutoff at 200 to at most 200 nucleotide accessions complementing these with all *Orthoflavivirus* isolates from ICTV as well as all *Flaviviridae* sequences priviously published for this dataset (Truong Nguyen *et al.* 2022). Polyprotein sequence was extracted excluding <1000 nt sequences and sequences with no BLASTP homology, aligned and trimmed as for other groups.

ML trees were created for all protein alignments with IQTREE v2.1.3 (Minh *et al.* 2020) with options –*ufboot 1000 --nmax 1000* and using ModelFinder (Kalyaanamoorthy *et al.* 2017) to find the best-fitted model. Generated trees were re-rooted on the outgroup for rhabdoviruses (infectious hematopoietic necrosis virus, acc. L40883) or midpoint for other viruses and annotated in TreeViewer (Bianchini and Sánchez-Baracaldo 2024).

3   RESULTS

3.1   Background filtering

Results for background filtering on simulated data are presented in Table 2. Filtering by mapping quality (mapq) was less effective than filtering by alignment score (AS). For the default threshold we selected AS>=100. At this level 99.998% of human reads were filtered, while none of viral reads that were not retroviruses. Notably, 96.375% of human endogenous retroviruses were removed (ERV), which is likely

due to ERV sequences in the human reference genome. To correctly bin ERV sequences we suggest using annotation strategies without background filtering (see discussion).

Table 2 Benchmarking background filtering on simulated data from Fosso et al. (Fosso et al. 2017). Selected filter highlighted. Filter, applied filter with AS representing alignment score and mapq mapping quality threshold, Total, total fraction of reads filtered, Homo sapiens, human reads filtered, Bacteria, Bacterial reads filtered, Viruses, all viral reads filtered, HERV, reads from Human Endogeneous Retroviruses filtered, Viruses excl. Retroviridae, filtered viral reads excluding Retroviridae reads.

Filter	Total	Homo sapiens	Bacteria	Viruses	HERV	Viruses excl. Retroviridae
AS0	99.587 %	100.000 %	91.696 %	97.529 %	99.544 %	96.804 %
AS30	94.687 %	100.000 %	0.021 %	21.982 %	96.453 %	0.151 %
AS50	94.686 %	99.999 %	0.002 %	21.913 %	96.450 %	0.046 %
AS75	94.685 %	99.999 %	0.001 %	21.887 %	96.444 %	0.008 %
AS100	94.684 %	99.998 %	0.000 %	21.861 %	96.354 %	0.000 %
AS125	94.681 %	99.997 %	0.000 %	21.701 %	95.649 %	0.000 %
AS150	76.305 %	80.631 %	0.000 %	11.793 %	51.980 %	0.000 %
mapq0	99.587 %	100.000 %	91.696 %	97.529 %	99.544 %	96.804 %
mapq10	90.708 %	94.938 %	16.278 %	26.474 %	52.932 %	20.282 %
mapq20	89.851 %	94.855 %	1.954 %	12.879 %	50.600 %	2.122 %
mapq30	89.159 %	94.219 %	0.382 %	10.595 %	45.364 %	0.459 %
mapq40	89.043 %	94.117 %	0.038 %	10.147 %	44.656 %	0.022 %
mapq50	88.564 %	93.623 %	0.005 %	8.680 %	38.240 %	0.005 %
mapq60	88.285 %	93.331 %	0.001 %	8.266 %	36.431 %	0.000 %

Results for background filtering for rat samples are presented in Supp. Table 1. Also for this data filtering by alignments score was the best option with threshold at 100 filtering 63-70% of rodent reads and only a few viral (<0.05%) or bacterial (<0.001) reads.

	Rat sample HEV-226				Rat sample HEV-227			
Filter	Total	Rodents	Bacteria	Viruses	Filter	Total	Rodents	Bacteria	Viruses
AS0	38.645 %	99.977 %	59.382 %	50.457 %	AS0	29.894 %	99.961 %	58.730 %	52.605 %
AS30	4.473 %	99.792 %	0.015 %	0.016 %	AS30	0.614 %	99.934 %	0.003 %	0.050 %
AS50	4.107 %	99.780 %	0.002 %	0.001 %	AS50	0.537 %	99.934 %	0.001 %	0.050 %
AS75	3.311 %	99.764 %	0.001 %	0.000 %	AS75	0.440 %	99.750 %	0.000 %	0.050 %
AS100	2.190 %	70.186 %	0.000 %	0.000 %	AS100	0.264 %	63.096 %	0.000 %	0.050 %
AS125	1.350 %	42.986 %	0.000 %	0.000 %	AS125	0.110 %	25.982 %	0.000 %	0.050 %
AS150	0.689 %	22.063 %	0.000 %	0.000 %	AS150	0.038 %	8.621 %	0.000 %	0.050 %
mapq0	38.645 %	99.977 %	59.382 %	50.457 %	mapq0	29.894 %	99.961 %	58.730 %	52.605 %
mapq10	16.435 %	2.751 %	29.209 %	23.875 %	mapq10	17.711 %	2.602 %	33.633 %	22.796 %
mapq20	15.604 %	2.640 %	27.513 %	21.395 %	mapq20	13.985 %	2.471 %	25.234 %	20.341 %
mapq30	15.562 %	2.215 %	27.495 %	21.206 %	mapq30	13.972 %	2.234 %	25.217 %	20.240 %
mapq40	15.549 %	2.046 %	27.487 %	21.193 %	mapq40	13.969 %	2.037 %	25.216 %	20.240 %
mapq50	0.320 %	1.239 %	0.131 %	4.981 %	mapq50	0.335 %	1.971 %	0.266 %	2.004 %
mapq60	0.089 %	0.797 %	0.002 %	0.001 %	mapq60	0.010 %	0.053 %	0.001 %	0.050 %

Supp. Table 1 Benchmarking background filtering on real data with Rattus norvegicus background (Erdin et al., 2025). Selected filter highlighted. Filter, applied filter with AS representing alignment score and mapq mapping quality threshold, Total, total fraction of reads filtered, Rodents, rodent reads filtered, Bacteria, Bacterial reads filtered, Viruses,Viral reads filtered.

3.2 Virus identification from simulated data

We evaluated LazypipeX with several one-round and two-round strategies, Lazypipe2 and web CZ ID on the human simulated metagenome (Fosso *et al.* 2017) as detailed in section 2.4.2. Precision, recall,

F1-score (harmonic mean of precision and recall) in terms of predicted virus taxa at species level,
execution time, maximum memory consumption and erroneous predictions are given in Table 3.

The lowest error rate was observed for LazypipeX annotations against complete NT without host
filtering, with both one-round (*Blastn.NT*) and two-round (*Minimap.NT$_{Vi}$-Blastn.NT*) strategies reporting
two false positive (FP) and two false negative (FN) species. FNs were *Mopeia Lassa virus reassortant*
29 (ML29) which was misclassified to *Mammarenavirus lassaenseı* and *Orthobornavirus bornaense*
with no annotation for the corresponding contig. Notably, ML29 contains S-segment of the *Lassa virus*
which caused the misclassification. FPs were *Orthopoxvirus cowpox* due to one misclassified *Vaccinia*
virus contig, and *African green monkey-like simian foamy virus* due to one misclassified *Human*
spumaretrovirus contig. The execution time for one-round version was ~117h and for two-round version
~27h showing a 4.3X speedup.

