

# tskit: WORKING WITH TREE SEQUENCES

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Code: github.com/petrelharp/progen-2023.

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### Interfaces and interoperability

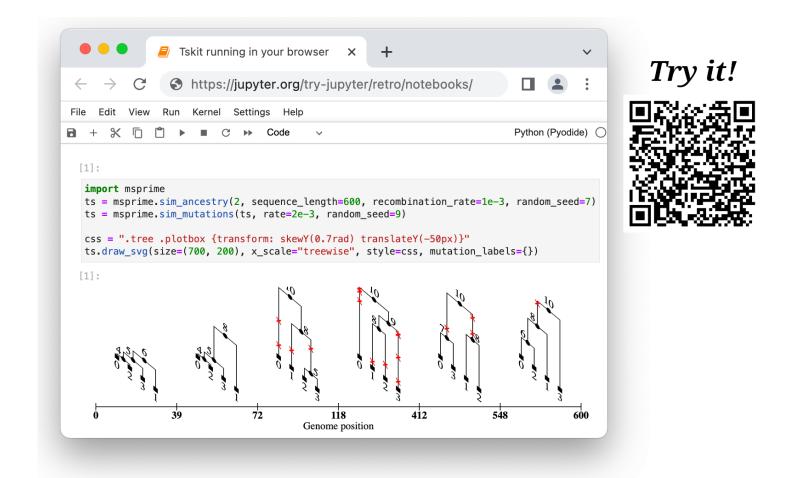
We aim to provide stable, well-tested and well-documented software so others can reliably build with it – including a backwards compatibility guarantee. tskit is already being used in a number of inference and simulation packages. The core functionality is implemented via a C API, and the primary interface is via a Python library, but others are available:

- Well-tested API used by many software packages: SLiM, fwdpy11, msprime, tsinfer/tsdate, Relate, slendr, etc.
- Available in multiple programming languages:





- Runs in-browser (no install required!) for quick demos / teaching (see screenshot).
- Can represent full Ancestral Recombination Graphs; includes ARG likelihood calculations.
- Interoperable with other packages (e.g., VCF output for sequence data, newick/nexus output to Dendropy, numpy arrays to scikit-



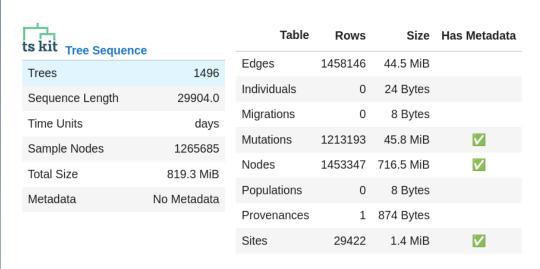
# METADATA

tskit now has integrated metadata for all objects (genomes, mutations, sites, etc). For instance (spoiler alert), the complete ARG for 1.26 million SARS-Cov-2 genomes (until mid-2021): fits in 57MB, and loads in under 1 second! It is 819 MB once decompressed, and has metadata attached to three tables.

-rw-rw-r-- 1 jk jk 57M Mar 2 13:32 SARS-Cov-2-ARG.ts.tsz

ts = tszip.decompress("SARS-Cov-2-ARG.ts.tsz")

CPU times: user 775 ms, sys: 533 ms, total: 1.31 s Wall time: 842 ms



The integrated data model links nodes, edges, sites and mutations, and now allows annotation of all objects with arbitrary external metadata. For instance, here's the first five sites in the SARS-Cov-2 ARG, and metadata for a sample (available as a dictionary!):

## ts.tables.sites[:5]

id	position	ancestral_state	metadata
0	56	G	{'masked_samples': 727232}
1	57	А	{'masked_samples': 726137}
2	58	Т	{'masked_samples': 725063}
3	59	С	{'masked_samples': 724533}
4	60	Т	{'masked_samples': 721663}

#### dataclasses.asdict(ts.node(1026732))

{'id': 1026732, 'flags': 1, 'time': 60.0, 'metadata': {'Imputed\_lineage': 'B.1.1.7', 'Nextclade\_pango': 'B.1.1.7', 'clade': '20I (Alpha, V1)', 'country': 'Germany', 'date': '2021-05-01', 'date\_submitted': '2021-05-17', 'gisaid\_epi\_isl': 'EPI\_ISL\_2122637', 'sc2ts\_qc': {'num\_masked\_sites': 150, 'original\_base\_composition': {'-': 103, 'A': 8902, 'C': 5473, 'G': 5847, 'T': 9578}], 'strain': 'Germany/un-RKI-I-137988/2021', 'totalSubstitutions': 36.0}}

## OVERVIEW

to use and build on. Why might you want to use tree sequences?

