



SOFTWARE TOOL ARTICLE

pysradb: A Python package to query next-generation sequencing metadata and data from NCBI Sequence Read Archive [version 1; peer review: 2 approved]

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Abstract

The NCBI Sequence Read Archive (SRA) is the primary archive of next-generation sequencing datasets. SRA makes metadata and raw sequencing data available to the research community to encourage reproducibility and to provide avenues for testing novel hypotheses on publicly available data. However, methods to programmatically access this data are limited. We introduce the Python package, pysradb, which provides a collection of command line methods to query and download metadata and data from SRA, utilizing the curated metadata database available through the SRADB project. We demonstrate the utility of pysradb on multiple use cases for searching and downloading SRA datasets. It is available freely at <https://github.com/saketkc/pysradb>.

Keywords

bioinformatics, metadata, SRA, NGS, NCBI, GEO



This article is included in the **Python Collection** collection.

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Referee Status:

	Invited Referees	
	1	2
version 1 published 23 Apr 2019	 report	 report
1 Simon Andrews , Babraham Institute, UK		
2 Ryan K. Dale , National Institutes of Health, USA		

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Introduction

Several projects have made efforts to analyze and publish summaries of DNA-¹ and RNA-seq^{2,3} datasets. Obtaining metadata and raw data from the NCBI Sequence Read Archive (SRA)⁴ is often the first step towards re-analyzing public next-generation sequencing datasets in order to compare them to private data or test a novel hypothesis. The NCBI SRA toolkit⁵ provides utility methods to download raw sequencing data, while the metadata can be obtained by querying the website or through the Entrez `efetch` command line utility⁶. Most workflows analyzing public data rely on first searching for relevant keywords in the metadata either through the command line utility or the website, gathering relevant sample(s) of interest and then downloading these. A more streamlined workflow can enable the performance of all these steps at once.

In order to make querying both metadata and data more precise and robust, the SRADB⁷ project provides a frequently updated SQLite database containing all the metadata parsed from SRA. SRADB tracks the five main data objects in SRA's metadata: submission, study, sample, experiment and run. These are mapped to five different relational database tables that are made available in the SQLite file. The metadata semantics in the file remain as they are on SRA. The accompanying package, SRADB⁸, made available in the R programming language⁹, provides a convenient framework to handle metadata queries and raw data downloads by utilizing the SQLite database. Though powerful, SRADB requires the end user to be familiar with the R programming language and does not provide a command-line interface for querying or downloading operations.

The `pysradb` package¹⁰ builds upon the principles of SRADB, providing a simple and user-friendly command-line interface for querying metadata and downloading datasets from SRA. It obviates the need for the user to be familiar with any programming language for querying and downloading datasets from SRA. Additionally, it provides utility functions that will further help a user perform more granular queries, which are often required when dealing with multiple datasets on a large scale. By enabling both metadata search and download operations at the command-line, `pysradb` aims to bridge the gap in seamlessly retrieving public sequencing datasets and the associated metadata.

`pysradb`¹⁰ is written in Python¹¹ and is currently developed on GitHub under the open-source BSD 3-Clause License. To simplify the installation procedure for the end-user, it is also available for download through PyPI and bioconda¹².

Methods

Implementation

`pysradb`¹⁰ is implemented in Python and uses `pandas`¹³ for data frame based operations. Since downloading datasets can often take a long time, `pysradb` displays progress for long haul tasks using `tqdm`¹⁴. The metadata information is read in the form of an SQLite¹⁵ database, made available by SRADB⁷.

Each sub-command of `pysradb` contains a self-contained help string that describes its purpose and usage example. The help text can be accessed by passing the '-help' flag. There is also additional documentation available for the sub-commands on [the project's website](#). We also provide example Jupyter¹⁶ notebooks that demonstrate the functionality of the Python API.

`pysradb`'s development primarily occurred on GitHub and the code is tested continuously using Travis CI web-hook. This monitors all incoming pull requests and commits to the master branch. The testing happens on Python version 3.5, 3.6, and 3.7 on an Ubuntu 16.04 LTS virtual machine, while testing webhooks on the bioconda channel provide additional testing on Mac-based systems. Nevertheless, `pysradb` should run on most Unix derivatives.

