DateLife Workflows

Luna L. Sanchez Reyes

2019-05-13

Taxon Primates

1. Query source chronograms

There are 526 species in the Open Tree of Life Taxonomy for the taxon Primates. Information on time of divergence is available for 355 of these species across 8 published and peer-reviewed chronograms. Original study citations as well as number of Primates species found across those source chronograms is shown in Table 1. All source chronograms are fully ultrametric and their maximum ages range from 62.766 to 90.4 million years ago (MYA). As a means for comparison, lineage through time plots of all source chronograms available in data base are shown in Fig. 1

2. Summarize results from query

LTT plots are a nice way to visually compare several trees. But what if you want to summarize information from all source chronograms into a single summary chronogram?

The first step is to identify the degree of species overlap among your source chornograms: if each source chronogram has a unique sample of species, it will not be possible to combine them into a single summary chronogram. To identify the set of trees or *grove* with the most source chronograms that have at least two overlapping taxa, we followed Ané et al. 2016. In this case, not all source chronograms found for the Primates have at least two overlapping species. The largest grove has 2 chronograms (out of 8 total source chronograms).

Now that we have identified a grove we can go on to summarize it by translating the source chronograms into patristic distance matrices and then averaging them into a single summary matrix; yes, this first step is that straightforward. We can average the source matrices by simply using the mean or median distances, or we can use methods that involve transforming the original distance matrices—such as the super distance matrix (SDM) approach of Criscuolo et al. 2006—by minimizing the distances across source matrices. As a result of such transformation, an SDM summary matrix can contain negative values. But, the SDM summary matrix of this taxon has no negative values.

Because our summary matrix is basically a distance matrix, a distance-based clustering algorithm could be used to reconstruct the tree. Algorithms such as neighbour joining (NJ) and unweighted pair group method with arithmetic mean (UPGMA) are fast and work very well when there are no missing values in the matrices. However, summary matrices coming from source chronograms usually have several NAs and missing rows. When this happens, variants of traditional clustering algorithms have been developed to deal with missing values. However, even these methods do not work well with our summary matrices, as shown in the following section. We should note that these clustering methods are usually applied to distance matrices representing substitution rates and not absolute time.

2.1. Clustering a summary matrix

NJ, UPGMA, BIONJ, minimum variance reduction (MVR) and the triangle method (TM) algorithms were used to cluster median and SDM summary distance matrices. All clustering algorithms returned very

similar trees with both types of summary matrices (Fig. 2, Appendix Fig. 5). UPGMA is the only algorithm that returns ultrametric trees, but they are considerably older than expected from source chronograms. The other methods returned trees with reasonable ages, but that are not ultrametric. An alternative to clustering algorithms is to use all data avilable in the summary matrix as calibrations over a consensus tree.

2.2. Calibrating a consensus tree with data from a summary matrix

Even if the branch lengths coming form the clustered chronograms are not adequate, the topology can still be used as a backbone tree that can be dated using data from the summary matrix as secondry calibrations. A summary of divergence times available for each node can be obtained from the summary matrix, simply by getting the nodes from the backbone tree that correspond to each pair of taxa in the matrix. Finally, this summary of node divergence times can be used with the consensus tree as input in any dating software that does not require data. The branch length aduster (BLADJ) algorithm [@Webb2000] is really fast and does not make any evolutionary assumptions on age distribution. Other software such as MrBayes and r8s can be used instead of BLADJ by running them without data. In here, we show summary chronograms obtained using minimum, mean and maximum distances from the summary of node divergence times of the backbone tree as fixed ages in BLADJ (Fig. 3). Summary chronograms from both types of summary matrices are quite similar. As expected, SDM chronograms using minimum, mean and maximum distances do not vary much in their maximum age, because ages are transformed to minimize the variance. In contrast, the median chronograms obtained with minimum, mean and maximum distances have wider variation in their maximum ages, as can be observed in the distance between the green arrows in Fig. 3.

3. Generate new chronograms

Another way to leverage information from the source chronograms is to use the node ages as secondary calibration points to date any tree topology (with or without branch lengths) given that at least two taxa from source chronograms are in the tips of that topology. In this data set we have 2251 calibrations in total (that basically corresponds to the sum of the number of nodes from each source chronogram). Once we have a target tree topology, we can map the calibrations to the target tree. Some nodes will have several calibrations and some others might have none. Also, some node ages can be conflicting, with descendant nodes being older than parent nodes. We performed a series of cross validation analyses with different dating methods by dating the topologies of each source chronogram using information from all other source chronograms as calibration points.

