

Points: 100

## Neighbor Joining Tree Construction

Write a Python program `nj.py` based on the `upgma` code that we gave in class that will read in a lower triangular distance matrix and generate the groupings and lengths of edges of a non-rooted tree that “fits” the data. You must work from the supplied `upgma.py` program. Your program does not need to draw a tree. An machine readable copy of sample data from Felsenstien is here: [../dogbear.dat](#).

To illustrate the format of the input it looks like:

```
Dog Bear Raccoon Weasel Seal SeaLion Cat Monkey
32
48 26
51 34 42
50 29 44 44
48 33 44 38 24
98 84 92 86 89 90
148 136 152 142 142 142 148
```

and is in the same form as for our `upgma`. Feel free to use the `lib515.py` library.

### OUTPUT

For your information, here is the tree for the `dogbear.dat` data in [Newick format](#) with distances following each leaf and internal node:

```
((Bear:6.87500, Raccoon:19.12500):1.75000, ((Seal:12.35000,
  Sealion:11.65000):7.81250, (Weasel:19.56250, (Cat:47.08333,
  Monkey:100.91667):20.43750):1.56250):3.43750, Dog:25.25000);
```

However, your program needs to output exactly three lists:

- The list we called `merge` in the `upgma` example.
- The list we called `size` in the `upgma` example.
- A list of edge pairs (as tuples) for each internal node except the last. These are the distances to each of the subclusters in a cluster. The last element in the list is the final distance between the last two clusters to be merged.

```
[0, 1, 2, 3, 4, 5, 6, 7, (7, 6), (5, 4), (2, 1), (10, 0), (8, 3), (11,
  9), (12, 13)]
[1, 1, 1, 1, 1, 1, 1, 1, 2, 2, 2, 3, 3, 5, 8]
[0, 0, 0, 0, 0, 0, 0, 0, (100.91666666666666, 47.083333333333334),
  (11.6500000000000006, 12.349999999999998), (19.125, 6.875), (1.75,
  25.25), (20.4375, 19.5625), (3.4375, 7.8125), 1.5625]
```

Notice that in order to get the correct output the final step must be to put the final distance on the list of edge pairs and the final cluster and size on their respective lists.

### THE NEIGHBOR-JOINING ALGORITHM

Given a distance matrix  $d$  compute an unrooted tree topology complete with edge lengths that tries to preserve the additive property:  $d_{i,m} + d_{j,m} - d_{i,j} = 2d_{k,m}$  where  $k$  is the first node on both routes from  $i$  and  $j$  to  $m$ .

1. Let the set of clusters be called  $L$  and initially  $i \rightarrow C_i \forall i$  that is  $|C_i| = 1$  and  $L = C_1, C_2, \dots, C_N$ .
2.  $d_{i,j}$  is the distance from the initial distance matrix.
3. Compute “normalized distance matrix”  $D_{i,j}$  for all  $i, j$  such that

$$D_{i,j} = d_{i,j} - (r_i + r_j) \quad \text{where} \quad r_i = \frac{1}{|L| - 2} \sum_{z \in L} d_{i,z}$$

This subtracts the average distance to all other nodes than the pair involved. Note: this is **not** where we use the distance identity.

4. Use normalized distance to find  $(i, j) = \underset{C_i, C_j \in L}{\operatorname{argmin}} D_{i,j}$
5. Merge  $C_i \cup C_j \rightarrow C_k$  where  $k$  is a new cluster number.
6. Mark old clusters as used so that effectively:  $L \leftarrow L - C_i - C_j$
7. Compute a new unnormalized distance matrix including the new cluster  $k$  and excluding  $i, j$ .

$$d_{k,z} = d_{z,k} = \frac{1}{2}(d_{i,z} + d_{j,z} - d_{i,j}) \quad \text{for all } z \in L$$

This uses the additivity of the distances to compute the distance to the new cluster from each other node.