Operation

`pysradb`¹⁰ can be run on either Linux- or Mac-based operating systems. It supports Python 3.5, 3.6 and 3.7. Requiring just two additional dependencies, `pysradb` can be easily installed using either a `pip`- or `conda`-based package manager via the bioconda¹² channel.

An earlier version of this article can be found on bioRxiv <https://doi.org/10.1101/578500>

Use cases

`pysradb`¹⁰ provides a chain of sub-commands for retrieving metadata, converting one accession to other and downloading. Each sub-command is designed to perform a single operation by default, while additional operations

can be performed by passing additional flags. In the following section we demonstrate some of the use cases of these sub-commands.

`pysradb` uses `SRAMetadb.sqlite`, a SQLite file produced and made available by `SRAdb`⁷ project. The file itself can be downloaded using `pysradb` as:

```
$ pysradb srametadb
```

The `SRAMetadb.sqlite` file is required for all other operations supported by `pysradb`. This file is required for all the sub-commands to function. By default, `pysradb` assumes that the file is located in the current working directory. Alternatively, it can be supplied using the `'-db path/to/SRAMetadb.sqlite'` argument. The `SRAMetadb.sqlite` is available at: <https://s3.amazonaws.com/starbuck1/sradb/SRAMetadb.sqlite.gz> or alternatively at <https://gbnci-abcc.ncifcrf.gov/backup/SRAMetadb.sqlite.gz>. The examples here were run using `SRAMetadb.sqlite` with schema version 1.0 and creation timestamp 2019-01-25 00:38:19.

Search

Consider a case where a user is looking for Ribo-seq¹⁷ public datasets on SRA. These datasets will often have 'ribosome profiling' appearing in the abstract or sample description. We can search for such projects using the 'search' sub-command:

```
$ pysradb search "ribosome profiling" | head
```

study_accession	experiment_accession	sample_accession	run_accession
DRP003075	DRX019536	DRS026974	DRR021383
DRP003075	DRX019537	DRS026982	DRR021384
DRP003075	DRX019538	DRS026979	DRR021385
DRP003075	DRX019540	DRS026984	DRR021387
DRP003075	DRX019541	DRS026978	DRR021388
DRP003075	DRX019543	DRS026980	DRR021390
DRP003075	DRX019544	DRS026981	DRR021391
ERP013565	ERX1264364	ERS1016056	ERR1190989

The results here list all relevant 'ribosome profiling' projects.

Getting metadata for a SRA project

Each SRA project (accession prefix 'SRP') on SRA consists of single or multiple experiments (accession prefix 'SRX') which are sequenced as single or multiple runs (accession prefix 'SRR'). Each experiment is carried out on an individual biological sample (accession prefix 'SRS').

`pysradb metadata` can be used to obtain all the experiment, sample, and run accessions associated with a SRA project as:

```
$ pysradb metadata SRP010679 | head
```

study_accession	experiment_accession	sample_accession	run_accession
SRP010679	SRX118285	SRS290854	SRR403882
SRP010679	SRX118286	SRS290855	SRR403883
SRP010679	SRX118287	SRS290856	SRR403884
SRP010679	SRX118288	SRS290857	SRR403885
SRP010679	SRX118289	SRS290858	SRR403886
SRP010679	SRX118290	SRS290859	SRR403887
SRP010679	SRX118291	SRS290860	SRR403888
SRP010679	SRX118292	SRS290861	SRR403889
SRP010679	SRX118293	SRS290862	SRR403890
SRP010679	SRX118294	SRS290863	SRR403891
SRP010679	SRX118295	SRS290864	SRR403892
SRP010679	SRX118296	SRS290865	SRR403893

However, this information by itself is often incomplete. We require detailed metadata associated with each sample to perform any downstream analysis. For example, the assays used for different samples and the corresponding treatment conditions. This can be done by supplying the '-desc' flag:

```
$ pysradb metadata SRP010679 -desc | head -5
```

study_accession	experiment_accession	sample_accession	run_accession	sample_attribute
SRP010679	SRX118285	SRS290854	SRR403882	source_name: PC3 human prostate cancer cells cell line: PC3 sample type: polyA RNA treatment: vehicle
SRP010679	SRX118286	SRS290855	SRR403883	source_name: PC3 human prostate cancer cells cell line: PC3 sample type: ribosome protected RNA treatment: vehicle
SRP010679	SRX118287	SRS290856	SRR403884	source_name: PC3 human prostate cancer cells cell line: PC3 sample type: polyA RNA treatment: rapamycin
SRP010679	SRX118288	SRS290857	SRR403885	source_name: PC3 human prostate cancer cells cell line: PC3 sample type: ribosome protected RNA treatment: rapamycin

