Identifying a species tree subject to random lateral gene transfer

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HIGHLIGHTS

- Lateral gene transfer (LGT) can mislead phylogenetic tree inference.
- Exact calculations show a zone of weak statistical inconsistency.
- If LGT rates are low enough then tree inference is still possible.
- Consistency can hold even if each gene is transferred many times on a large tree.
- A tree inference method is described and applied to two data sets.

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ABSTRACT

A major problem for inferring species trees from gene trees is that evolutionary processes can sometimes favor gene tree topologies that conflict with an underlying species tree. In the case of incomplete lineage sorting, this phenomenon has recently been well-studied, and some elegant solutions for species tree reconstruction have been proposed. One particularly simple and statistically consistent estimator of the species tree under incomplete lineage sorting is to combine three-taxon analyses, which are phylogenetically robust to incomplete lineage sorting. In this paper, we consider whether such an approach will also work under lateral gene transfer (LGT). By providing an exact analysis of some cases of this model, we show that there is a zone of inconsistency when majority-rule three-taxon gene trees are used to reconstruct species trees under LGT. However, a triplet-based approach will consistently reconstruct a species tree under models of LGT, provided that the expected number of LGT transfers is not too high. Our analysis involves a novel connection between the LGT problem and random walks on cyclic graphs. We have implemented a procedure for reconstructing trees subject to LGT or lineage sorting in settings where taxon coverage may be patchy and illustrate its use on two sample data sets.

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1. Introduction

Phylogenetic trees inferred from different genes often suggest different evolutionary histories for the species from which they have been sampled. This problem of gene tree ‘incongruence’ is widely recognized in molecular systematics (Cranston et al., 2009; Galtier and Daubin, 2008). It is particularly relevant to the question of the extent to which the history of life on earth can be represented by a phylogenetic tree, rather than a complex network of reticulate evolutionary events, such as species hybridization, lateral gene transfer (LGT) and endosymbiosis. There are several well-recognized causes of gene tree incongruence, most of which apply even in the absence of reticulate evolution, and we begin by discussing these.

Firstly, there is always an expected amount of disagreement under any model of tree-based Markovian evolution, simply due to random sampling effects (i.e. the sequences are of finite rather than infinite length); moreover, this effect becomes magnified as branches in the tree become very short, or very long (Martyn and Steel, 2012). Furthermore, regardless of how much data one has, certain tree reconstruction methods may exhibit systematic errors, due to phenomena such as long branch attraction, or where the model assumed in the analysis differs significantly from the process that generated the data (‘model mis-specification’) (Felsenstein, 2004).

A second basic reason for gene trees to differ in topology from the underlying species tree is the population-genetic phenomenon of incomplete lineage sorting. As one traces the history of a gene sampled from different extant species back in time, the
resulting tree of coalescent events can differ from the species tree within which these lineages lie. Recent theoretical work (Rosenberg, 2002) based on the multi-species coalescent has shown that the most probable gene tree topology can differ from the species tree topology, when the number of taxa is greater than three. By contrast, it has long been known that for triplets, the matching topology is the most probable topology (Nei, 1987; Tajima, 1983). Gene tree discordance due to incomplete lineage sorting is well-established in many data sets, and has been proposed as an explanation for why, for example, up to 30% of the gene trees in the tree (humans, chimpanzee, gorilla) do not support this species relationship (Hobolt et al., 2007). Further recent work on lineage sorting has investigated the statistical consistency of building species trees from gene trees (or the clades they contain) according to various consensus criteria (Allman et al., 2011; Degnan et al., 2009). Additional reasons for gene tree discordance that are still consistent with a species tree are gene duplication and loss, and recombination.

If we turn now to reticulate evolution, it helps to distinguish between two types: hybridization (which will include, for example, endosymbiosis, the transfer of a sizable percentage of the genome of one species into another or the combination of two genomes into a larger genome) and LGT (which is widespread in bacteria and includes the transfer of one or a small number of genes from one organism to another). In the case of hybridization, it is clear that no single tree can adequately describe the evolution of the taxa under study, and that a network (or a set of species trees) is usually a more appropriate representation.

