# TNet (version 1.0)

## Description

TNet is a phylogeny-based method for reconstructing transmission networks for infectious diseases. It takes as input a phylogeny of the strain (pathogen) sequences sampled from infected hosts and analyzes it to estimate the underlying transmission network. TNet relies on the availability of multiple strain sequences from each sampled host to infer transmissions and is simpler and more accurate than existing approaches. Each run of TNet on the same input tree can result in a different estimate transmission network (constructed based on a single optimal sampled uniformly at random from among all optima), and so TNet should be executed multiple times (say 100) on the input phylogeny and an aggregated transmission network should be constructed from the resulting outputs. The method is parameter-free and highly scalable and can be easily applied within seconds to datasets with hundreds of strain sequences and hosts.

TNet implements algorithms described in the following paper:

> *TNet: Phylogeny-Based Inference of Disease Transmission Networks Using Within-Host Strain Diversity*
> Saurav Dhar, Chengchen Zhang, Ion Mandoiu, Mukul S. Bansal.
> Under review.

TNet is implemented in Python and requires version 3.0 or greater. TNet is freely available open source under GNU GPL.

## Usage

TNet takes as input a single rooted binary phylogeny on all strain sequences sampled from the infected hosts considered in the analysis. This phylogeny must be in Newick format. Such a phylogeny can be constructed using standard phylogeny construction tools such as RAxML or PhyML and then rooting the resulting unrooted phylogeny using standard rooting methods. Each leaf labell in this phylogeny must be of the form <hostID>_<sequenceID>. Only <hostID> is used by TNet.

To use TNet, an input file and an output file must be specified as follows:

```
tnet.py inputFile outputFile
```
or
```
python3 tnet.py inputFile outputFile
```

Each execution of TNet on an input file outputs a single transmission network based on a randomly sampled optimal solution of an underlying computational problem. Thus, TNet should be executed multiple times (say 100 times) on a single input file and results should be aggregated across all output transmission networks. To further improve inference accuracy we also suggest aggregating across multiple bootstrap replicates of the input phylogeny, as done in the paper named above.

## Interpretation of the output

Each execution of TNet on an input file outputs a single transmission network based on a randomly sampled optimal solution of an underlying computational problem. The output file lists all inferred edges (connecting two hosts) in the transmission network. A sample output follows:

```
None  124
124   123
124   122
124   121
124   121
```

In this output, the first line indicates that host 124 is the source of the transmission, the second line indicates that there was a transmission from host 124 to host 123, and so on. Note that some transmission edges may be listed multiple times (e.g., the transmission from 124 to 121 in the sample output above). Any repeated edges can be ignored; alternatively, repeated edges may indicate that multiple distinct pathogen lineages were transmitted during the transmission.

As noted above, each execution of TNet on an input file outputs a single transmission network based on a randomly sampled optimal solution of an underlying computational problem. Thus, TNet should be executed multiple times (say 100 times) on a single input file and results should be aggregated across all output transmission networks. To further improve inference accuracy, we also suggest aggregating across multiple bootstrap replicates of the input phylogeny, as done in the paper named above.

## Example datasets

The “input” directory in the git repository contains a sample rooted phylogeny that can be used as input for TNet.