This can be further expanded to reveal the data in 'sample_attribute' column into separate columns via '-expand' flag. This is most useful for samples that have associated treatment or cell type metadata available.

```
$ pysradb metadata SRP010679 -desc -expand
```

... [truncated]

run_accession	cell_line	sample_type	source_name	treatment
SRR403882	pc3	polya rna	pc3 human prostate cancer cells	vehicle
SRR403883	pc3	ribosome protected rna	pc3 human prostate cancer cells	vehicle
SRR403884	pc3	polya rna	pc3 human prostate cancer cells	rapamycin
SRR403885	pc3	ribosome protected rna	pc3 human prostate cancer cells	rapamycin
SRR403886	pc3	polya rna	pc3 human prostate cancer cells	pp242
SRR403887	pc3	ribosome protected rna	pc3 human prostate cancer cells	pp242
SRR403888	pc3	polya rna	pc3 human prostate cancer cells	vehicle
SRR403889	pc3	ribosome protected rna	pc3 human prostate cancer cells	vehicle
SRR403890	pc3	polya rna	pc3 human prostate cancer cells	rapamycin
SRR403891	pc3	ribosome protected rna	pc3 human prostate cancer cells	rapamycin
SRR403892	pc3	polya rna	pc3 human prostate cancer cells	pp242
SRR403893	pc3	ribosome protected rna	pc3 human prostate cancer cells	pp242

Any SRA project might consist of experiments involving multiple assay types. The assay associated with any project can be obtained by providing '-assay' flag:

```
$ pysradb metadata SRP000941 -assay | tr -s ' ' | cut -f5 -d ' ' | tail -n +2 | sort | uniq -c
```

```
999 Bisulfite-Seq
768 ChIP-Seq
121 OTHER
353 RNA-Seq
28 WGS
```

Getting SRPs from GSE

The Gene Expression Omnibus database (GEO)¹⁸ is the NCBI data repository for functional genomics data.

It accepts array and sequence-based data from gene profiling experiments. For sequence-based data, the corresponding raw files are deposited to the SRA. GEO assigns a dataset accession (accession prefix 'GSE') that is linked to the corresponding accession on the SRA (accession prefix 'SRP'). It is often necessary to interpolate between the two accessions. `gse-to-srp` sub-command allows converting GSE to SRP:

```
$ pysradb gse-to-srp GSE24355 GSE25842

study_alias  study_accession
GSE24355    SRP003870
GSE25842    SRP005378
```

It can be further expanded to obtain the corresponding experiment and run accessions:

```
$ pysradb gse-to-srp -detailed -expand GSE100007 | head

study_alias  study_accession  experiment_accession  sample_accession  experiment_alias  sample_alias
GSE100007    SRP109126        SRX2916198           SRS2282390       GSM2667747       GSM2667747
GSE100007    SRP109126        SRX2916199           SRS2282391       GSM2667748       GSM2667748
GSE100007    SRP109126        SRX2916200           SRS2282392       GSM2667749       GSM2667749
GSE100007    SRP109126        SRX2916201           SRS2282393       GSM2667750       GSM2667750
GSE100007    SRP109126        SRX2916202           SRS2282394       GSM2667751       GSM2667751
GSE100007    SRP109126        SRX2916203           SRS2282395       GSM2667752       GSM2667752
GSE100007    SRP109126        SRX2916204           SRS2282396       GSM2667753       GSM2667753
GSE100007    SRP109126        SRX2916205           SRS2282397       GSM2667754       GSM2667754
GSE100007    SRP109126        SRX2916206           SRS2282400       GSM2667755       GSM2667755
```