The next most accurate strategies were one and two-round annotations against NT$_{Vi}$/NT$_{ABV}$ with
background filtering to remove eukaryotic sequences. Both two-round (*Minimap.NT$_{Vi}$-Blastn.NT$_{ABV}$*) and
one-round (*Blastn.NT$_{ABV}$*) reported two FP and three FN species. FNs include two endogenous
retroviruses (*Human endogeneous retrovirus* and *Human endogenous retrovirus W*) that were filtered
with the host genome and *ML29* which was misclassified to *Mammarenavirus lassaense*. About half
(~1000) of *Orthopneumovirus hominis* reads were misclassified as *O. bovis* (FP)*.* One *Vaccinia virus*
contig was misclassified as a *Orthopoxvirus cowpox* (FP)*.* The execution time for the one-round version
(*Blastn.NT$_{ABV}$*) was ~32h and for two-round version (*Minimap.NT$_{Vi}$-Blastn.NT$_{ABV}$*) ~4h showing 8.4X
speedup. Annotating with just one round of Minimap2 against NT viruses (*Minimap.NT$_{Vi}$*) was ~3.4X
faster than with two rounds (*Minimap.NT$_{Vi}$-Blastn.NT$_{ABV}$*) with no decrease in accuracy.

Annotating against smaller RefSeq database with one and two rounds (*Minimap.RefSeq$_{Vi}$-*
Blastn.RefSeq$_{ABV}$ and *Minima.RefSeq$_{Vi}$*) had higher error rate compared to annotations against NT
resulting in three FP and six FN species.

Annotating with Lazypipe2 resulted in three FN and 16 FP predictions and execution time comparable
to that of the newer 2-round strategy (*Minimap.NT$_{Vi}$-Blastn.NT$_{ABV}$*). FNs included the same three viruses
as in *Minimap.NT$_{Vi}$-Blastn.NT$_{ABV}$*. Lazypipe2 also reported *Orthopneumovirus bovis* (FP) and additional
14 false positive bacteriophage species (Table 3).

Annotating with DIAMOND Blastp against complete NR (*Diamondp.NR*) had low recall and precision
compared to LazypipeX NT-based two-round annotations. *Diamondp.NR* reported five FPs and failed
to report 15 viral pathogens present in the benchmark (Table 3).

Annotating with online CZ ID pipeline resulted in four FNs and 20 FPs. FNs included two retroviruses
likely filtered with host genome, *Betapapillomavirus 2* and *Bocaparvovirus primate1,* which were likely
misclassified to *Human papillomavirus* (FP) and *Human bocavirus* (FP). The latter two of the reported
viruses refer to taxonomic entities very close to the correct predictions and may be attributed to
discrepancies between taxonomies applied in analysis and evaluation. In addition to these CZ ID

reported another four FP eukaryotic viruses and 14 unlabelled bacteriophages leading to an overall low precision (Table 3).

To summarize, switching from one-round annotations (*Blastn.NT* and *Blastn.NT$_{ABV}$*) to two-round annotations (*Minimap.NT$_{Vi}$-Blastn.NT* and *Minimap.NT$_{Vi}$-Blastn.NT$_{ABV}$)* resulted in an ~4-8X speedup with no decrease in accuracy. Background filtering and annotating in two rounds on the NT$_{ABV}$ segment versus annotating on complete NT resulted in ~7X speedup. Altogether, switching from BLASTN annotation on complete NT (*Blastn.NT*) to two-round annotation on NT$_{ABV}$ (*Minimap.NT$_{Vi}$-Blastn.NT$_{ABV}$*) resulted in a ~30X speedup with only slight decrease in precision and identical recall (Table 3). Also, the updated code performed better then Lazypipe2 and web CZID pipelines with fewer classification errors (Table 3).

Table 3 Accessing Lazypipe accuracy of virus taxon retrieval, execution time and memory consumption on simulated data from (Fosso et al., 2017). Compared pipelines are ordered by the descending F1-score. Column headers: TP, number of true positive species, FP, number of false positive species, FN, number of false negative species, F1, F1-score, Maxrss, maximum memory consumption.

[1]Other false negatives for Diamond.NR: Alphainfluenzavirus influenzae, Cytomegalovirus humanbeta5, Morbillivirus hominis, Orthoebolavirus zairense, Orthohepadnavirus hominoidei, Orthomarburgvirus marburgense, Phlebovirus napoliense, TTV-like mini virus and Uukuvirus uukuniemiense.
[2]Other false positives for Lazypipe2: Bacteriophage sp., Bifidobacterium phage Bbif-1, Caudoviricetes sp., Escherichia phage Lambda_ev058 and phi458, Herelleviridae sp., Inoviridae sp., Maltschvirus maltsch, Myoviridae sp. ctk251 and ctZhz2, RockefellervirusIPLA5, Siphoviridae sp. ctX581, Streptococcus phage IPP16, Streptococcus satellite phage Javan375 and uncultured human fecal virus.
[3]Other false positives for CZID: Biseptimavirus BU01, Caudoviricetes sp., Crassvirales sp., Myoviridae sp. ctk251/ctuJM17, Phietavirus B236, Podoviridae sp. ct7Ex2, Siphoviridae sp. ctCCX1/ctGz830/ctSqC25/ctX581/ctxvK3, uncultured human fecal virus and virus sp. ctJrn16.

Strategy	TP	FP	FN	Precision	Recall	F1	Execution Time	Maxrss (MB)	Betapapillomavirus 2	Betaretrovirus maspifimon	Bocaparvovirus primate1	Human endogenous retrovirus K	Human endogenous retrovirus	Human endogenous retrovirus W	Mopeia Lassa virus reassortant 29	Orthobornavirus bornaense	Other False Negatives	AGM simian foamyvirus	Betaretrovirus squmon	Borna disease virus	Human bocavirus	Human papillomavirus	Orthobunyavirus khatangaense	Orthopneumovirus bovis	Orthopoxvirus cowpox	Phlebovirus toscanaense	Similspumavirus chlaet	Similspumavirus pantrotro	Simplexvirus humanalpha1	Spleen focus-forming virus	unknown Hepeviridae	Other False Positives
Minimap.NT$_{Vi}$ - Blastn.NT	80	2	2	97.6%	97.6%	97.6%	1:03:09:17	110,985								FN	FN	FP							FP							
Blastn.NT	80	2	2	97.6%	97.6%	97.6%	4-21:17:37	25,857								FN	FN	FP							FP							
Minimap.NT$_{Vi}$ - Blastn.NT$_{ABV}$	79	2	3	97.5%	96.3%	96.9%	3:53:04	20,827				FN	FN	FN											FP	FP						
Blastn.NT$_{ABV}$	79	2	3	97.5%	96.3%	96.9%	1-08:28:43	25,713				FN	FN	FN											FP	FP						
Minimap.NT$_{Vi}$	79	2	3	97.5%	96.3%	96.9%	1:08:28	12,166				FN	FN	FN											FP			FP				
Minimap.RefSeq$_{Vi}$ - Blastn.RefSeq$_{ABV}$	77	3	5	96.3%	93.9%	95.1%	0:50:09	10,374	FN			FN	FN	FN	FN							FP						FP	FP			
Minimap.RefSeq$_{Vi}$	77	3	5	96.3%	93.9%	95.1%	0:45:51	10,066	FN			FN	FN	FN	FN							FP						FP	FP			
Lazypipe2 --ann minimap	79	16	3	83.2%	96.3%	89.3%	3:43:05	70,548				FN	FN	FN											FP							FP[2]
Diamondp.NR	66	5	15	93.0%	81.5%	86.8%	1-07:50:35	33,550		FN	FN	FN	FN	FN	FN		FN[1]	FP	FP	FP								FP			FP	
CZID NT/NR	78	20	4	79.6%	95.1%	86.7%	NA	NA	FN		FN	FN	FN										FP	FP	FP			FP		FP		FP	FP[3]

3.3 Virus discovery from real data

3.3.1 Benchmarking against BLAST baseline

We compared identification of viral sequences from 90 mosquito samples (Truong Nguyen *et al.* 2022) by five chained strategies *Vi.chain1, Vi.chain2, Vi.chain3, Vi.chain3.env and Vi.chain4,* and Lazypipe2 --ann minimap using viral sequences identified by *Vi.chain0* as our "ground truth" baseline (for strategy definitions see section 2.2.3). This approach reflects limitations posed by real data, for which "true" viral sequences or taxa are unknown. The aim was not to measure absolute accuracy, but rather relative accuracy and relative time savings when comparing viral sequence identification to BLASTN/BLASTP annotations commonly applied in viral metagenomics. In this evaluation, true positives (TP) were

defined as viral sequences (>=500 bp) identified by both evaluated and the baseline strategy, False
positives (FP) as viral sequences (>=500 bp) identified by the evaluated but not the baseline strategy,
and False negatives (FN) as viral sequences (>=500 bp) identified by the baseline but not the evaluated
strategy. Precision, recall and F1-metric were calculated from the corresponding TP, FP and FN
basepair counts for the entire set of 90 samples with results presented in Fig 1 and Supp. Table 2.