3.1. Calibrate a backbone tree without branch lengths

To date a tree in the absence of data on relative evolutionary rates (molecular or morphological) we follow the same methodology as the one used to obtain summary chronograms. First, we obtained the nodes that correspond to each pair of taxa in the data set of total calibrations to construct a summary of node calibrations for the backbone tree. Then, we used mean ages as secondary calibrations for the backbone tree with the software BLADJ. In general, the time of divergence information from other source chronograms allows to recover the divergence times from the original study. In some cases, it is evident that information from a particular study really affects the summary of divergence times. In some other cases, the root of the tree is not calibrated. Since BLADJ has no underlying model of evolution, there is no way for the algorithm to calculate this age. So, we simply added a unit of the mean difference across ranked ages from secondary calibrations (Fig. 4)

3.2. Calibrate a tree with data

If you have atree with branch lengths proportional to relative substitution rates, you can use PATHd8 You can get data on relative substitution rates from BOLD using DAteLife's functions. To deal with conflicting calibrations, we can expand them to make them agree, or we can summarize them.

3.2.1. Expanding calibrations

3.2.2. Summarizing calibrations

4. Example with subspecies tree

As an example, we're gonna date the subspecies tree of the group (coming from otol).

Now, let's say you like the Open Tree of Life Taxonomy and you want to stick to that tree. Dates from available studies were tested over the Open Tree of Life Synthetic tree of Primates and a tree was constructed, but all branch lengths are NA. We also tried each source chronogram independently, with the Dated OToL and with each other, as a form of cross validation in Table 2. This is not working perfectly yet, but we are developping new ways to use all calibrations efficiently.

5. Simulate data/ Add missing taxa

An alternative to generate a dated tree from a set of taxa is to take the available information and simulate into it the missing data. We will take the median and sdm summary chronograms to date the Synthetic tree of Life:

Tables and Figures

Table 1: Primates source chronogram studies information.

	Citation	Source N	Taxon N
1.	Bininda-Emonds, Olaf R. P., Marcel Cardillo, Kate E. Jones, Ross D. E. MacPhee,	3	215/526
	Robin M. D. Beck, Richard Grenyer, Samantha A. Price, Rutger A. Vos, John L.		
	Gittleman, Andy Purvis. 2007. The delayed rise of present-day mammals. Nature		
	446 (7135): 507-512		
2.	Hedges, S. Blair, Julie Marin, Michael Suleski, Madeline Paymer, Sudhir Kumar.	1	294/526
	2015. Tree of life reveals clock-like speciation and diversification. Molecular Biology		
	and Evolution 32 (4): 835-845		
3.	Springer, Mark S., Robert W. Meredith, John Gatesy, Christopher A. Emerling,	4	330/526
	Jong Park, Daniel L. Rabosky, Tanja Stadler, Cynthia Steiner, Oliver A. Ryder, Jan		
	E. Janečka, Colleen A. Fisher, William J. Murphy. 2012. Macroevolutionary		
	dynamics and historical biogeography of primate diversification inferred from a		
	species supermatrix. PLoS ONE 7 (11): e49521.		

 $Source\ N$: Number of source chronograms reported in study.

Taxon N: Number of queried taxa found in source chronograms.

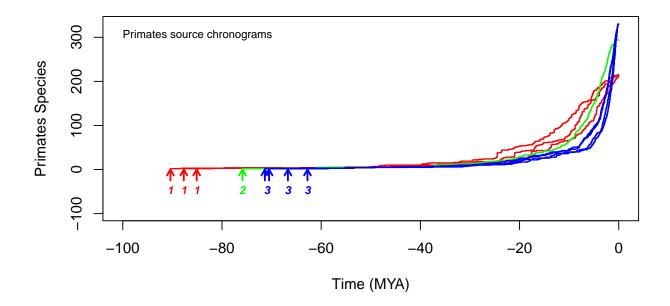


Figure 1: Lineage through time (LTT) plots of source chronograms available in data base for species in the Primates. Numbers correspond to original studies in Table 1. Arrows indicate maximum age of each chronogram.