Getting a list of GEO experiments for a GEO study

Any GEO study (accession prefix 'GSE') will involve a collection of experiments (accession prefix 'GSM'). We can obtain an entire list of experiments corresponding to the study using the `gse-to-gsm` sub-command from `pysradb`:

```
$ pysradb gse-to-gsm GSE41637 | head

study_alias  experiment_alias
GSE41637    GSM1020640_1
GSE41637    GSM1020641_1
GSE41637    GSM1020642_1
GSE41637    GSM1020643_1
GSE41637    GSM1020644_1
GSE41637    GSM1020645_1
GSE41637    GSM1020646_1
GSE41637    GSM1020647_1
GSE41637    GSM1020648_1
```

However, a list of GSM accessions is not useful if one is performing any downstream analysis, which essentially requires more detailed information about the metadata associated with each experiment. This relevant metadata associated with each sample can be obtained by providing `gse-to-gsm` additional flags:

```
$ pysradb gse-to-gsm -desc GSE41637 | head

study_alias  experiment_alias  sample_attribute
GSE41637    GSM1020640_1     source_name: mouse_brain || strain: DBA/2J || tissue: brain
GSE41637    GSM1020641_1     source_name: mouse_colon || strain: DBA/2J || tissue: colon
GSE41637    GSM1020642_1     source_name: mouse_heart || strain: DBA/2J || tissue: heart
GSE41637    GSM1020643_1     source_name: mouse_kidney || strain: DBA/2J || tissue: kidney
GSE41637    GSM1020644_1     source_name: mouse_liver || strain: DBA/2J || tissue: liver
GSE41637    GSM1020645_1     source_name: mouse_lung || strain: DBA/2J || tissue: lung
GSE41637    GSM1020646_1     source_name: mouse_skm || strain: DBA/2J || tissue: skeletal muscle
GSE41637    GSM1020647_1     source_name: mouse_spleen || strain: DBA/2J || tissue: spleen
GSE41637    GSM1020648_1     source_name: mouse_testes || strain: DBA/2J || tissue: testes
```

The metadata information can then be parsed from the `sample_attribute` column. To obtain more structured metadata, we can use an additional flag `'-expand'` :

```
$ pysradb gse-to-gsm -desc -expand GSE41637 | head

study_alias  experiment_alias  source_name  strain  tissue
GSE41637    GSM1020640_1    mouse_brain  dba/2j  brain
GSE41637    GSM1020641_1    mouse_colon  dba/2j  colon
GSE41637    GSM1020642_1    mouse_heart  dba/2j  heart
GSE41637    GSM1020643_1    mouse_kidney dba/2j  kidney
GSE41637    GSM1020644_1    mouse_liver  dba/2j  liver
GSE41637    GSM1020645_1    mouse_lung   dba/2j  lung
GSE41637    GSM1020646_1    mouse_skm    dba/2j  skeletal muscle
```

Getting SRR from GSM

`gsm-to-srr` allows conversion from GEO experiments (accession prefix 'GSM') to SRA runs (accession prefix 'SRR'):

```
$ pysradb gsm-to-srr GSM1020640 GSM1020646

experiment_alias  run_accession
GSM1020640_1     SRR594393
GSM1020646_1     SRR594399
```

Downloading SRA datasets

`pysradb` enables seamless downloads from SRA. It organizes the downloaded data following the NCBI hierarchy: 'SRP => SRX => SRR' of storing data. Each 'SRP' (project) has multiple 'SRX' (experiments) and each 'SRX' in turn has multiple 'SRR' (runs). Multiple projects can be downloaded at once using the `download` sub-command:

```
$ pysradb download -p SRP000941 -p SRP010679
```

`download` also allows Unix pipes-based inputs. Consider our previous example of the project SRP000941 with different assays. However, we want to be able to download only 'RNA-seq' samples. We can do this by subsetting the metadata output for only 'RNA-seq' samples:

```
$ pysradb metadata SRP000941 -assay | grep 'study|RNA-Seq' | pysradb download
```

This will only download the 'RNA-seq' samples from the project.

Summary

`pysradb`¹⁰ provides a command-line interface to query metadata and download sequencing datasets from the SRA. It enables seamless retrieval of metadata and conversion between different accessions. `pysradb` is written in Python 3 and is available on Linux and Mac OS. The source code is hosted on GitHub and licensed under BSD 3-clause license. It is available for installation through PyPI and bioconda.