Figure 1 Overall performance of chained strategies (and Lazypipe2 –ann minimap) on the mosquito dataset.
Precision, recall, F1-metric and Time were calculated relative to Vi.chain0 baseline strategy as described in the
text. Precision, precision of identified viral base pairs, Recall, recall of viral base pairs, F1, F1-metric (harmonic
mean of precision and recall), Time, execution time relative to Vi.chain0.

Supp. Table 2. Accessing Lazypipe accuracy of viral sequence identification with different annotation strategies on
a dataset of 90 mosquito pools (Truong Nguyen et al., 2022). Ground truth baseline is defined here as viral
sequences identified by Vi.chain0. TP, number of viral sequences (>=500 bp) identified by both evaluated and the
baseline strategy, FP, number of viral sequences (>=500 bp) identified by the evaluated but not the baseline
strategy, FN, number of viral sequences (>=500 bp) identified by the baseline but not the evaluated strategy,
TP.bp/FP.bp/FN.bp, number of base pairs in TP/FP/FN sequences, F1, F1-metric, Time, execution time relative to
Vi.chain0, Maxrss, maximum memory consumption relative to Vi.chain0.

Annotation strategy	Annotation settings	TP	FP	FN	TP.bp	FP.bp	FN.bp	Precision	Recall	F1	Time	Maxrss
Vi.chain0	—ann1 Blastn.NT$_{ABV}$, Blastp.UniRef$_{ABV}$	2,195	0	0	4,212,822	0	0	100.0 %	100.0 %	100.0 %	100.0 %	100.0 %
Vi.chain1	—ann1 Minimap.NT$_{ABV}$, Blastn.NT$_{ABV}$, Blastp.UniRef$_{ABV}$	2,194	0	1	4,211,738	0	1,084	100.0 %	100.0 %	100.0 %	22.2 %	105.6 %
Vi.chain2	—ann1 Minimap.NT$_{ABV}$, Blastp.UniRef$_{ABV}$	2,154	3	41	4,181,975	11,386	30,847	99.7 %	99.3 %	99.5 %	16.0 %	98.9 %
Vi.chain3	—ann1 Minimap.NT$_{ABV}$, Diamondp.UniRef$_{ABV}$ very-sensitive	2,137	6	58	4,165,647	13,169	47,175	99.7 %	98.9 %	99.3 %	4.8 %	131.1 %
Vi.chain3.env	—ann1 Minimap.NT, Diamondp.UniRef —very-sensitive	2,102	5	93	4,121,336	12,659	91,486	99.7 %	97.8 %	98.8 %	49.5 %	417.4 %
Vi.chain4	—ann1 Minimap.NT$_{ABV}$, SANSparallel	1,832	13	363	3,663,913	8,629	548,909	99.8 %	87.0 %	92.9 %	5.0 %	63.1 %
Lazypipe2	—ann1 minimap	1626	0	569	3,268,076	0	944,746	1	77.6 %	87.4 %	8.0 %	63.9 %

Here we compare evaluated chains to the *Vi.chain0* baseline (BLASTN against NT$_{ABV}$, then contigs with
no hits with BLASTP against UniRef100$_{ABV}$). In *Vi.chain1* we added Minimap2 before BLASTN, which
resulted in ~1 kbp of FNs, no FPs and a ~5-fold speedup (Fig 1). In *Vi.chain2* we substituted BLASTN

to Minimap2, which resulted in ~11 kbp of FPs, ~31 kbp of FNs and ~6-fold speedup. In *Vi.chain3* we

substituted BLASTN with Minimap2 and BLASTP with DIAMOND blastp, which resulted in ~13 kbp of

FPs, ~47 kbp of FNs and ~20-fold speedup. In *Vi.chain3.env* we used the same searches as *Vi.chain3*

without prior background filtering and annotating against complete NT/UniRef100 databases. This had

no effect on FPs (~13 kbp) but, nearly doubled the number of false negatives (~91 kbp). We note that

it is debatable whether these are truly negative findings since here we compare against BLAST

searches on NT/UniRef100 that exclude eukaryotic sequences, thus FNs might represent contigs with

close homologs in eukaryotic sequences. Execution time for *Vi.chain3.env* was ~10X higher than for

Vi.chain3 likely due annotating a larger assembly against larger databases. In *Vi.chain4* we substituted

BLASTN with Minimap2 and BLASTP with SANSparallel, which resulted in ~9 kbp of FPs, ~549 kbp of

FNs and ~20-fold speedup. Precision of all LazypipeX strategies was at least 99.7%. Recall was lowest

for strategy incorporating SANSparallel (87.0%), and above 97.8% for all other strategies. Recall by

Lazypipe2 was expectedly lower (77.6%) since this version only allowed annotations against either

nucleotide or protein database. For benchmarking we selected the nucleotide NT database leading to

poor recall of distant homologs.

The average runtime for mosquito samples (mean 74 Mbp, median 46 Mbp, range 0.2-338 Mbp) ranged

from ~5 hours (*Vi.chain1*) to ~1 hour (*Vi.chain3* and *Vi.chain4*) illustrating time savings of up to 19 out

of 20 hours compared to the BLAST baseline.

3.3.2   Identified viral sequences

Here we characterise viral sequences identified by LazypipeX *Vi.chain1* (Minimap2, BLASTN, BLASTP)

from the mosquito dataset and compare these to the previously published findings (Truong Nguyen *et*

al. 2022). We also examine a novel rhabdovirus discovered from the horse fly sample with *Vi.chain1*

augmented with a HMMscan search. Finally, we compare reported rhabdo-, nege-, toti- and flavivirus

sequences reported by *Vi.chain1* and CZ ID pipelines for the horse fly sample and two mosquito

samples (acc. SRR19859382-83). All viral contigs reported by *Vi.chain1* are listed in Supp. Table 3

and CZ ID reports in Supp. File 2.

For the mosquito dataset, *Vi.chain1* identified 1,242 viral contigs exceeding 1 kbp and covering a total

of 3,559 kbp. From these Minimap2 identified 78.1 %, BLASTN 1.6% and BLASTP 20.3%. Keeping in

mind application of chained annotations for virus discovery we next focus on >=1 kbp viral contigs

identified by Minimap2 and BLASTP. We expected that Minimap2 would identify viral sequences

already published for this dataset, while BLASTP would identify divergent and yet unpublished viruses.

3.3.2.1   Identified rhabdoviruses

Minimap2 in *Vi.chain1* identified 44 contigs (>=1kbp) covering *Enontekio merhavirus, E. ohlsrhavirus,*

Hattula rhabdovirus, Inari rhabdovirus, Joutseno rhabdovirus 1 and 2, and Ohldorf virus (see Supp.

Table 3). These include all rhabdoviruses previously published for this dataset (Truong Nguyen *et al.*

2022). BLASTP in *Vi.chain1* identified two contigs covering complete rhabdovirus genomes that were

 provisionally named here according to municipalities they were discovered from as *Kuusamo*

525   *primrhavirus* and *Inari merhavirus* (Fig 2). Comparing these against rhabdoviruses published in

526   previous analysis (Truong Nguyen *et al.* 2022) confirmed very low nucleotide level homology (qcov

527   <9%) for both sequences.