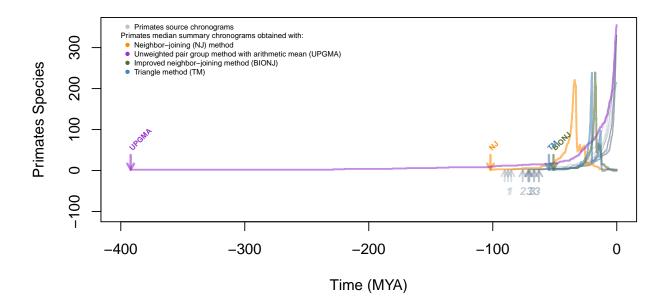


Figure 2: Lineage Through Time plots of Primates median summary chronograms obtained with different clustering algorithms. Not all algorithms worked with this summary matrix and we are only showing here the ones that worked. Chronograms obtained from the SDM summary matrix are very similar to the ones from the median summary matrix with all clustering algorithms (Appendix Fig. 5).

References

Appendix

The following species were completely absent from the chronogram data base: Alouatta arctoidea, Alouatta discolor, Alouatta stramineus, Alouatta ululata, Aotus azarae, Aotus jorgehernandezi, Aotus zonalis, Avahi mooreorum, Avahi ramanantsoavani, Cacajao rubicundus, Callicebus aureipalatii, Callicebus baptista, Callicebus barbarabrownae, Callicebus caquetenesis, Callicebus caquetensis, Callicebus discolor, Callicebus lucifer, Callicebus medemi, Callicebus melanochir, Callicebus miltoni, Callicebus ornatus, Callicebus pallescens, Callicebus regulus, Callicebus stephennashi, Callicebus toppinii, Callicebus urubambensis, Callicebus vieirai, Callithrix cf. emiliae, Callithrix chrysoleuca, Carlito syrichta, Cebus aequatorialis, Cebus brunneus, Cebus cesarae, Cebus cuscinus, Cebus imitator, Cebus leucocephalus, Cebus malitiosus, Cebus polykomos, Cebus unicolor, Cebus versicolor, Cebus yuracus, Cephalopachus bancanus, Cercocebus lunulatus, Cercocebus sanjei, Cercopithecus denti, Cercopithecus doggetti, Cercopithecus kandti, Cercopithecus lomamiensis, Cercopithecus lowei, Cheirogaleus andysabini, Cheirogaleus lavasoensis, Cheirogaleus minusculus, Cheirogaleus thomasi, Cheracebus lugens, Cheracebus purinus, Cheracebus torquatus, Chiropotes utahickae, Chlorocebus djamdjamensis, Daubentonia robusta, Euoticus matschiei, Galagoides cocos, Galagoides orinus, Galagoides rondoensis, Galagoides thomasi, Hylobates entelloides, Hylobates funereus, Lemur indri, Lemur tardigradus, Lemur volans, Lepilemur grewcockorum, Lepilemur hollandorum, Lepilemur jamesorum, Lepilemur mitsinjoensis, Lepilemur scottorum, Lepilemur tymerlachsoni, Lophocebus johnstoni, Lophocebus opdenboschi, Lophocebus osmani, Lophocebus uqandae, Macaca balantak, Macaca leucogenys, Macaca speciosa, Mico acariensis, Mico intermedius, Mico leucippe, Mico marcai, Mico nigriceps, Microcebus lokobensis, Microcebus marchita, Microcebus myonixus, Microcebus tanosi, Nomascus annamensis, Nycticebus bancanus, Nycticebus borneanus, Nycticebus kayan, Papio japonicus, Papio kindae, Phaner electromontis, Phaner parienti, Piliocolobus bouvieri, Piliocolobus epieni, Piliocolobus oustaleti, Piliocolobus parmentieri, Piliocolobus semlikiensis, Piliocolobus temminckii, Piliocolobus waldronae, Pithecia cazuzai, Pithecia chrysocephala, Pithecia hirsuta, Pithecia inusta, Pithecia isabela, Pithecia milleri, Pithecia mittermeieri, Pithecia napensis, Pithecia pissinattii, Pithecia rylandsi, Pithecia vanzolinii, Plecturocebus bernhardi, Plecturocebus brunneus, Plecturocebus caliqatus, Plecturocebus cinerascens, Plecturocebus cupreus, Plecturocebus donacophilus, Plecturocebus hoffmannsi, Plecturocebus miltoni, Plecturocebus moloch, Presbutis bicolor, Presbutis canicrus, Presbutis mitrata, Presbutis natunae. Presbytis sabana, Presbytis senex, Presbytis siamensis, Presbytis siberu, Presbytis sumatrana, Propithecus candidus, Pseudopotto martini, Pygathrix cinerea 1 RL-2012, Pygathrix cinerea 2 RL-2012, Rhinopithecus bieti 1 RL-2012, Rhinopithecus bieti 2 RL-2012, Saguinus cruzlimai, Saguinus illigeri, Saguinus lagonotus, Saquinus leucogenys, Saquinus nigrifrons, Saquinus pileatus, Saquinus ursulus, Saquinus weddelli, Saimiri cassiquiarensis, Saimiri macrodon, Sapajus apella, Sapajus cay, Sapajus flavius, Sapajus libidinosus, Sapajus nigritus, Sapajus xanthosternos, Sciurocheirus cameronensis, Sciurocheirus makandensis, Semnopithecus ajax, Semnopithecus hypoleucos, Semnopithecus schistaceus, Tarsius banacanus, Tarsius fuscus, Tarsius pelengensis, Tarsius tarsius, Tarsius tumpara, Trachypithecus ebenus, Trachypithecus mauritius, Trachypithecus selangorensis, Trachypithecus shortridgei

