Phylogenetics

ANGES: reconstructing ANcestral GEnomeS maps

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1 INTRODUCTION

The reconstruction of the organization of ancestral genomes is a long-standing problem, motivated by several applications in comparative genomics (Pennisi, 2005). It has received a renewed interest during the past few years, due, among others, to the increasing number of sequenced and assembled genomes and major methodological advances. Two general approaches have been followed by recent methods. The global parsimony aims at computing ancestral gene orders that minimize the number of genome rearrangements; following the pioneering work of Sankoff et al., it can now address hard problems of ancestral genome reconstruction (Zheng and Sankoff, 2012). The local parsimony approach that was pioneered in Ma et al. (2006) follows principles used in computing physical maps of extant genomes and was explored in several recent articles (Chauve and Tannier, 2008; Chauve et al., 2010; Ma et al., 2006; Muffato et al., 2011).

In a series of recent articles, we followed the latter approach and described different methods for computing ancestral genome maps from the comparison of extant species. All methods are centered on variants of a combinatorial framework widely used to compute physical maps of extant genomes, the Consecutive-Ones Property (Chauve and Tannier, 2008; Chauve et al., 2010; Gavranovic et al., 2011). We implemented this framework for the reconstruction of animal, fungi and plant ancestral genomes and showed its validity on well-accepted ancestral genomes as well as its robustness to parameter change.

In the present note, we describe ANGES, a suite of Python programs implementing and extending several of these methods, to reconstructing both eukaryotic and prokaryotic ancestral genome maps. It is currently the only available method able to handle a wide range of genetic or genomic data on any domain of life.

2 METHODOLOGICAL OVERVIEW

ANGES requires two kinds of data as input: a ‘species tree’, annotated to indicate an ancestral species (the ancestor), and a set of homologous ‘markers’ families.

The markers are genomic segments that are present in the current extant species. A family describes an ancestral marker, which is assumed to have been present in the ancestral genome, in single copy and with no overlap between different markers. They can be obtained through whole-genome alignments to detect families of genomic segments that evolved only through limited local rearrangements, as in Chauve and Tannier (2008), Gavranovic et al. (2011) and Ma et al. (2006) or using gene families as in Muffato et al. (2010) and Chauve et al. (2010). ANGES computes the organization of these markers into chromosomal segments of the ancestor, i.e. an ancestral genome map. ANGES proceeds in two stages:

(1) For each pair of species whose evolutionary path in the species tree contains the ancestor, ANGES detects genomic segments with a similar organization (in terms of markers) in both compared species. It then derives ‘Ancestral Contiguous Sets’ (ACS), which are sets of ancestral markers that should be contiguous in an ancestral genome. Each ACS is given a weight according to the pattern of occurrence of its support in the considered extant species.

(2) ANGES organizes the markers into linearly or circularly ordered ancestral chromosomal segments called ‘Contiguous Ancestral regions’ (Ma et al., 2006), by extracting a subset of ACS that satisfies a variant of the Consecutive-Ones Property (the CIP).

This general approach can be implemented in several ways, and ANGES implements a wide range of variations.

ACS can be of two types. Adjacencies are sets of two markers that are believed to be contiguous in the ancestral genomes. Common intervals are sets of two or more markers that are also believed to be contiguous in the ancestral genomes.

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Robustness studies (Ouangraoua et al., 2011; Schmidt and Stoye, 2004). Markers missing in some extant genomes, due for example to gene loss, are a major issue, especially when considering a large number of extant genomes (Gavranovic et al., 2011). To address this issue, ANGES implements several methods to post-process ACS in order to account for missing markers, which are described in detail in the ANGES manual. ANGES can, for example, handle whole-genome duplications, possibly followed by massive gene loss (Chauve et al., 2010; Gavranovic et al., 2011).

Ancestral genome maps are then computed from the set of ACS by selecting a subset of these ACS that satisfies the CIP (for multichromosomal linear genomes) or the circular CIP (for unichromosomal circular genomes). ANGES computes such subsets of ACS using a greedy heuristic defined in Ma et al. (2006), a branch-and-bound algorithm or a spectral seriation algorithm (Atkins et al., 1998) (the latter method is a new feature that was not described in our previous articles). Ancestral genome maps are represented by the PQ-tree data structure [multichromosomal linear genomes, see Chauve and Tannier, 2008] or the related PC-tree (circular genomes). The ability to compute circular ancestral genome is an important new feature of ANGES with respect to our previous works that considered only eukaryotic ancestral genomes. ANGES permits the use of CIP variants, such as the sandwich-CIP (Gavranovic et al., 2011), by translating ACS into a correlation matrix that is used as input to a spectral seriation algorithm or the CIP with multiplicity that can include information about telomeric ACS (Chauve et al., 2011).

The methods implemented by ANGES are predictive and in most cases cannot be assessed against true results, as ancestral genomes at the evolutionary scale of tens or hundred millions of years are unavailable. Some specific ancestors, regarded as references by the genomics and cytogenetics communities, have been used to assess the accuracy of ancestral genome reconstruction methods, such as the boreoeutherian or saccharomyces ancestors [see (Chauve and Tannier, 2008; Chauve et al., 2010)]. Comparison with simulations (Ma et al., 2006), robustness studies (Ouangraoua et al., 2011) or statistical support of local features (Ma et al., 2006) has also been reported. Probably more useful, the number of discarded ACS to satisfy the CIP is a good indicator of the conflicting nature of the conserved synteny signal that defines ACS. For all the ancestors reported using the methods implemented in ANGES, we could observe a very low conflicting signal with well under 5% of ACS discarded.

3 IMPLEMENTATION
The general principle of our implementation is that every computational step is implemented into a single Python script, which takes as input a set of text files and computes, as output, a set of text files. Hence, the Python scripts that compose ANGES can be organized into a complete ancestral genome computation pipeline or can be used as stand-alone scripts. ANGES contains efficient implementations for generic tasks in comparative genomics such as computing common intervals and checking several variants of the CIP on binary matrices. In some cases, it is, as far as we know, the first available implementation of recent algorithms (Bergeron et al., 2008; Chauve et al., 2011; McConnell, 2004).

To use the ANGES scripts into an ancestral genome map computation pipeline, ANGES contains a ‘master script’ that reads the input files (markers and species tree, as well as possible optional files) and a ‘parameters file’ that records all user-defined choices that can be made regarding data processing. A simple graphical interface is provided to generate, read and modify parameters files.

Running times can vary depending on the nature of data and the ancestral map computation methods. The examples provided with the distribution have all completed within a few minutes on a bi-processor desktop computer. Larger datasets we analyzed, especially with non-unique markers or using the branch-and-bound method, required up to 2 days of computation.

4 FUTURE WORK
In the next release, we will include alternative methods to compute ancestral genome maps from ACS, and in particular an extension of the greedy heuristic for the sandwich-CIP framework. We will also include methods to compute ancestral genome structures that do not rely on the notion of contiguity but on the less rigid notion of conserved synteny, as described in Ouangraoua et al. (2011). And we are working on allowing multichromosomal genomes with circular and linear chromosomes, to be able to handle in a single framework all known genome forms in the living world. We are also integrating the possibility to provide gene families trees as input.

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