Figure 2 ORF structure of novel Rhabdo-like sequences from mosquito samples and the horse-fly sample. Labels at the center represent closest homologs in UniRef with identity values in parenthesis. *Provisional virus names, isolates and accession numbers.

The *Kuusamo primrhavirus* (identified in acc. SRR19859383) and *Inari merhavirus* (identified in acc.

SRR19859382) both had five ORFs corresponding to N, P, M, G and L genes in a canonical

rhabdovirus genome (Fig 2). In *Kuusamo primrhavirus* ORF1 had 36.5% aa identity to nucleoprotein,

ORF2 29.8% identity to phosphoprotein, ORF4 35.9% identity to glycoprotein and ORF5 52.3%

identity to RdRp in *Primus virus* (UniRef acc. A0A6H0C864_9VIRU, A0A6H0C6Q8_9VIRU,

A0A6H0C7L9_9VIRU and A0A6H0CA79_9VIRU), while ORF3 had no homologs in UniRef, although

by genomic position this appears a putative M gene. In *Inari merhavirus* ORF1 had 35.8% identity to

Merida virus nucleoprotein (A0A894KPN5_9RHAB), ORF2 had no homologs (a putative

phosphoprotein), ORF3 had 26.7% identity to matrix protein in unclassified rhabdovirus

(A0A140DDD9_9RHAB), ORF4 had 49.6% identity to *Merida virus* glycoprotein

(A0A5Q0TW57_9RHAB) and ORF5 52.1% identity to *Culex rhabdovirus* RdRp

(A0A7L7QQ11_9RHAB). Notably, the *Inari merhavirus* had no significant nucleotide similarity to *Inari*

rhabdovirus (acc. ON955143) previously identified from the same pooled sample.

HMMscan identified a single 12.7 kbp rhabdo-like sequence from the horse fly sample with ORF

structure following genomic organisation of a rhabdovirus (Fig 2). This virus was provisionally named

Lieksa stangrhavirus after the municipality of origin. In this contig ORF1 had 24.7% aa identity to

nucleoprotein (acc. A0A0B5KRF8_9RHAB, *Sanxia Water Strider Virus 5*), ORF4 24.4% aa identity to

glycoprotein (acc. A0A7D7F3C3_9RHAB, *Coleopteran rhabdo-related virus OKIAV29*), ORF5 21.9%

aa identity to another glycoprotein (A0A8K1YQP7_9RHAB, *Xiangshan rhabdo-like virus 2*) and ORF6

35.8% aa identity to Replicase (A0A8K1YQP8_9RHAB). ORF2 and ORF3 had no homologs in UniRef

but likely represent P and M genes.

Bootstrapped maximum likelihood (ML) trees were constructed from the L-gene for *Kuusamo*
primrhavirus, *Lieksa stangrhavirus* and *Inari merhavirus* as described in the methods section (Fig 3 and
4)*. In the phylogenetic tree (Fig 3), *Stangrhavirus lieksa* (acc. PV591081) was placed as a sister clade
to a clade with five ICTV *Stangrhavirus* isolates, while *Primrhavirus kuusamo* (acc. BK070983) was
placed within a monophyletic clade of ICTV *Primrhavirus* isolates, confirming the assignment of these
sequences to *Strangrhavirus* and *Primrhavirus* genera. *Inari merhavirus* (acc BK070984) was placed
as an outgroup to the clade of Merida viruses, and within the clade formed by other merhaviruses (Fig
4), confirming its assignment to the *Merhavirus* genus.

Comparing *K. primhavirus* (acc. BK070983) protein sequences to ICTV *Primrhavirus* isolates (acc.
MN661034, BK059423, MN567480, OL700129) showed above 66%, 46% and 63% aa divergence for
the N, L and G genes, respectively, suggesting this is a new *Primrhavirus*. Unlike ICTV primrhaviruses,
which were discovered in metagenomic studies of *Culex and Aedes* mosquitos (Walker *et al.* 2022), *K.*
primrhavirus was sequenced from *Ochlerotatus communis*, further supporting demarcation into a
separate species. Comparing *L. stangrhavirus* (acc. PV591081) protein sequences to ICTV
Stangrhavirus isolates (acc. MW434768, MN013386, MW434775, OL700091, OL700136) showed
above 65%, 46% and 63% aa divergence for the N, L and G genes, respectively, suggesting this is a
new *Stangrhavirus*. Additionally, the G gene in the new genome is split into two ORFs (Fig 2), and the
host species is a horse-fly, while ICTV stangrhaviruses are hosted by mosquitoes.

Comparing *Inari merhavirus* (acc. BK070984) protein sequences to ICTV *Merhavirus* isolates (acc.
OR479699, BK059424, ON955142, ON955143, KU194360, LC775065, AB604791) showed above
67%, 48% and 52% amino acid (aa) divergence for the N, L and G genes, suggesting this is a new
Merhavirus. The aa divergence was high (>70%, >60% and >70%, for N, L and G, respectively) also to
the *Merhavirus inari* (acc. ON955143) which was previously identified from the same sample. Thus, in
the pooled sample individual mosquitoes may have been infected or coinfected with the two different
merhaviruses.

Figure 3 Phylogeny of Lieksa stangrhavirus* and Kuusamo primrhavirus*. Tip colours: red, novel sequences, pink, Rhabdoviridae previously published for the mosquito dataset (Truong Nguyen et al. 2022), green, ICTV Deltarhabdovirinae isolates, blue, homologs retrieved with BLASTX. Columns on the right denote virus genus and host species. Rooted on outgroup (infectious hematopoietic necrosis virus, acc. L40883, not shown), divergent clades collapsed for clarity. *Names assigned to new sequences are provisional.

Figure 4 Phylogeny of Inari merhavirus. Tip colours: red, novel sequence, pink, rhabdoviruses previously identified from the mosquito dataset (Truong Nguyen et al. 2022), green, ICTV Alpharhabdovirinae isolates, blue, homologs retrieved with BLASTX. Columns on the right denote virus genus and host species. Rooted on the outgroup (infectious hematopoietic necrosis virus, acc. L40883, not shown), divergent clades collapsed for clarity. *Names assigned to new sequences are provisional.*

3.3.2.2 Identified negeviruses

Minimap2 in *Vi.chain1* identified 56 contigs (<=1kbp) including *Kustavi negevirus, Mekrijärvi negevirus, Utsojoki negeviruses 1-3, Dezigoudou virus and Corboda virus* (Supp. Table 3), which covers all Negeviruses previously published for this dataset (Truong Nguyen *et al.* 2022). BLASTP in *Vi.chain1* identified four whole-genome negevirus-like sequences ranging between 11.1-11.4 kbp and 21 shorter sequences. Comparing these BLASTP findings against previously published Negeviruses (Truong Nguyen *et al.* 2022) confirmed very low nucleotide level homology (qcov <=3%). ORF structure for the

seven largest BLASTP findings is illustrated in Fig 5. These were provisionally named *Kuusamo* and

Hanko negev-like viruses, represented by four and three isolates, respectively (Fig 5).

Figure 5 ORF structure of novel Negev-like virus sequences from mosquito samples. Labels at the center represent
*closest homologs in UniRef with identity values in parenthesis. *Provisional virus names, isolates and accession*
numbers.

Bootstrapped ML tree was constructed with the RdRp-gene for the seven largest Negev-like virus

sequences as described in the methods section (Fig 6). In the phylogenetic tree novel sequences were

placed into two sister clades with isolates in the first claded hosted by *Ochlerotatus caspius*, and in the

second claded by *O. communis*. Both clades were placed at a distance of known members of the

proposed genus *Negevirus* and closest to negev-like viruses identified from *Bactrocera, Zeugodacus,*

Ceratitis and other *Diptera* (Fig 6). Comparing ORF1 protein sequence (concatenated

Methyltransferase, RNA-helicase and RdRp) between the *Kuusamo negev-like virus* (acc. BK070997),

Hanko negev-like virus (acc. BK070992) and exemplar isolates of the proposed genus *Negevirus* (as

listed in (Nunes *et al.* 2017)) showed below 7% aa divergence in within clade pairwise comparisons,

above 25% aa divergence in between clade comparisons and above 76% divergence to exemplar

Negevirus isolates (Supp. Table 4). These divergence levels (<7% within clades and >25% between

clades) suggest that the seven isolates identified here likely represent two novel negeviruses hosted by

Ochlerotatus mosquitoes.