Hybridization will lead to gene tree incongruence but it leaves a statistically different signature to the processes of lineage sorting or sampling error discussed above. For example, in the case of lineage sorting, for a given triplet of taxa, we expect that one of the two topologies will be well supported, and the other two topologies will have lower but approximately equal support. On the other hand, under hybridization, we expect to find support for two of the three topologies that reflect the hybridization event but little support for the third. A number of authors have explored this question of distinguishing hybridization from lineage sorting (Chung and Ané, 2011; Holland et al., 2008; Holder et al., 2001; Joly et al., 2009; Yu et al., 2012).

The second type of reticulate evolution, LGT, is the main concern of this paper, and is particularly relevant for prokaryotic evolution (Dagan and Martin, 2007; Doolittle, 1999; Jain et al., 1999). A fundamental and much-debated question is whether a species tree can be reconstructed from gene trees if genes are randomly transferred between the lineages of the tree (Abby et al., 2012; Szöllő et al., 2012). One viewpoint holds that if the vast majority of genes have been transferred during their history then few gene tree topologies will agree with any species tree topology, so it makes little sense to talk about a single species tree (Bapteste et al., 2005; Dagan and Martin, 2006). An alternative view is that one can still recover statistical support for a central species tree even in the presence of relatively high rates of LGT (cf. Abby et al., 2012; Roch and Snir, 2013; Szöllő et al., 2012), particularly if these transfers occur in a mostly random (rather than concerted) fashion. The debate is complex because extensive gene tree incongruence does not, by itself, provide strong evidence against a underlying species tree, including the phylogenetic distance between them. In Roch and Snir (2013), the most recent of these papers, the authors establish a strong transition result, which shows how the species tree can be reconstructed from a given (logarithmic) number of gene trees, provided that the expected number of LGT events lies below a certain threshold; above this threshold, it becomes impossible to distinguish the underlying species tree from alternative trees based only on the given gene trees. Our results are complementary to this work, as our interest is more in the statistical consistency of species tree reconstruction and, in particular the consistency of tree reconstruction of triplets in a larger tree.

In our paper we begin by setting up some definitions to formally describe the way in which an arbitrary sequence of LGTs on a species tree determines the topology of the associated gene tree. We consider the combinatorial aspects of this process for any given triplet of leaves. We then introduce the model of random LGT events from Linz et al. (2007) along with an extension to allow the rates of LGT to vary with time and with phylogenetic distance. Under these models, we provide an exact analysis of this model on three-taxon and four-taxon trees, showing that for three-taxon trees, the species topology always has strictly higher probability than the other two competing topologies, but for certain four-taxon trees there is a zone of (weak) statistical inconsistency. In Section 6, we consider trees with an arbitrary number of leaves and establish a sufficient condition for statistically consistent species tree reconstruction from gene trees. Essentially, this condition is an upper bound on the expected number of transfers of certain types in the tree.

We then discuss how estimates of the rate of LGT in a large tree could impact on this analysis, and consider the difficult problem of reconstructing a tree when the topology of each gene tree can be influenced by both LGT and incomplete lineage sorting processes. Finally, we describe and illustrate a simple algorithm for reconstructing a species tree from gene trees which may have patchy taxon coverage; here tree reconstruction is statistically consistent if each gene evolves under the random model of LGT or incomplete lineage sorting, provided the rate of LGT is sufficiently low. We end with a brief discussion and some questions for further work.

2. Combinatorial LGT analysis

2.1. Definitions

Throughout this paper X will denote a set of species of size n, and A will denote a subset of X of size 3. Consider a rooted phylogenetic ‘species’ tree T, with leaf set X, and a vertex ρ. We will regard T as a 1-dimensional simplicial complex (i.e. the edges as intervals) so each ‘point’ p in T is either a vertex or an element of the interval that corresponds to an edge. Consider a coalescence time scale: t : T → [0, ∞) of the tree with the coalescence time increasing into the past. Then

- t(ρ) = 0 ⇔ ρ is a leaf,
- if u is a descendant of υ then t(u) < t(υ).

We refer to t(p) as the t-value of p and to t(ρ) as the time span of the tree (the time from the present to the most recent common ancestor (MRCA) of all the species in X).

A lateral gene transfer (LGT) on T (or, more briefly a transfer event) is an arc from p ∈ T to ρ ∈ T where p and ρ are contemporaneous, i.e. t(p) = t(ρ). We will also assume that neither p nor ρ are vertices of T (i.e. transfers go between points on the edges of the tree).