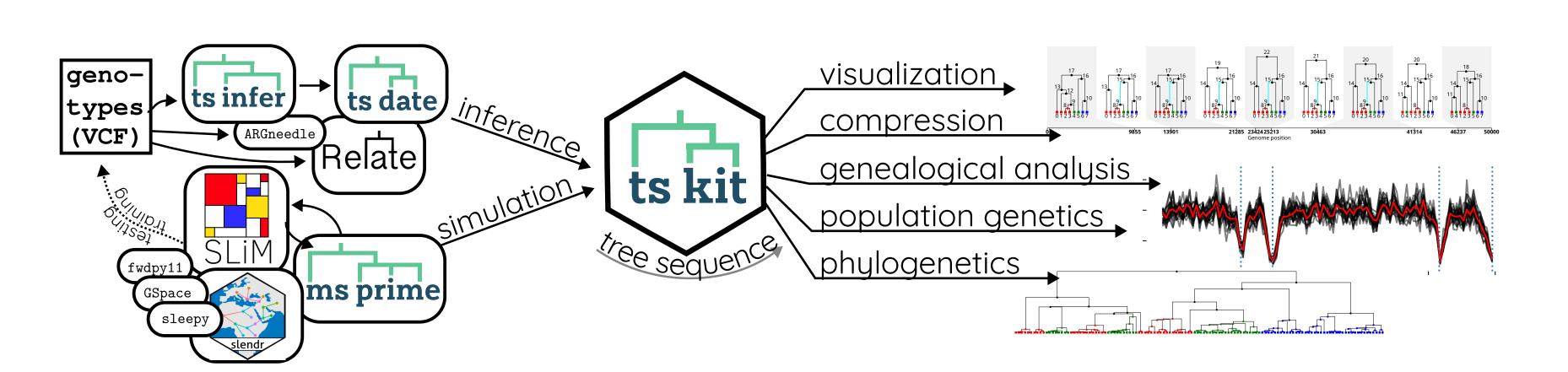
- For large samples, stores genotypes losslessly using (estimated) underlying genealogical relationships in orders of magnitude less space,
- ... and allows fast processing and exploration, in seconds, not hours.
- Genealogical relationships "the trees" are often closer to things we want to learn about
- ... and explicitly include a time dimension.
- History of a process can be recorded in a simulation, not just the genotypic outcome,
- ... and simulations can be much faster/more efficient.

tskit is the C and python library providing tools for working with suc- In summary: by representing genomes using the genealogical process that cinct tree sequences. We provide solid, stable, well-tested software for you generated the data, we get both a huge advantage storing and manipulating genomic data, as well as a more direct look at the processes that generated the data.

Here's some silly slogans, care to suggest any more?

- "tskit: launching your genomes into the time dimension!"
- "tskit: tree thinking, for popgen"
- "tskit: stable software for genome-scale trees"
- "tskit: all your insights, much faster!"