3.3.2.3   Identified totiviruses and artiviruses

The first search in *Vi.chain1*, Minimap2, identified 315 contigs (>=1kbp) covering 49 out of 52 *Totivirdae* species previously published for this dataset (Truong Nguyen *et al.* 2022) (Supp. Table 2). This search did not report *Hanko totivirus 10 (2.7 kbp), Ilomantsi totivirus 3* (1.2 kbp) and *Karstula totivirus (1.1 kbp)* published previously perhapse due to close similarities to other totivirus-like sequences.

BLASTP identified five whole-genome totivirus-like whole genome contigs ranging from 7.5 to 7.7 kbp and provisionally named as *Joensuu artivirus* (Fig 7). Comparison against previously published *Totivirus* sequences (Truong Nguyen *et al.* 2022) showed very low nucleotide level homology (qcov<14%) for all of these contigs.

Bootstrapped ML tree was constructed with concatenated ORF1 and ORF2 aa sequences. *Joensuu artivirus* formed a monophyletic sister clade to an unclassified *Grogonang virus 29* (OR270267) (Fig 8). The two closest classified viruses were *Artivirus sani* (acc. EU715328) and *Artivirus ni* (acc. GQ342961) confirming assignment to *Artivirus* (Fig 8). *Joensuu artivirus FIN/PK-2018/77* (BK071019)

was selected as the representative isolate and its RdRp aa sequence was compared to other members

of the clade and established member species of *Artivirus* (acc. KX882995, KX882989, AY570982,

MT293125, GQ342961, KX882972, EU715328, AB555544, KX882992). The representative isolate had

<40% aa identity to established *Artivirus* species and >99% aa identity within the *Joensuu artivirus*

sequences suggesting these represented isolates of a novel *Artivirus*.

Figure 7 ORF structure of novel Artivirus sequences from mosquito samples. Labels at the center represent closest
*homologs in UniRef with identity values in parenthesis. *Provisional virus names, isolates and accession numbers.*

3.3.2.4 Identified flaviviruses

The first search in *Vi.chain1*, Minimap2, identified 10 contigs (>=1kbp) covering all *Flaviviridae* species

previously published for this dataset (Truong Nguyen *et al.* 2022) (Supp. Table 2). BLASTP in *Vi.chain1*

identified six contigs mapped to *Flaviviridae*. Comparison against previously published *Flaviviridae*

showed 72.0% nt identity between sequence provisionally named as *Hameenlinna flavivirus (acc.*

BK071037) and *Hanko virus* (acc. ON949929.1), while other sequences were non-homologous. All six

contigs had a long ORFs complete on the 5' end and partial on 3' end with flavivirus polyprotein

homologs in UniRef (Figure 9).

Bootstrapped ML tree was constructed for the six BLASTP *Flaviviridae* sequences (>=1kbp) as

described in the methods section (Fig 10). *Hameenlinna flavivirus FIN/KH-2018/38-2* was placed as a

new isolate of previously published *Hameenlinna flavivirus FIN/KH-2018/38* (acc. ON955057). *Palkane*

flavi-like virus (acc. BK071041) was placed as an outgroup to *Placeda* (acc. MW434111) and *Culex*

tritaeniorhynchus flavi-like (acc. LC514290) viruses. Four novel sequences, including two isolates of

Lestijarvi flavivirus (acc. BK071038 and BK071040) hosted by *Ochlerotatus diantaeus* and two isolates

of *Palkane flavivirus* (acc. BK071036 and BK071039) hosted by *O. communis*, formed a monophyletic

group very distant from previously published *Flaviviridae* and BLASTX homologs (Figure 10). Due to

partial coverage of the polyprotein the *Lestijarvi* and *Palkane* sequences were not characterised in

details, however the performed preliminary phylogenetic analysis indicates that these may represent

isolates of novel flaviviruses in *O. diantaeus* and *O. communis* mosquitoes from Finland. Whether this

is indeed so remains to be shown by future studies. We note that *Palkane flavi-like virus*, *Lestijarvi*

flavivirus and *Palkane flavivirus* are provisional names, which may be subject to change.

Figure 9   ORF structure of novel Flavi-like sequences from mosquito samples. Labels at the center represent
*closest homologs in UniRef with identity values in parenthesis. *Provisional virus names, isolates and accession*
numbers.

*Figure 10 Phylogeny of novel Flavi-like sequences from mosquito samples. Tip colours: novel sequences in red, homologs previously published for this data in light red, BLASTX homologs in blue, ICTV Orthoflavivirus exemplars in a collapsed clade in the upper part of the tree. Columns on the right denote genus and host species. Scale bar is in Mya. *Provisional virus names, isolates and accession numbers.*

3.3.3   Comparison of virus discovery by CZ ID and Lazypipe

Comparing CZ ID and LazypipeX findings for *Rhabdoviridae* from the horse fly and the two mosquito samples (acc. SRR19859382-83) showed that CZ ID was able to identify complete genome for *Inary merhavirus*, but only the L-gene for *Lieksa stangrhavirus* and 138-606 bp fragments of *Kuusamo primrhavirus* reported by LazypipeX (Supp. File 2). For negeviruses CZ ID was able to recover 0.1-1.1 kbp fragments, but not the full genome of *Kuusamo negev-like virus FIN/PP-2018/83* reported by LazypipeX (Supp. File 2). For totiviruses CZ ID did not report divergent totivirus contigs reported by LazypipeX for acc. SRR19859382 and also did not report the full genome *Joensuu artivirus FIN/PP-2018/83* reported by LazypipeX for acc. SRR19859383. For samples analysed with CZ ID LazypipeX and CZ ID did not report *Flaviviridae* contigs. There were also no large (>= 1kbp) novel rhabdo-, nege-, arti- or totivirus sequences reported by CZ ID that would not be reported by LazypipeX (Supp. File 2).

DISCUSSION

LazypipeX is a versatile and efficient pipeline for virus discovery and identification from metagenomic next-generation sequencing (mNGS) data, offering significant advancements in speed, accuracy, and flexibility.

LazypipeX's background filtering enables transparent removal of host and contamination sequences by

specifying database accessions and metadata in the configuration file. This approach achieved a 7-fold

speedup in a simulated benchmark with heavy host sequence backgrounds (Table 3), driven by reduced

data size and a smaller reference database for annotation. Benchmarking on real and synthetic datasets

demonstrated high filtering accuracy (Table 2 and Supp Table 1). The pipeline supports pre-built filters

for common use cases, custom filter creation, and simultaneous application of multiple filters. These

capabilities make LazypipeX adaptable to diverse scenarios, such as clinical samples from human or

animal patients, vector pools (e.g., mosquitoes and ticks), and wildlife specimens with unknown

species-level identification.

Similar to its predecessors LazypipeX employs *de novo* assembling to generate contigs, enhancing the

sensitivity and specificity of homology searches. This approach is shared by pipelines that classify

assemblies instead or in addition to classifying raw reads (Morfopoulou and Plagnol 2015; Scheuch,

Höper and Beer 2015; Alawi *et al.* 2019; Kalantar *et al.* 2020; Zwagemaker *et al.* 2021). LazypipeX

integrates both nucleotide and protein alignments, a feature rare among mNGS pipelines. A recent

comparison found that only two of 13 pipelines used both alignment types (Alawi et al., 2019; Vilsker et

al., 2019), and only DAMIAN incorporated Hidden Markov Models (de Vries et al., 2021). LazypipeX

users can tailor the annotation procedure by chaining homology searches (e.g., Minimap2, BLASTN,

BLASTP, BLASTX, Diamond, SANSparallel, HMMscan) against different databases, optimizing for

sensitivity or specificity depending on the dataset. This flexibility supports iterative adjustment of the

pipeline to match a variety of goals in viral metagenomics.