We write e = (p, ρ) to denote this transfer event and we write t(σ) for the common value of t(p) and t(ρ). We will assume that no two transfer events occur at exactly the same time.
Let $\sigma = \sigma_1, \ldots, \sigma_k$ be a sequence of transfer events ($\sigma_i = (p, p_i)$) arranged in increasing $t$-value order (thus $\sigma_1$ is the most recent transfer event and $\sigma_k$ is the most ancient; moreover, the total ordering is well defined by the assumption that no two transfer events took place at the same date). Thus, we will always assume in what follows that:

$$0 < t(\sigma_1) < t(\sigma_2) < \cdots < t(\sigma_k) < t(p).$$

We refer to $\sigma$ as a transfer sequence on the species tree $T$. In the biological context, we view $\sigma$ as describing the transfer history of a particular gene, so different genes will have different associated transfer sequences (including, possibly, the empty transfer sequence, if no transfer events occur on $T$).

Given a species tree $T$ with the leaf set $X$ and a transfer sequence $\sigma = \sigma_1, \ldots, \sigma_k$ on $T$, we obtain an associated gene tree $T[\sigma]$. To describe this tree more precisely, we assume, as in Linz et al. (2007), that an LGT arc from point $p$ to $p'$ replaces the gene that was present on the edge at $p'$ with the transferred gene from $p$. Thus, if we trace the history of the gene from the present to the past (i.e., in increasing coalescence time), each time we encounter an incoming horizontal arc into this edge we follow this arc (against the direction of the arc). In this way, the species tree, along with any sequence of LGTs describes an associated gene tree. We can formalize this by deleting the interval of this 1-complex immediately above $p_i$ and consider the minimal connected subgraph of the resulting complex that contains $X$. Call this tree $T[\sigma]$. An example to illustrate these concepts is presented in Fig. 1.

It will also be helpful to describe the tree $T[\sigma]$ as the final tree in a sequence, where each tree is obtained from its predecessor by applying the the next LGT transfer, as follows.

- Given the pair $T, \sigma = \sigma_1, \ldots, \sigma_k$, define the following sequence of derived X-trees, for $0 \leq r \leq k$:
  $$T_0 = T, \quad T_r = T_{r-1}[\sigma_r].$$
  Thus, $T_k = T[\sigma]$, and for $1 \leq r < k$, $T_r = T[\sigma_1, \ldots, \sigma_r]$ is the gene tree obtained from $T$ by performing just the $r$ most recent transfers. For example, for the tree $T$ shown in Fig. 1, $T_1 = T_0[\sigma_1]$ and $T_2 = T_1[\sigma_2]$ have the same topology as $T$, but $T_3 = T_2[\sigma_3]$ has the topology shown in Fig. 2.
- Given $T' \in \{T_0, T_1, \ldots, T_k\}$, a point $p \in T'$ and any non-empty subset $Y$ of $X$, let $\text{des}_T(T', p)$ denote the subset of $Y$ whose elements are the descendants of $p$ (i.e., which become separated from the root of $T'$ if $p$ is deleted).

### 3. Triplet analysis

Consider a species tree $T$ on $X$. For the rest of this paper, we will assume that $T$ is binary (i.e. fully resolved). Let $A = \{a, b, c\}$ be a subset of $X$ of size 3. We write $T[A]$ to denote the binary phylogenetic tree on leaf set $A$ that is induced by restricting the leaf set of $T$ to $A$ (Semple and Steel, 2003). We refer to $T[A, b, c]$ as a triplet (species tree) topology, and we write $a|bc$ to denote the triplet topology in which the root of this tree separates leaf $a$ from the pair $b, c$.

#### 3.1. Key triplet definitions

For a sequence $\sigma$ of transfer events, we say that:

- $\sigma$ induces a match for $A = \{a, b, c\} \subseteq X$ if the species tree $T$ and its associated gene tree $T[\sigma]$ resolve $a$, $b$, $c$ as the same three-taxon tree; i.e., if $T[\sigma]|A = T|A$.
- Otherwise, if $T[\sigma]|A$ is one of the other two rooted binary tree topologies on leaf set $A$, we say that $\sigma$ induces a mismatch for $A$.

An example is provided in Fig. 3.

For a transfer event $\sigma = (p, p')$, we say that $\sigma$ is into an $A$-lineage if $\text{des}_T(T, p)$ is a single element of $A$ (two such transfers are shown in Fig. 3).