Dated induced subtree could not be obtained for the Primates.

This taxon's SDM matrix has NO negative values. This taxon's Median matrix has NO negative values.

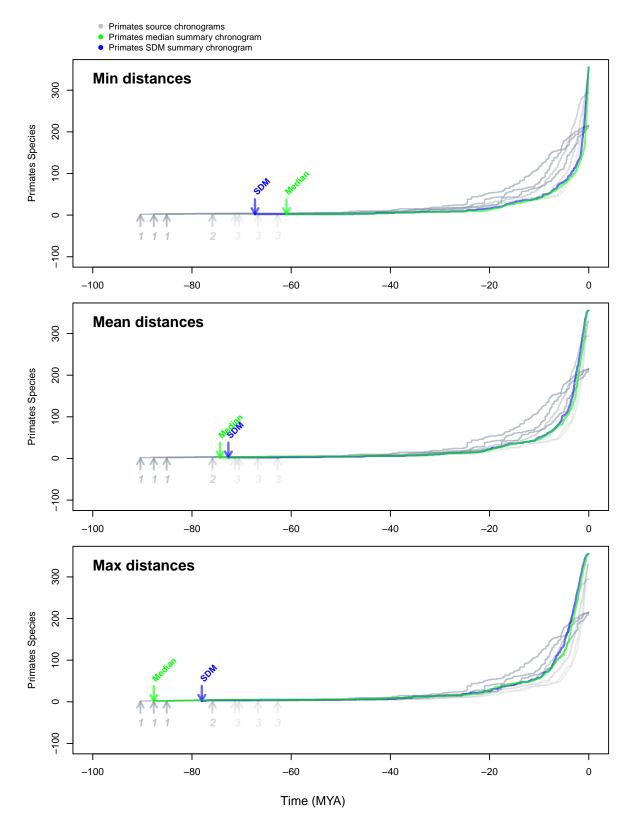
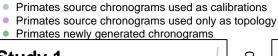


Figure 3: Primates lineage through time (LTT) plots from source chronograms (gray), median (green) and SDM (blue) summary chronograms obtained by calibrating a consensus tree tropology with distance data from respective summary matrices and then adjusting branch lengths with BLADJ.



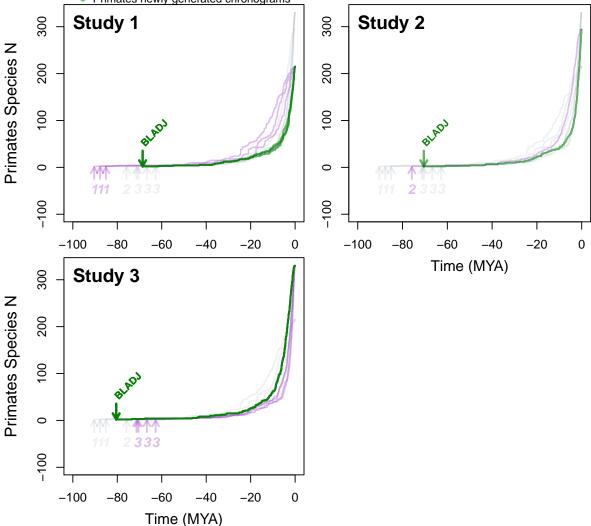


Figure 4: Primates lineage through time (LTT) plots from source chronograms used as secondary calibrations (gray), source chronograms used as topology (purple) and chronograms resulting from calibrating the latter with the former, using BLADJ (green).