The two-round annotation strategy in LazypipeX significantly enhances virus identification efficiency.

An initial fast search against a virus-only database, followed by a second round using BLASTN against

a larger database, achieved an 8-fold speedup without compromising accuracy (Table 3). When

combined with eukaryotic background filtering, this approach yielded a 30-fold speedup compared to

single-round BLASTN against the complete NT database. Compared to Lazypipe2, DIAMOND blastp

on NR, and CZ ID, LazypipeX's two-round strategies (*Minimap.NT$_{Vi}$-Blastn.NT* and *Minimap.NT$_{Vi}$-*

Blastn.NT$_{ABV}$) demonstrated superior accuracy in identifying eukaryotic viral taxa (Table 3). For

example, DIAMOND blastp missed 15 viral pathogens, likely due to high protein-level similarity among

related viruses, while CZ ID reported multiple false positives. These results highlight the strength of

LazypipeX's nucleotide-based searches for detecting known viruses.

LazypipeX two-round strategy is not unique it's attempts to perform faster alignments prior to the more

accurate but slower alignments. The idea was inspired by VirusSeeker pipeline that starts by fast

identification of candidate viral sequences against virus-only nucleotide and protein databases and then

removes false positives with alignments against NCBI NT and NR databases (Zhao *et al.* 2017). Also,

in CZ ID reads are first mapped with fast nucleotide and protein aligners prior to aligning contigs with

BLAST against targeted databases (Kalantar *et al.* 2020). In GenomeDetective viral reads are identified

with fast DIAMOND search on UniRef90 prior to assembling and running BLASTN/BLASTX on RefSeq.

Lazypipe complements these approaches by offering flexibility to choose different search engines and
databases for both scanning and refinement stages.

For virus discovery, LazypipeX's chained annotation strategies proved highly effective. In a real dataset
of 90 mosquito pooled samples, the *Vi.chain1* strategy (Minimap2/BLASTN/BLASTP) reduced
execution time 5-fold compared to BLASTN/BLASTP baseline, with precision and recall above 99.7%
and 98.9%, respectively. Substituting BLASTP with faster tools like DIAMOND or SANSparallel
(*Vi.chain3, Vi.chain4*) achieved a 20-fold speedup, though *Vi.chain4* showed lower recall (87.0%).
Vi.chain1 retrieved nearly all previously reported viral sequences (Truong Nguyen et al., 2022) via its
initial Minimap2 search, while subsequent BLASTP search identified novel, distantly related sequences.
This also explains why Lazypipe2 had low recall (78.6%): this version only allowed application of one
search in each analysis, thus, failing to report distant homologs. Incorporating HMMscan into *Vi.chain1*
enabled the discovery of a novel rhabdovirus whole-genome sequence in a horse fly sample, which CZ
ID only partially recovered (L gene only). Comparisons with CZ ID (Kalantar *et al.* 2020) for rhabdo-,
nege-, arti/toti-, and flaviviruses further demonstrated LazypipeX's superior recovery of novel genomes
(Supp. File 2). Findings by BLASTP in Vi.chain1 were characterized with ORF, homology and
phylogenetic analysis, which confirmed that most of these findings were unrelated to viral sequences
previously reported for this dataset and covering complete genomes for at least six novel viruses. These
results underscoring LazypipeX utility for detecting divergent viruses in complex mNGS datasets.

Based on our benchmarking we can suggest general guidelines for selecting Lazypipe strategies for
different use cases (Table 4). For virus identification we advocate using two-round strategies either
against NT_{ABV} for tissue samples or against complete NT for samples with potential retroviral sequences
and/or samples with complex eukaryotic backgrounds (Table 4). For virus discovery from tissue
samples, we advocate using strategies *Vi.chain1* or *Vi.chain3*. Vi.chain3 is particularly well suited for
large datasets, since this showed a 20-fold speedup relative to BLASTN/BLASTP with a very small
reduction in precision and recall (Supp. Table 2). *Vi.chain1* provided a 5-fold speedup with nearly no
reduction in precision/recall, thus this is a viable option for analysing smaller datasets or running
analysis of larger HPCs. For virus discovery from environment, faecal and other samples with complex
eukaryotic background we advocate using *Vi.chain3.env* as this will also bin eukaryotic sequences.

Table 4. Recommended annotation strategies for different user cases

Sequence Data	Analysis Goals	Recommended Annotation Strategy
Illumina libraries from tissue samples and other samples with clearly defined eukaryotic background	Virus identification excluding retroviruses. BLASTN accuracy with up to 30X speedup compared to BLASTN on complete NT.	*background filter + Minimap.NT$_{Vi}$ - Blastn.NT$_{ABV}$*
	Virus identification including retroviruses. BLASTN accuracy with up to 8X speedup compared to BLASTN.	*Minimap.NT$_{Vi}$ - Blastn.NT*
	Virus discovery excluding retroviruses. Precision/recall at ~100%/100% of combined BLASTN/BLASTP with up to 5X speedup.	*background filter + Vi.chain1*
	Virus discovery excluding retroviruses. Precision/recall at ~100%/99% of combined BLASTN and BLASTP with up to 20X speedup.	*background filter + Vi.chain3*
Illumina libraries from environmental, fecal and other samples with complex eukaryotic backgrounds	Virus identification including retroviruses. BLASTN accuracy with up to 8X speedup compared to BLASTN.	*Minimap.NT$_{Vi}$ - Blastn.NT*
	Virus discovery including retroviruses. Precision/recall at ~100%/98% of combined BLASTN and BLASTP with up to 20X speedup.	*Vi.chain3.env*

Future perspectives

761 We plan to extend Lazypipe for use in field settings, particularly for integration with the Oxford Nanopore
762 MinION platform. This version will be optimized for both offline and online modes, enabling real-time
763 analysis of mNGS data in remote locations.

4   Author Contributions

765 Conceptualization of the project: I.W., R.K., T.S. and O.V.; Conceptualization of the pipeline: I.W., R.K.
766 and T.S.; Implementation, updates and code distribution: I.W.; Reference Database Curation: I.W.;
767 Benchmarking: I.W.; Testing and Reporting: I.W., R.K. and T.S.; Writing — Original Draft Preparation:
768 I.W.; Writing — Review & Editing: O.V., R.K. and T.S.; Supervision of the study: R.K., T.S. and O.V.;
769 Funding Acquisition: R.K. and O.V.

5   Acknowledgements

771 The authors wish to acknowledge CSC – IT Center for Science, Finland, for computational resources.

6   Conflicts of Interest

773 The authors have no conflict of interest to declare.

7 Data Availability Statement

LazypipeX user manual and source code are available at https://bitbucket.org/plyusnin/lazypipe/ and https://www.helsinki.fi/en/projects/lazypipe. The produced virus sequences are publicly available in the NCBI GenBank database (for accession codes see Supp. Table 5).

8 Funding

This research was funded by the Academy of Finland (grant number 339510), PREPARE-TID (EU grant number 101137132) DURABLE (EU HERA), VEO – European Union's Horizon 2020 (grant number 874735), the Jane and Aatos Erkko Foundation, and Helsinki University Hospital Funds. Funding bodies were not directly involved in designing or implementing research described in this manuscript.

9 Supplementary Materials

Supplementary Tables 1, 2, 3, 4 and 5

Supplementary Files 1 and 2.

10 References

Alawi M, Burkhardt L, Indenbirken D *et al.* DAMIAN: an open source bioinformatics tool for fast, systematic and cohort based analysis of microorganisms in diagnostic samples. *Sci Rep* 2019;**9**, DOI: 10.1038/s41598-019-52881-4.