Data availability

Underlying data

Dataset from DDBJ Sequence Read Archive, Accession number DRP003075: <https://identifiers.org/insdc.sra/DRP003075>

Dataset from EMBL-EBI Sequence Read Archive, Accession number ERP013565: <https://identifiers.org/insdc.sra/ERP013565>

Dataset from Gene Expression Omnibus, Accession number GSE24355: <https://identifiers.org/geo/GSE24355>

Dataset from Gene Expression Omnibus, Accession number GSE25842: <https://identifiers.org/geo/GSE25842>

Dataset from Gene Expression Omnibus, Accession number GSE100007: <https://identifiers.org/geo/GSE100007>¹⁹

Dataset from Gene Expression Omnibus, Accession number GSE41637: <https://identifiers.org/geo/GSE41637>²⁰

Dataset from NCBI Sequence Read Archive, Accession number SRP010679: <https://identifiers.org/insdc.sra/SRP010679>²¹

Dataset from NCBI Sequence Read Archive, Accession number SRP000941: <https://identifiers.org/insdc.sra/SRP000941>²²

Software availability

Software available from: <https://pypi.org/project/pysradb/>.

Source code available from: <https://github.com/saketkc/pysradb>.

Archived source code at time of publication: <https://doi.org/10.5281/zenodo.2579446>¹⁰.

License: BSD 3-Clause

Author endorsement

Dr. Luiz O. Penalva confirms that the author has an appropriate level of expertise to conduct this research, and confirms that the submission is of an acceptable scientific standard. Dr. Luiz O. Penalva declares they have no competing interests. Affiliation: UT Health San Antonio, Children's Cancer Research Institute, San Antonio, Texas, 78229, USA

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Open Peer Review

Current Referee Status:  

Version 1

Referee Report 07 May 2019

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Ryan K. Dale 

Bioinformatics and Scientific Programming Core, National Institute of Child Health and Human Development, National Institutes of Health, Bethesda, MD, USA

Pysradb is a Python package that extends the existing SRADB R package by exposing command-line functionality such that metadata queries can be piped to other command-line tools.

Overall this is a well-written tool with good documentation and tests. As-is, it is very useful and my comments here are only aimed at making the tool more useful.

Downloading:

It would be really nice to be able to download fastqs, perhaps using SRADB's strategy of downloading them from EBI.

When trying to run the example from the docs, "pysradb download -p SRP000941", I got a 550 HTTP error. Upon closer inspection, some of the SRAs in this entry are no longer available, but it causes the whole download to fail. I'm not quite sure of the underlying cause—this seems to be a mismatch between what's in SRA and the metadata database even though I just downloaded a fresh database with a timestamp of a week ago.

Two suggestions for making the download more robust. The first is to move on if there are any failures, reporting any failures at the end to stderr. The second is to provide a --list option that only prints the table with URLs such that the user can extract the URLs as needed (for example, to allocate to separate nodes on a cluster where failures can be handled by other mechanisms). With the current set of arguments, the only way to do the latter is to interactively answer "N" to the question of whether to continue downloading.

It could be useful to just download individual SRRs; I do not see that functionality available.

Searching:

An explanation of the search syntax would be helpful (e.g., SQL syntax, which allows wildcards but not regular expressions). More complex examples would be good here. Given the heterogeneity of user-entered metadata in SRA, it would be convenient to use regular expressions when searching (which appears to be possible with Python and sqlite3 using a callback function).

I would prefer a nicer-formatted "no results found" printed to stderr rather than RuntimeError and UserWarning.

Other:

An interface to check the timestamp from the metaInfo table of the database would be helpful.

This may be a personal preference, but I would rather have output be tab-separated instead of requiring additional `tr -s " " | cut -d " "` commands (or `awk`) to do any sort of manipulation on the command line.

Is the rationale for developing the new software tool clearly explained?

Yes

Is the description of the software tool technically sound?

Yes

Are sufficient details of the code, methods and analysis (if applicable) provided to allow replication of the software development and its use by others?

Yes

Is sufficient information provided to allow interpretation of the expected output datasets and any results generated using the tool?

Yes

Are the conclusions about the tool and its performance adequately supported by the findings presented in the article?

Yes

Competing Interests: No competing interests were disclosed.

Reviewer Expertise: bioinformatics, genomics, chromatin, gene expression throughout development

I have read this submission. I believe that I have an appropriate level of expertise to confirm that it is of an acceptable scientific standard.