![Fig. 1.](image)

Fig. 1. (i) A rooted binary X-tree $T$ with a sequence $\sigma$ of six transfer events, labelled in increasing order into the past; (ii) the simplicial complex obtained from $T$ by deleting the interval immediately above the endpoint of each transfer; (iii) the resulting tree $T[\sigma]$. In this example, $\sigma$ induces a match for $a, b, c$ and a mismatch for $a, b, d$. 

![Fig. 2.](image)

Fig. 2. The tree $T_3$ in the sequence $T_0, T_1, T_2, T_3, \ldots, T_6$ where $T_6 = T$ is the tree shown in Fig. 1(i).
3.2. Triplet combinatorics under LGT transfers

Now, suppose that \( \sigma = \sigma_1, \ldots, \sigma_k \) is a sequence of transfers on \( T \). Consider the resulting sequence \( \{T_r; 0 \leq r \leq k\} \) of derived trees and let \( \sigma_r = (p, p') \).

- If \( \text{des}_A(T_{r-1}, p') = \{x\} \) for some \( x \in A \), and \( |\text{des}_A(T_{r-1}, p')| \leq 1 \), we say that \( \sigma_r \) is an \( A \)-transfer, and that it transfers \( x \).
- If \( \sigma_r \) transfers \( x \) and \( \text{des}_A(T_{r-1}, p') = \emptyset \), we say that \( \sigma_r \) moves \( x \) and we refer to \( \sigma_r \) as an \( A \)-moving transfer.
- If \( \sigma_r \) transfers \( x \), and if \( \text{des}_A(T_{r-1}, p') = \{y\} \), we say that \( \sigma_r \) joins \( x \) to \( y \) and we refer to \( \sigma_r \) as an \( A \)-joining transfer.

Note that any \( A \)-transfer is either a moving or joining transfer (but not both). Examples of \( A \)-transfers are shown in Fig. 3(i).

We begin by recalling the model of LGT described in Linz et al. (2007), which made the following assumptions: (1) a binary, labeled, rooted and clocklike species tree \( T \) is given, as well as all the splitting times along this tree; (2) differences between a gene tree and \( T \) are only caused by LGT events; (3) the transfer rate is homogeneous per gene and unit time; (4) genes are transferred independently; (5) one copy of the transferred gene still remains in the donor genome; (6) the transferred gene replaces any existing orthologous counterpart in the acceptor genome. We refer to this model from Linz et al. (2007) as the standard LGT model and we will consider extensions of it which relax assumptions (2) and (3). In particular, consider the following relaxation of assumption (3) in which transfer events on \( T \) occur as a Poisson process through time, in which the rate of transfer event from point \( p \) on a lineage to a contemporaneous point \( p' \) on another lineage at time \( t \) occurs at the rate \( f(d(p, p'), t) \) where \( f(d, t) \) is a constant or at least monotone non-increasing function in \( d \) (though it can vary non-monotonically in \( t \)) and \( d(p, p') \) is the evolutionary distance in the tree between contemporaneous points \( p \) and \( p' \) in the tree. We call this model the extended LGT model.

Both the standard and extended LGT models induce a well-defined probability distribution on gene tree topologies, both for the original set of taxa, and for any subset. In the standard LGT model, the number of transfers has a Poisson distribution, with a mean equal to the rate of LGT transfer out of any given point in the tree times the sum of the branch lengths (phylogenetic diversity) of the tree. We first establish that the Poisson distribution for the number of LGT transfers (in total or just the transfers in an \( A \)-lineage) still holds for the extended LGT model.

4. Statistical signals for the central tree under LGT and incomplete lineage sorting

Finally transfer the time dating from \( T_1 \) across to \( T_{1+1} \).

An example of the process \( T \rightarrow T_1 \) is illustrated in Fig. 3(ii) for \( \sigma = \sigma_1 \).

Lemma 2. Suppose \( \sigma = \sigma_1, \ldots, \sigma_k \) is a sequence of transfer events on a rooted binary \( X \)-tree \( T \) with \( t(\sigma_1) < t_0 \), and with no \( A \)-joining transfers. Then \( T[\sigma]\{A = T_1\} \).

(i) Deleting all \( p \in T_1 \) with \( t(p) < t(\sigma_1) \).
(ii) Labeling \( p_i \) by \( x \).
(iii) For each \( z \in A - \{x\} \), assigning label \( z \) to the unique point \( p_i \) of \( T_1 \) that has \( t(p_i) = t(\sigma_1) \) and \( z \in \text{des}(T_1, p_i) \).
(iv) We will regard the other leaves in the tree as unlabeled.