Alburkat H, Smura T, Bouilloud M *et al.* Evolution and genetic characterization of Seoul virus in wild rats Rattus norvegicus from an urban park in Lyon, France 2020–2022. *PLoS Negl Trop Dis* 2024;**18**:e0012142.

Alfaro-Alarcón A, Hetzel U, Smura T *et al.* Boid Inclusion Body Disease Is Also a Disease of Wild Boa Constrictors. *Microbiol Spectr* 2022;**10**:e01705-22.

Altschul SF, Gish W, Miller W *et al.* Basic local alignment search tool. *J Mol Biol* 1990;**215**:403–10.

Altschul SF, Madden TL, Schäffer AA *et al.* Gapped BLAST and PSI-BLAST: a new generation of protein database search programs. *Nucleic Acids Res* 1997;**25**:3389–402.

Bassi C, Guerriero P, Pierantoni M *et al.* Novel Virus Identification through Metagenomics: A Systematic Review. *Life* 2022;**12**:2048.

Bianchini G, Sánchez-Baracaldo P. TreeViewer: Flexible, modular software to visualise and manipulate phylogenetic trees. *Ecol Evol* 2024;**14**:e10873.

Buchfink B, Reuter K, Drost H-G. Sensitive protein alignments at tree-of-life scale using DIAMOND. *Nat Methods* 2021;**18**:366–8.

Buchfink B, Xie C, Huson DH. Fast and sensitive protein alignment using DIAMOND. *Nat Methods* 2015;**12**:59–60.

Camacho C, Coulouris G, Avagyan V *et al.* BLAST+: architecture and applications. *BMC Bioinformatics* 2009;**10**:421.

Capella-Gutiérrez S, Silla-Martínez JM, Gabaldón T. trimAl: a tool for automated alignment trimming in large-scale phylogenetic analyses. *Bioinformatics* 2009;**25**:1972–3.

Eddy SR. Accelerated Profile HMM Searches. *PLOS Comput Biol* 2011;**7**:e1002195.

Erdin M, Altan E, Suomalainen N *et al.* Genetic characterization of rat hepatitis E virus (Rocahepevirus ratti) in urban brown rats (Rattus norvegicus) in Helsinki, Finland. *Arch Virol* 2025;**170**:215.

Erdin M, Smura T, Kalkan KK *et al.* Detection of divergent Orthohantavirus tulaense provides insight into wide host range and viral evolutionary patterns. *Npj Viruses* 2024;**2**:1–13.

Forbes KM, Webala PW, Jääskeläinen AJ *et al.* Bombali Virus in Mops condylurus Bat, Kenya. *Emerg Infect Dis* 2019;**25**:955–7.

Fosso B, Santamaria M, D'Antonio M *et al.* MetaShot: an accurate workflow for taxon classification of host-associated microbiome from shotgun metagenomic data. *Bioinformatics* 2017;**33**:1730–2.

Ibañez-Lligoña M, Colomer-Castell S, González-Sánchez A *et al.* Bioinformatic Tools for NGS-Based
Metagenomics to Improve the Clinical Diagnosis of Emerging, Re-Emerging and New Viruses.
Viruses 2023;**15**:587.
Kalantar KL, Carvalho T, de Bourcy CFA *et al.* IDseq—An open source cloud-based pipeline and
analysis service for metagenomic pathogen detection and monitoring. *GigaScience*
2020;**9**:giaa111.
Kalyaanamoorthy S, Minh BQ, Wong TKF *et al.* ModelFinder: fast model selection for accurate
phylogenetic estimates. *Nat Methods* 2017;**14**:587–9.
Katoh K, Standley DM. MAFFT Multiple Sequence Alignment Software Version 7: Improvements in
Performance and Usability. *Mol Biol Evol* 2013;**30**:772–80.
Korhonen EM, Truong Nguyen PT, Faolotto G *et al.* Complete coding sequence of an Aedes flavivirus
strain isolated from Aedes albopictus collected in Northern Italy. *Microbiol Resour Announc*
2024;**13**:e00146-24.
Kuivanen S, Levanov L, Kareinen L *et al.* Detection of novel tick-borne pathogen, Alongshan virus, in
Ixodes ricinus ticks, south-eastern Finland, 2019. *Eurosurveillance* 2019;**24**, DOI: 10.2807/1560-
7917.ES.2019.24.27.1900394.
Lehikoinen J, Nurmi K, Ainola M *et al.* Epstein-Barr Virus in the Cerebrospinal Fluid and Blood
Compartments of Patients With Multiple Sclerosis and Controls. *Neurol Neuroimmunol*
Neuroinflammation 2024;**11**:e200226.
Li H. Aligning sequence reads, clone sequences and assembly contigs with BWA-MEM. *ArXiv Prepr*
ArXiv13033997 2013.
Li H. Minimap2: pairwise alignment for nucleotide sequences. *Bioinformatics* 2018;**34**:3094–100.
Li Y, Wang H, Nie K *et al.* VIP: an integrated pipeline for metagenomics of virus identification and
discovery. *Sci Rep* 2016;**6**:23774.
Menzel P, Ng KL, Krogh A. Fast and sensitive taxonomic classification for metagenomics with Kaiju.
Nat Commun 2016;**7**:11257.
Minh BQ, Schmidt HA, Chernomor O *et al.* IQ-TREE 2: New Models and Efficient Methods for
Phylogenetic Inference in the Genomic Era. *Mol Biol Evol* 2020;**37**:1530–4.
Mistry J, Chuguransky S, Williams L *et al.* Pfam: The protein families database in 2021. *Nucleic Acids*
Res 2021;**49**:D412–9.
Morfopoulou S, Plagnol V. Bayesian mixture analysis for metagenomic community profiling.
Bioinformatics 2015;**31**:2930–8.
Mull N, Erdin M, Smura T *et al.* Novel Ozark Orthohantavirus in Hispid Cotton Rats (Sigmodon hispidus),
Arkansas, USA. *Emerg Infect Dis* 2023;**29**:2576–8.
Noguchi H, Taniguchi T, Itoh T. MetaGeneAnnotator: detecting species-specific patterns of ribosomal
binding site for precise gene prediction in anonymous prokaryotic and phage genomes. *DNA Res*
Int J Rapid Publ Rep Genes Genomes 2008;**15**:387–96.
Nunes MRT, Contreras-Gutierrez MA, Guzman H *et al.* Genetic characterization, molecular
epidemiology, and phylogenetic relationships of insect-specific viruses in the taxon Negevirus.
Virology 2017;**504**:152–67.
Ogola JG, Alburkat H, Smura T *et al.* Detection and genetic characterization of alphacoronaviruses in
co-roosting bat species, southeastern Kenya. 2024:2024.12.23.24319537.
Pérot P, Bigot T, Temmam S *et al.* Microseek: A Protein-Based Metagenomic Pipeline for Virus
Diagnostic and Discovery. *Viruses* 2022;**14**:1990.
Plyusnin I, Vapalahti O, Sironen T *et al.* Enhanced Viral Metagenomics with Lazypipe 2. *Viruses*
2023;**15**:431.
Robinson JT, Thorvaldsdottir H, Turner D *et al.* igv.js: an embeddable JavaScript implementation of the
Integrative Genomics Viewer (IGV). *Bioinformatics* 2023;**39**:btac830.
Sakaguchi S, Nakano T, Nakagawa S. NeoRdRp2 with improved seed data, annotations, and scoring.
Front Virol 2024;**4**, DOI: 10.3389/fviro.2024.1378695.
Scheuch M, Höper D, Beer M. RIEMS: A software pipeline for sensitive and comprehensive taxonomic
classification of reads from metagenomics datasets. *BMC Bioinformatics* 2015;**16**, DOI:
10.1186/s12859-015-0503-6.
Shen W, Le S, Li Y *et al.* SeqKit: A Cross-Platform and Ultrafast Toolkit for FASTA/Q File Manipulation.
PLOS ONE 2016;**11**:e0163962.
Shen W, Ren H. TaxonKit: A practical and efficient NCBI taxonomy toolkit. *J Genet Genomics Yi Chuan*
Xue Bao 2021;**48**:844–50.
Singh U, Wurtele ES. orfipy: a fast and flexible tool for extracting ORFs. *Bioinformatics* 2021;**37**:3019–
20.