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Pysradb is a command line utility, written in python, which provides an easy scriptable interface for querying metadata and datasets from the SRA database.

The authors correctly point out that interacting with GEO/SRA through the official site and APIs can be slow and frustrating, and thus having a tool which can make this process more streamlined is of great value.

Both the paper and the documentation for the tool are well written and easy to understand. The tool provides a series of individual operations performing queries, translations or downloads and these can be run individually or chained together through pipes (which is really nice!).

The tool requires an initial download of the sqlite database from the SRADB project. Whilst I can see how this then makes all subsequent operations quick, it does mean that you have to download a >2GB file (which expands to >30GB), taking 30+ mins before you can do anything with the program. It presumably also means that you need to re-download this file every time there is an update to the data in GEO otherwise your searches are likely to be out of date. On our site at least, people are often getting data for papers which have just been released so this is going to entail a lot of waiting for this file to download. It would be great if there was a way to point to a publicly accessible SQL server to do queries without having to do the local download, and then providing the option of pulling a local copy if you need greater performance. Also having a way to do incremental updates to this file instead of re-downloading the whole thing would be nice. Neither of these is a deal breaker, but they mightn't be too hard to implement?

The individual tools all worked as described, with the exception of the issues listed at the bottom, and the experience was generally very good with the tool.

One frustrating limitation is that the piping support is not universal throughout the tool. You can pipe into the download command, but not, for example, into the metadata command. Being able to chain operations such as:

```
pysrdb gse-to-srp GSE24355 | pysrdb metadata | pysra download
```

..or

```
pysrdb search "oocyte development" | head | pysrdb metadata
```

..would be really nice and presumably not too hard to support?

The downloading side of the tool is very useful and probably the part which is hardest to achieve in the main SRA site. Whilst this worked as described there are some aspects of the way it works which make it a little frustrating. Firstly, it downloads SRA files, which hardly anyone wants - having a way to get the fastq files directly would be a really useful addition rather than having to run fastq-dump manually afterwards. It also downloads into a structured set of folders, which makes sense, but for large downloads means your files are scattered through multiple folders which makes life harder when you want to process them. Even the --out-dir option doesn't mean the files are in that directory, but just that it's used as a basename. For the names of the files it would be nicer to have a name which incorporated the relevant SRR/SRX ids and maybe the user submitted sample name so that you can actually have a meaningful and complete name from the file. For example, the types of filenames generated by SRA explorer (<https://ewels.github.io/sra-explorer/>) are a nice compromise between being predictable, unique and yet informative at the same time.

If I'm being really picky I'd also quibble a bit at the choice of some of the defaults in the API. For example, I can't see why the --desc and --expand options aren't the default for the metadata sub-program - give me everything in a nice format and let me cut that down if I don't need everything.

Overall this tool is really nice and will be useful for a lot of people. With a small amount of refinement this is likely to become part of our standard toolbox.

Minor points to address:

- 1) The API seems to have changed since the paper was written. The option to download the metadata is now pysradb metadb, and not pysradb srametadb. This is wrong in both the paper and the Jupyter notebook example on the github page and should be changed. It might be nice to allow srametadb as a fallback if people have been using the old name?
- 2) Some of the documentation is incomplete. For example the quickstart documentation at <https://saket-choudhary.me/pysradb/quickstart.html#the-full-list-of-possible-pysradb-operations> doesn't list the search operation so I couldn't look up the options I had for that.
- 3) The piping option is great, but on my system generates a crash if there is too much output (possibly more than the pipe buffer can hold?). Submitted as bug #7
- 4) The metadata command line in the paper is broken. You need double dashes before the options, so --desc rather than -desc. This seems to happen elsewhere as well and might just be an auto-format problem.
- 5) If using wget to download the progress bar for downloading doesn't work. The data comes down but the progress stays at 0%.

Is the rationale for developing the new software tool clearly explained?

Yes

Is the description of the software tool technically sound?

Yes

Are sufficient details of the code, methods and analysis (if applicable) provided to allow replication of the software development and its use by others?

Yes

Is sufficient information provided to allow interpretation of the expected output datasets and any results generated using the tool?

Yes

Are the conclusions about the tool and its performance adequately supported by the findings presented in the article?

Yes

Competing Interests: No competing interests were disclosed.

Reviewer Expertise: bioinformatics

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