Finally transfer the time dating from \( T_1 \) across to \( T_{1+1} \).

A second combinatorial lemma, extends case (b) of Lemma 1 slightly and is used in the proof of Theorem 5. In order to state it, we need some further definitions.

Suppose \( \sigma = \sigma_1, \ldots, \sigma_k \) is a sequence of transfer events on a rooted binary \( X \)-tree \( T \) with \( t(\sigma_1) < t_0 \), and with no \( A \)-joining transfers. Construct an associated sequence of trees \( T_1, T_2, \ldots, T_k \) as follows. Set \( T_0 = T \) and construct \( T_{i+1} \) from \( T_i \) by the following procedure. If \( \sigma_1 \) is not \( A \)-moving then set \( T_{i+1} = T_1 \). If \( \sigma_1 = (p, p') \) moves \( x \in A = \{a, b, c\} \) then let \( T_{i+1} \) be the tree obtained from \( T_i \) by:

(i) Deleting all \( p \in T_i \) with \( t(p) < t(\sigma_1) \).
(ii) Labeling \( p_i \) by \( x \).
(iii) For each \( z \in A - \{x\} \), assigning label \( z \) to the unique point \( p_i \) of \( T_i \) that has \( t(p_i) = t(\sigma_1) \) and \( z \in \text{des}(T_i, p_i) \).
(iv) We will regard the other leaves in the tree as unlabeled.

Finally transfer the time dating from \( T_i \) across to \( T_{i+1} \).

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4. Statistical signals for the central tree under LGT and incomplete lineage sorting

We begin by recalling the model of LGT described in Linz et al. (2007), which made the following assumptions: (1) a binary, labeled, rooted and clocklike species tree \( T \) is given, as well as all the splitting times along this tree: (2) differences between a gene tree and \( T \) are only caused by LGT events; (3) the transfer rate is homogeneous per gene and unit time; (4) genes are transferred independently; (5) one copy of the transferred gene still remains in the donor genome; (6) the transferred gene replaces any existing orthologous counterpart in the acceptor genome. We refer to this model from Linz et al. (2007) as the standard LGT model and we will consider extensions of it which relax assumptions (2) and (3). In particular, consider the following relaxation of assumption (3) in which transfer events on \( T \) occur as a Poisson process through time, in which the rate of transfer event from point \( p \) on a lineage to a contemporaneous point \( p' \) on another lineage at time \( t \) occurs at the rate \( f(d(p, p'), t) \) where \( f(d, t) \) is a constant or at least monotone non-increasing function in \( d \) (though it can vary non-monotonically in \( t \)) and \( d(p, p') \) is the evolutionary distance in the tree between contemporaneous points \( p \) and \( p' \) in the tree. We call this model the extended LGT model.

Both the standard and extended LGT models induce a well-defined probability distribution on gene tree topologies, both for the original set of taxa, and for any subset. In the standard LGT model, the number of transfers has a Poisson distribution, with a mean equal to the rate of LGT transfer out of any given point in the tree times the sum of the branch lengths (phylogenetic diversity) of the tree. We first establish that the Poisson distribution for the number of LGT transfers (in total or just the transfers in an \( A \)-lineage) still holds for the extended LGT model.

Lemma 3. Under the extended LGT model, the total number of transfers, and the number of transfers into an \( A \)-lineage (for any given subset \( A \) of \( X \) of size 3) each has a Poisson distribution.

Proof. We begin by recalling a general property for any collection \( (P_1, \ldots, P_k) \) of independent Poisson processes, where process \( P_i \) has intensity \( r(i) \). Let \( Y_i(t_0) \) count the number of times process \( P_i \) occurs up to time \( t_0 \) and let \( Y = \sum_{i=1}^{k} Y_i(t_0) \). Then \( Y_i(t_0) \) has a Poisson distribution with mean \( \int_{0}^{t_0} r(i) \) dt. Moreover, since the sum of independent Poisson random variables has a Poisson
distribution, with a mean equal to the sum of the individual means, it follows that \( Y \) has a Poisson distribution, with mean \( \sum_{i=1}^{k} E(t_i) \). (For background on these stochastic results, the reader may wish to consult Grimmett and Stirzaker, 2001.) We apply this result as follows: let \( N \) be the total number of transfers and \( N_k \) the number of transfers into an \( A \)-lineage in the tree. We can express \( N \) by considering all intervals \( I \) between speciation events in \( T \) and, within each interval, consider all ordered pairs of lineages (restricted to that interval) \( I_1, I_2 \). For each such ordered triplet \( i = (I_1, I_2) \), let \( P_i \) be the Poisson process of transfers from \( I_1 \) to \( I_2 \) (which may depend non-homogeneously on time) and \( N_i \) denotes the number of these transfers for this triplet. Then the \( P_i \) are independent processes, and \( N = \sum_{i=1}^{k} (I_1, I_2) N_i \), so \( N \) has a Poisson distribution.