Somervuo P, Holm L. SANSparallel: interactive homology search against Uniprot. *Nucleic Acids Res*
2015;**43**:W24–9.
Tangherlini M, Dell'Anno A, Zeigler Allen L *et al.* Assessing viral taxonomic composition in benthic
marine ecosystems: reliability and efficiency of different bioinformatic tools for viral metagenomic
analyses. *Sci Rep* 2016;**6**:28428.
Tarasov A, Vilella AJ, Cuppen E *et al.* Sambamba: fast processing of NGS alignment formats.
Bioinformatics 2015;**31**:2032–4.
Törönen P, Holm L. PANNZER-A practical tool for protein function prediction. *Protein Sci Publ Protein*
Soc 2022;**31**:118–28.
Truong Nguyen PT, Culverwell CL, Suvanto MT *et al.* Characterisation of the RNA Virome of Nine
Ochlerotatus Species in Finland. *Viruses* 2022;**14**:1489.
Velden N van der. *Geneviewer: Gene Cluster Visualizations*., 2025.
Vilsker M, Moosa Y, Nooij S *et al.* Genome Detective: an automated system for virus identification from
high-throughput sequencing data. *Bioinformatics* 2019;**35**:871–3.
Virtanen J, Hautala K, Utriainen M *et al.* Equine dermatitis outbreak associated with parapoxvirus.
2023:2023.09.01.555671.
Virtanen J, Zalewski A, Kołodziej-Sobocińska M *et al.* Diversity and transmission of Aleutian mink
disease virus in feral and farmed American mink and native mustelids. *Virus Evol*
2021;**7**:veab075.
de Vries JJC, Brown JR, Fischer N *et al.* Benchmark of thirteen bioinformatic pipelines for metagenomic
virus diagnostics using datasets from clinical samples. *J Clin Virol* 2021;**141**:104908.
Walker PJ, Freitas-Astúa J, Bejerman N *et al.* ICTV Virus Taxonomy Profile: Rhabdoviridae 2022. *J*
Gen Virol 2022;**103**:001689.
Zakham F, Albalawi AE, Alanazi AD *et al.* Viral RNA Metagenomics of Hyalomma Ticks Collected from
Dromedary Camels in Makkah Province, Saudi Arabia. *Viruses* 2021;**13**:1396.
Zhao G, Wu G, Lim ES *et al.* VirusSeeker, a computational pipeline for virus discovery and virome
composition analysis. *Virology* 2017;**503**:21–30.
Zwagemaker F, Schmitz D, Nooij S *et al.* DennisSchmitz/Jovian: 1.2.07. 2021, DOI:
10.5281/zenodo.4899547.

907
908

Kuusamo negev-like virus* FIN/PP-2018/83 (BK070990)
Bactrocera dorsalis negev-like virus isolate Bc (RdRp 41.3%)

Hanko negev-like virus* FIN/U-2018/92 (BK070991)
Bactrocera dorsalis negev-like virus isolate Bc (RdRp 41.8%)

Hanko negev-like virus* FIN/VS-2018/100-1 (BK070992)
Bactrocera dorsalis negev-like virus isolate Bc (RdRp 41.8%)

Hanko negev-like virus* FIN/VS-2018/100-2 (BK070993)
Bactrocera dorsalis negev-like virus isolate Bc (RdRp 41.0%)

Kuusamo negev-like virus* FIN/Po-2018/31-1 (BK070994)
Bactrocera dorsalis negev-like virus isolate Bc (RdRp 40.7%)

Kuusamo negev-like virus* FIN/Po-2018/31-2 (BK070995)
Bactrocera dorsalis negev-like virus isolate Bc (RdRp 39.0%)

Kuusamo negev-like virus* FIN/PK-2018/70 (BK070997)
Bactrocera dorsalis negev-like virus isolate Bc (RdRp 38.0%)

0.0 1.0k 2.0k 3.0k 4.0k 5.0k 6.0k 7.0k 8.0k 9.0k 10k 11k

■ ORF4 ■ ORF3 ■ Glycoprotein ■ RdRp

Joensuu artivirus* FIN/KP-2018/32 (BK071015)

Penaeid shrimp infectious myonecrosis virus (RdRp 37.1%)

Joensuu artivirus* FIN/L-2018/24 (BK071016)

Penaeid shrimp infectious myonecrosis virus (RdRp 37.1%)

Joensuu artivirus* FIN/PP-2018/82 (BK071017)

Penaeid shrimp infectious myonecrosis virus (RdRp 37.0%)

Joensuu artivirus* FIN/PP-2018/83 (BK071018)

Penaeid shrimp infectious myonecrosis virus (RdRp 37.1%)

Joensuu artivirus* FIN/PK-2018/77 (BK071019)

Penaeid shrimp infectious myonecrosis virus (RdRp 37.1%)

0.0   1.0k   2.0k   3.0k   4.0k   5.0k   6.0k   7.0k

■ Coat protein   ■ RdRp

Label	Host
Wenling toti-like virus 2 (KX882992.1)	crustacean
Wenzhou toti-like virus 1 (KX882995.1)	Charybdis crab
Hanko toti-like virus 2 (OP019860.1)	Ochlerotatus caspius
Ilomantsi toti-like virus 2 (OP019883.1)	Ochlerotatus hexodontus
Vaasa toti-like virus (OP019898.1)	Ochlerotatus excrucians
Kustavi toti-like virus (OP019890.1)	Ochlerotatus caspius
Palkane toti-like virus (OP019895.1)	Ochlerotatus hexodontus
Palkane toti-like virus (OP019896.1)	Ochlerotatus communis
Penaeid shrimp infectious myonecrosis virus (KR815474.1)	Penaeus (Litopenaeus) vannamei
Totiviridae sp. (MZ218620.1)	NA
Golden shiner totivirus (NC_030295.1)	Notemigonus crysoleucas (Golden shiner)
Hubei diptera virus 22 (KX882972.1)	Diptera
Soybean thrips-associated totivirus 2 (MT293125.1)	
Wuhan insect virus 31 (NC_033467.1)	flea and ants
Totiviridae sp. (MZ218641.1)	NA
Omono River virus (AB555544.1)	Culex pipiens pallens
Totivirus sp. (OQ443038.1)	mosquitoes
Omono River virus (MT066059.1)	Aedes albopictus
Yongshan totivirus (PP782332.1)	Culicoides sp.
Culex tritaeniorhynchus totivirus (MN614415.1)	Culex tritaeniorhynchus
Omono River virus (KY264024.1)	Culex vishnui
Omono River virus (KY264025.1)	Culex vishnui
XiangYun toti-like virus 2 (OL700191.1)	Culex pipiens
Drosophila melanogaster totivirus SW-2009a (GQ342961.1)	Drosophila melanogaster
Culex tritaeniorhynchus totivirus (NC_040670.1)	Culex tritaeniorhynchus
Crogonang virus 20 (OR270253.1)	Gonidea angulata
Armigeres subalbatus virus SaX06-AK20 (EU715328.1)	Armigeres subalbatus
Crogonang virus 29 (OR270267.1)	Gonidea angulata
Joensuu artivirus* FIN/PK-2018/77 (BK071019)	Ochlerotatus communis
Joensuu artivirus* FIN/L-2018/24 (BK071016)	Ochlerotatus communis
Joensuu artivirus* FIN/PP-2018/83 (BK071018)	Ochlerotatus communis
Joensuu artivirus* FIN/KP-2018/32 (BK071015)	Ochlerotatus communis
Joensuu artivirus* FIN/PP-2018/82 (BK071017)	Ochlerotatus communis

0.34