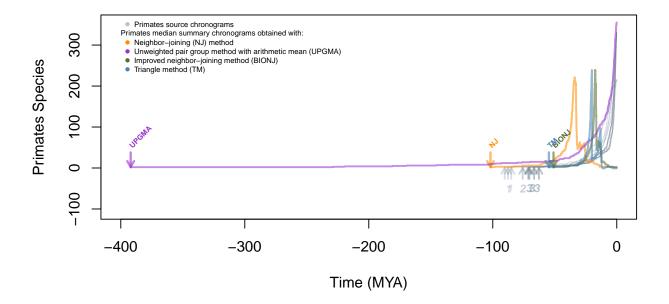


Figure 5: Lineage Through Time plots of Primates SDM summary chronograms obtained with different clustering algorithms. Not all algorithms worked with the SDM summary matrix and we are only showing here the ones that worked. Chronograms obtained from the median summary matrix are very similar to the ones shown here with all algorithms (mainFig. 2).

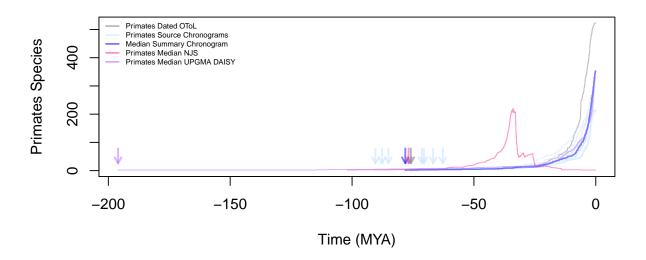


Figure 6: Primates lineage through time (LTT) plots from source chronograms and Median summary matrix converted to phylo with different methods (NJ and UPGMA). Clustering algorithms used often are returning non-ultrametric trees or with maximum ages that are just off (too old or too young). So we developed an alternative algorithm in datelife to go from a summary matrix to a fully ultrametric tree.

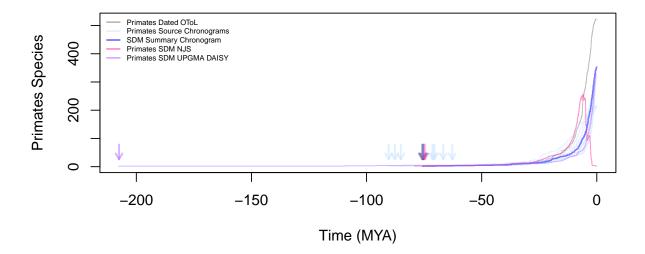


Figure 7: Primates lineage through time (LTT) plots from source chronograms and SDM summary matrix converted to phylo with different methods (NJ and UPGMA). Clustering algorithms used often are returning non-ultrametric trees or with maximum ages that are just off (too old or too young). So we developed an alternative algorithm in datelife to go from a summary matrix to a fully ultrametric tree.

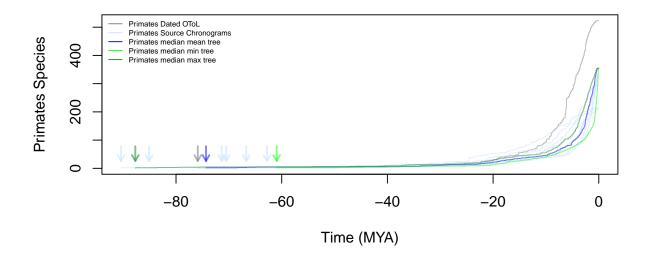


Figure 8: Primates lineage through time (LTT) plots from source chronograms and Median summary matrix converted to phylo with datelife algorithm.

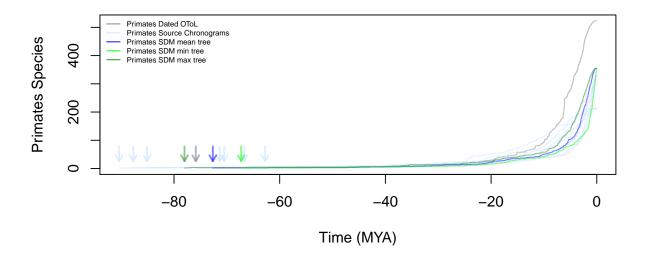


Figure 9: Primates lineage through time (LTT) plots from source chronograms and SDM summary matrix converted to phylo with datelife algorithm.