Regarding \( N_A \), we consider all intervals \( I \) between speciation events in \( T \) and, within each interval, consider all ordered pairs of lineages (restricted to that interval) \( I_1, I_2 \), where \( I_2 \) (respectively \( I_1 \)) has exactly one (respectively at most one) of \( a, b, c \) as a descendant. Let \( P_i \) be the Poisson process of transfers from \( I_1 \) to \( I_2 \) (which may depend non-homogeneously on time) and let \( N_i \) denote the number of such transfers for this triplet. Note that the \( P_i \) are independent processes, and that \( N_A = \sum_{i=1}^{k} (I_1, I_2) N_i \), and so this quantity again has a Poisson distribution.

Although the process of LGT transfers (under the standard or extended) model is a continuous time process, there is an associated discrete process that induces an identical distribution on gene trees; it is obtained by considering the decomposition described in the proof of Lemma 3. Thus with any sequence \( \sigma = \sigma_1, \ldots, \sigma_k \) of LGT transfers on \( T \), we may associate the discrete sequence \( s = s_1, \ldots, s_k \) where \( s_i \) refers to a triple \((I_1, I_2)\), in which case \( T(\sigma) = T(s) \) (i.e., all that matters in determining the resulting tree topology is the sequence of transfers between branches of the tree and their relative ordering, not the actual times that they occur). Consequently, each such discrete sequence \( s = s_1, \ldots, s_k \) for \( T \) has a positive probability, and under the standard LGT model, this probability has a Poisson distribution that just depends on \( k \).

5. Exact analysis of triplets in three- and four-taxon trees

For a transfer sequence \( \sigma \) generated by this type of LGT process, and any triplet \( a, b, c \in X \), we are interested in the probability that \( \sigma \) induces a match for \( a, b, c \). For three leaves, an exact analysis is straightforward, as we now show.

**Proposition 4.** If \( T \) has just three taxa, then under the extended LGT model, the probability that a transfer sequence induces a match for the three taxa is strictly greater than the probability it induces either one of the two mismatch topologies (which have equal probability).

**Proof.** When \( A = \{a, b, c\} \) then there are no \( A \)-moving transfers, and a transfer \( \sigma \) is \( A \)-Joining if and only if \( t(\sigma) < t_A \); let \( N \) denote the number of such transfers. We can express \( N \) as the sum of six random variables, which counts the number of transfers from the lineage \( x \) to lineage \( y \) (for \( x, y \in A, x \neq y \)). By the assumptions of the model, and Lemma 3, \( N \) has a Poisson distribution with some fixed mean \( m \). Now if \( N = 0 \) we obtain a matching topology for the three taxa, while if \( N > 0 \) then, by Lemma 1, the topology of the tree is determined by the first transfer (since it is, by necessity, \( A \)-Joining) and under the assumption of the model (in particular invoking the property that \( f \) is non-increasing), the probability that this first transfer is between two of the most closely related taxa is at least \( \frac{1}{2} \). Thus, the probability of a matching topology for the three taxa is at least
\[
P(N = 0) + \frac{1}{4} P(N > 0) = e^{-m} + \frac{1}{4}(1 - e^{-m}),
\]
while for either mismatch topology, the probability is equal to the other mismatch topology and is no more than \( \frac{1}{4}(1 - e^{-m}) \). This completes the proof.

5.1. Four-taxon case

The analysis of the distribution of triplet gene tree topologies generated by random LGT transfers on a four-taxon tree is considerably more interesting than the three-taxon case. Note that with four taxa, there are two rooted binary tree shapes—the ‘fork-shaped’ tree (with two cherries, as shown in Fig. 4) and the pectinate tree (with one cherry). We will provide an exact analysis for the first of these tree shapes under the standard LGT model, as this suffices to demonstrate a zone of statistical inconsistency, though we discuss briefly how an analogous (but more complex) analysis could be carried out for the other rooted tree shape. Our analysis requires no approximations, nor any imposition of an upper bound on the number or rate of LGT transfers. It relies on associating a random walk on a six-cycle to the LGT process.