8. Compute the length of the edges from  $k$  to  $i$  and  $j$ . Even though  $C_k$  has assumed the role of both  $C_i$  and  $C_j$  you still need the edge length to  $i$  and  $j$  from  $k$  in order to “draw” the tree.

$$\operatorname{edge}_{i,k} = \frac{1}{2}(d_{i,j} + r_i - r_j), \quad \operatorname{edge}_{j,k} = \frac{1}{2}(d_{i,j} + r_j - r_i)$$

9. Define height  $h_k = d_{i,j}/2$  where  $h_k$  is the height of node that is the ancestor to all in  $C_k$ . When drawing the tree  $h_k$  is the height above the baseline (where all the leaves are).
10.  $L \leftarrow L \cup C_k$
11. While there is more than two clusters left go to step 3
12. Finally, join the remaining two clusters with:

$$\operatorname{edge}_{j,k} = d_{i,j}$$

### IMPLEMENTATION NOTES

Consider this part of the computation:

$$D_{i,j} = d_{i,j} - (r_i + r_j) \quad \text{where} \quad r_i = \frac{1}{|L| - 2} \sum_{z \in L} d_{i,z}$$

The values of  $r_z$  can be computed once each time we want to compute matrix  $D$ . This saves a vast amount of time. Furthermore, since  $D_{i,j}$  is only used to find the argmin of  $D_{i,j}$  we actually don't have to save array  $D$ ; we only need to find the argmin of it. So first compute all the  $r$  and then combine the argmin step with the computation of  $D_{i,j}$ .

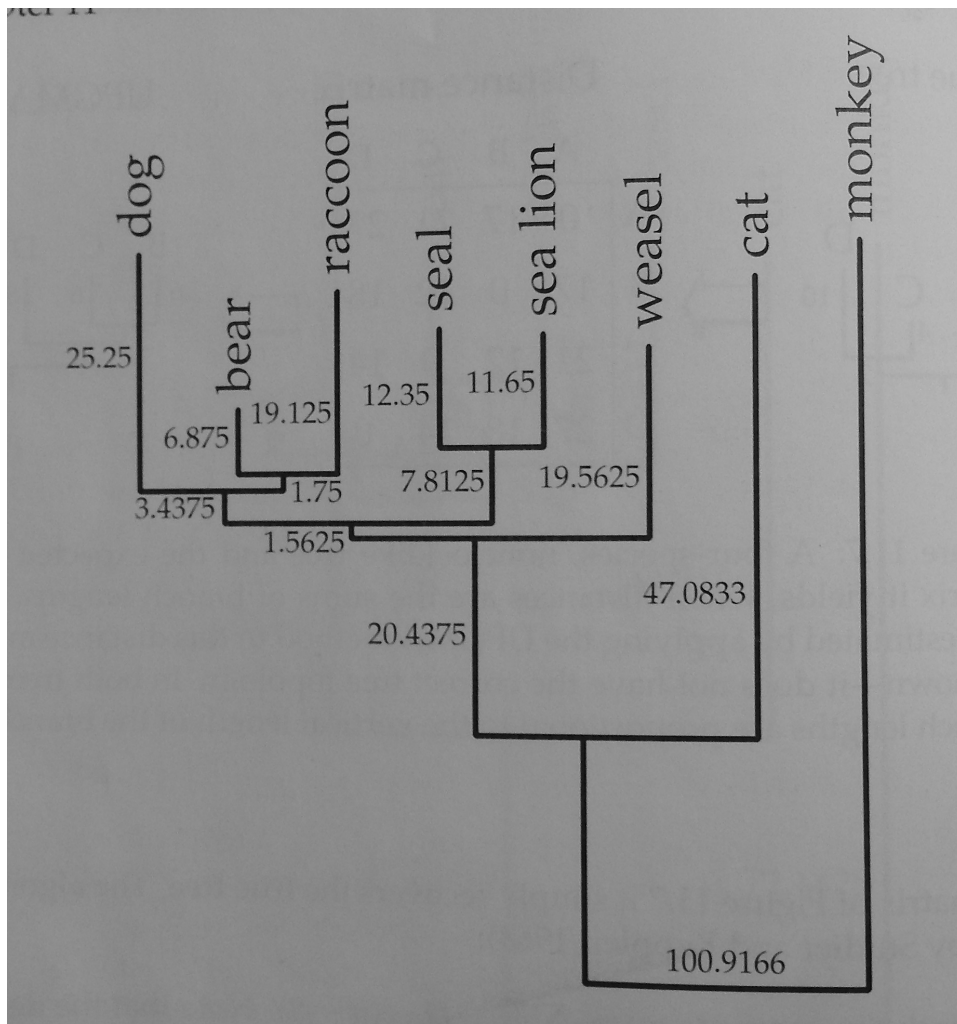


Figure 1: The non-rooted tree for the given problem drawn as a rooted tree.

In the box is an excerpt from the class notes on the algorithm.

### Submission

As always, do your own work. Do not copy from others or from the internet. Using what we did in class is perfectly legal. Submit a single file named `nj.py` to the class submission page. The test script will ignore the spacing you choose and focus on non-whitespace. Remember no late papers and so always turn something in for partial credit.