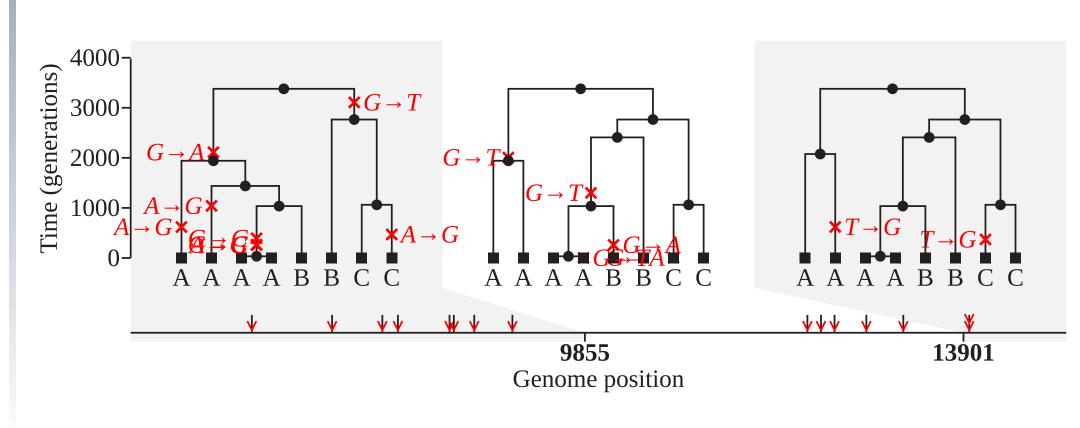
Documentation and examples: https://tskit.dev/



# $V_{\rm ISUZALIZATION} - { m see\ more\ at\ https://tskit.dev/tutorials/viz.html}$

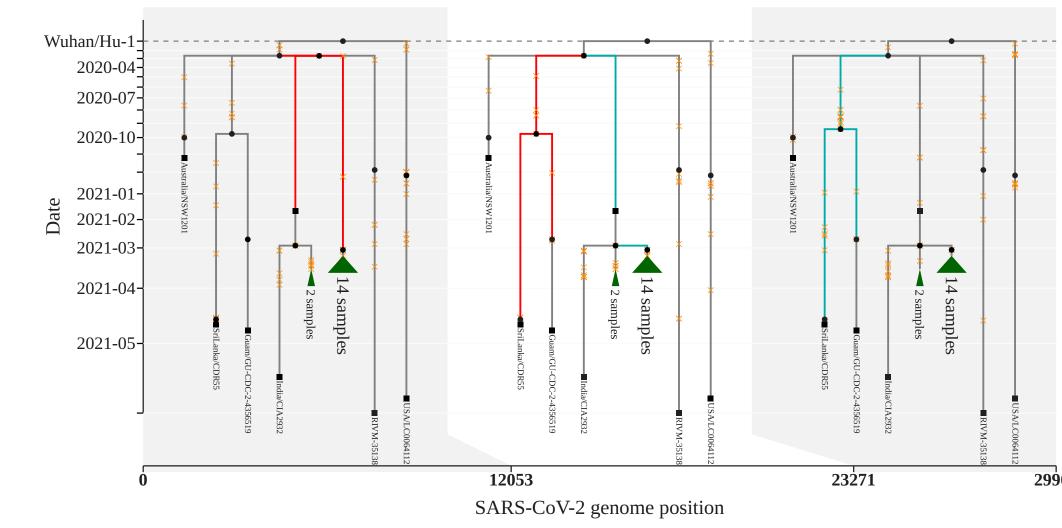
SVG-based visualization allows flexible styling of local trees.

- set labels for nodes, mutations, and tickmarks: e.g., using metadata simp\_ts.draw\_svg(
- color elements: e.g., to highlight branches that change between trees, mutations by type, or samples by location
- transform elements: e.g., rotate labels, alter node symbols, even 3D effects!
- timescale titles show time units by default (scaling can be linear, log, or rank)
- interaction possible via mouseover events and javascript animation
- text-based plots also available for simple debugging



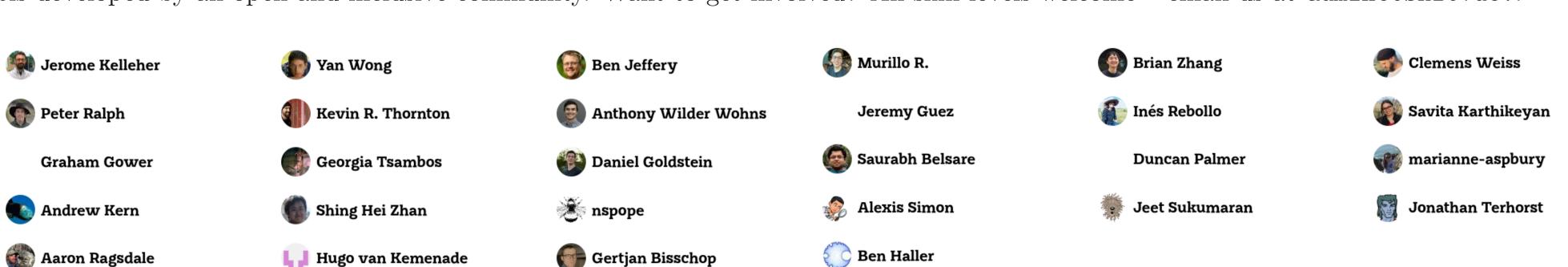
#### Example:

size=(800, 400), canvas\_size=(850, 405), style=style + "".join(node\_styles), y\_axis=True, time\_scale="log\_time", symbol\_size=4.5, y\_label = "Date", x\_label = "SARS-CoV-2 genome position", y\_ticks = y\_ticks, mutation\_labels={}, y\_gridlines=True, node\_labels=node\_labels, root\_svg\_attributes={"id": "ns\_rec"},



## CONTRIBUTORS

tskitis developed by an open and inclusive community. Want to get involved? All skill levels welcome – email us at admin@tskit.dev.



# BIG DATA UNIVERSITY OF INSTITUTE OXFORD





### STATISTICS

imb = pd.DataFrame({

tskit lets you perform efficient calculations of statistics along the genome, often many times quicker than other software! You may be interested in calculating:

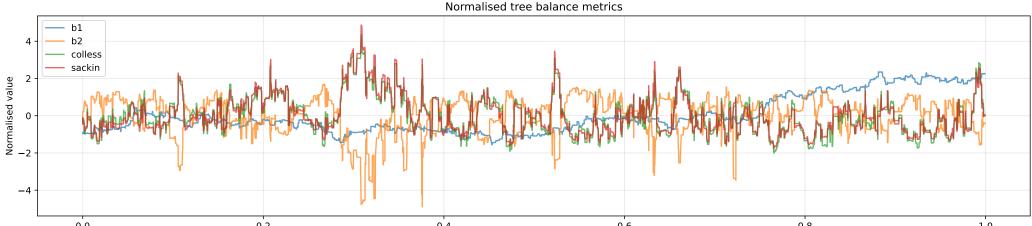
- the allele frequency spectrum or statistics derived from it, like nucleotide diversity, Tajima's D, f4, ...
- IBD-based quantities,
- summaries of tree topology, e.g., genealogical nearest neighbours and tree balance metrics.
- cross-coalescence rates (coming soon!)

Example: newly-implemented tree balance statistics. A balanced (binary) tree is perfectly symmetric in some way: each node has an equally sized subtree descending from its left- and right- branches, where 'size' is determined by some metric involving the tree's nodes and edges. tskit now implements several different metrics to calculate how unbalanced each tree is:

```
"b2" : [t.b2_index() for t in ts.trees()],
    "colless" : [t.colless_index() for t in ts.trees()],
   "sackin" : [t.sackin_index() for t in ts.trees()]
}).set_index("genomic position")
imb = ((imb - imb.mean()) / imb.std())
imb.plot(figsize=(16, 4), alpha=0.7)
```

"genomic position" : [t.interval[0] for t in ts.trees()],

"b1" : [t.b1\_index() for t in ts.trees()],



## NOTABLE NEW FEATURES

tskit's contributors are actively working on new features, bug fixes, and improvements to the usability of existing features. Here's a shortlist of some recent additions:

**Reference sequences** By default, the sites in a tree sequence only define ancestral nucleotides at polymorphism sites. Remaining positions can now be specified using the TreeSequence.reference\_sequence, and individual sample alignments can be obtained with the new TreeSequence.alignments() iterator.

**Structural operations** We've expanded the set of utility functions for large edits on tree sequences. For instance, the TreeSequence.decapitate method removes all parts of a tree sequence that are older than some user-specified time, and TreeSequence.union allows joining together of separate tree sequences to allow parallel simulation across different branches of a phylogenetic tree.

Efficient array access The relationships between nodes in each tree can now be extracted as numpy arrays. When used in conjunction with numba, Python-based calculations on the trees can act as speedily as machinelevel code. Here is pure python computing total branch length just as fast as the "built-in" method (implemented in C):