We begin with some definitions. For four leaves \( x, y, z, w \), we write \((x,y; w,z)\) to denote the rooted binary tree, with the leaves \( x, y \) on one side of the root and \( w, z \) on the other, and with the MRCA of \( x, y \) at a fixed time \( t_{x,y} \) and the MRCA \( w, z \) at a later fixed time \( t_{w,z} \) (Thus \((x,y; w,z) = (y,x; w,z) = (x,z; y,w) = (x,y; z,w) \)), but otherwise the same symmetries hold. For example, the tree \((a, \emptyset ; b, \emptyset ; c, \emptyset ; d, \emptyset )\) is shown in Fig. 4(i) and the tree \((b, c; a, \emptyset ; \emptyset ; \emptyset )\) in Fig. 4(ii). Here \( \emptyset \) refers to the fourth taxon, the identity of which plays no role when we come to consider the topology of the triplet \( a, b, c \).

We now state the main result of this section.

**Theorem 5.** Suppose \( T \) is a rooted four-taxon tree and \( A = \{a, b, c\} \) is a subset of the leaf set \( X \) of \( T \), and suppose that \( T(A) = a \) \( b,c \) (thus \( T \) is a tree of type \( \tau^p \) in \( T \) is an Jenningsia). Let \( P \) be the probabilistic mapping from the transformation of \( T(A) = a \) \( b,c \) under the standard LGT model of Linz et al. (2007), \( (x,y,z) \to A \). Let \( \lambda \) denote the instantaneous rate of transfer events out of any given lineage.

(i) Suppose \( T \) is of type \( \tau^p \). Then for any \( t \)-value for the MRCA of \( \{a, \emptyset \} \), there is a sufficiently large \( t \)-value for the MRCA of \( \{b, c\} \), for which the matching gene tree topology \( a:b:c \) has a lower probability than either of the alternative mismatch topologies. More precisely, for \( \mu = \frac{1}{2} \lambda t_{a} \) \( \lambda t_{b,c} \), we have
\[
P(a:b:c) < \frac{1}{4}[1 - e^{-\mu}(1 - e^{-\mu_1})(1 + e^{-\mu_2})],
\]
where \( \mu_1 = \lambda t_{a} \) and \( \mu_2 = \lambda t_{b,c} \).

(ii) The species tree topology \( \tau^p = \{b, c; a, \emptyset \} \) which, in contrast to (i), always has higher probability under the standard LGT model for a matching gene tree topology \( a:b:c \) than for either mismatch topology.
and $P(b|ac) = P(c|ab)$; moreover, $P(a|bc) < P(b|ac) = P(c|ab)$ if and only if $t_{b,c} - t_{a,c}$ is greater than $(1/3)\ln((1+e^{-2\mu})/(1-e^{-2\mu}))$.

(ii) Suppose $T$ is of type $\tau_a$. Then for any $t$-values for the MRCAs of $(a,*)$ and $(b, c)$, the matching gene tree topology $a|bc$ has a higher probability than either of the alternative mismatch topologies. More precisely, for $\tau_a$, we have

\[
P(a|bc) = \frac{1}{3}(t_{b,c} - t_{a,c}).
\]

In this case, $P(a|bc) > P(b|ac)(= P(c|ab))$ for all values of $t_{a,c}$ and $t_{b,c} - t_{a,c}$.

**Proof.** First observe that, by symmetry, we have

\[
P(b|ac) = P(c|ab),
\]

under the standard LGT model (indeed this holds here even under the extended LGT model).

Consider the cyclic graph whose nodes are the six trees

$\tau_a = (a,*,b,c), \quad \tau'_a = (b,c;a,*), \quad \tau_b = (b,*,a,c), \quad \tau'_b = (a,c;b,*), \\
\tau_c = (c,*,a,b), \quad \tau'_c = (a,b,c,*),$

which are connected into a cycle as shown in Fig. 5.

Now, let $Z_t: t \geq 0$ be a continuous-time symmetric random walk on this 6-cycle graph, where the instantaneous rate of moving from one node to either given neighboring node is 1. Let $p_i(t) r = 0,1,2,3$ be the probability that, after running the process for time $t$, this Markov process is at a node that is graph distance $r$ (by the shortest path) from its initial state.

Note that $p = p(t) = [p_0(t), p_1(t), p_2(t), p_3(t)]^T$ satisfies the system of first-order linear differential equations

\[
\frac{dp}{dt} = Bp,
\]

where $B$ is the $4 \times 4$ tridiagonal matrix

\[
\begin{bmatrix}
-2 & 1 & 0 & 0 \\
2 & -2 & 1 & 0 \\
0 & 1 & -2 & 2 \\
0 & 0 & 1 & -2 \\
\end{bmatrix}
\]

and so

\[
p(t) = \exp(\beta t)p(0) \quad \text{where} \quad p(0) = [1,0,0,0]^T.
\]

The eigenvalues of $B$ are 0, $-1$, $-3$, $-4$, and the random walk on the 6-cycle is a reversible Markov process with uniform

---

**Fig. 5.** Conditional on there being no $A$-joining transfers between the present and $t_a$, the topology of the tree sequence $T'$ induced by $A$-moving transfers in this interval is equivalent to a simple random walk on the six-cycle graph shown.
equilibrium frequency, and so
\[
\lim_{t \to \infty} \mathbf{p}(t) = \{1/6, 1/3, 1/3, 1/6\},
\]
and each component \(p_j(t)\) is of the form
\[
p_j(t) = a_j + b_j e^{-t} + c_j e^{-3t} + d_j e^{-4t},
\]
for constants \(a_j, \ldots, d_j\) that are determined by the eigenvectors of \(B\).

Using standard matrix diagonalization techniques from linear algebra, we obtain the following solution to Eq. (4)
\[
\mathbf{p}(t) = \begin{bmatrix}
\frac{1}{3} & 1 & 1 & \frac{1}{3} \\
1 & 1 & -1 & -1 \\
-1 & 1 & 1 & 1 \\
\frac{1}{2} & -1 & 1 & -\frac{1}{2}
\end{bmatrix}
\begin{bmatrix}
e^{-t} \\
e^{-3t} \\
e^{-4t}
\end{bmatrix}.
\]

From this, one immediately obtains the following result, which will be required later.

**Lemma 6.** For all \(t > 0\)
\[
p_1(t) - 2p_3(t) = e^{-t} - e^{-3t} > 0,
\]
and
\[
2p_0(t) - p_2(t) = e^{-t} + e^{-3t} > 0.
\]

We return now to the proof of Theorem 5. We will establish part (i), and indicate how the proof of part (ii) follows by a directly analogous argument.

Let \(E\) denote the event that the random sequence of transfer events \(\sigma\) generated by the model induces a match for \(A\). Let \(J\) denote the number of \(A\)-joining transfers between \(t=0\) and \(t=t_{a,b,c}\). Then \(J\) has a Poisson distribution with mean \(2 \lambda t_{a,b,c} = 6 \mu\), since at any moment in the interval \([0, t_{a,b,c}]\), there are four lineages, three of which lead to leaves in \(A\) (therefore, for any \(x \in A\), the rate of transfer from that \(x\)-lineage to any other lineage that also lead to \(A\) is \(\lambda (2/3)\)). Thus the cumulative rate of an \(A\)-joining transfer is \(3\lambda (2/3) = 2\lambda\). Consequently
\[
P(\lambda > 0) = 1 - e^{-6\mu}.
\]
\[
\text{Now, Lemma 1 (part (b)(ii)) implies that}
\]
\[
P(\epsilon|J > 0) = \frac{1}{b},
\]
and by the law of total probability, Eqs. (5) and (6) give
\[
P(\epsilon) = \frac{1}{4} (1 - e^{-6\mu}) + e^{-6\mu} P(\epsilon|J = 0).
\]

Now the \(A\)-moving transfers between \(t=0\) and \(t=t_{a,b,c}\) constitute a continuous-time Poisson process for which the rate at which any given \(x \in A\) is moved is \(\frac{1}{2}\lambda\). Note that this process is independent of \(J\) since the source point of an \(A\)-moving transfer is always on a different type of lineage (having an element of \(A\) as a descendant) from an \(A\)-moving transfer. As the \(A\)-moving process proceeds in time (from \(t=0\) to \(t=t_{a,b,c}\)), the resulting sequence of trees \(T\) described in the preamble to Lemma 2 corresponds to a simple symmetric random walk on the nodes of the 6-cycle shown in Fig. 5, starting with tree \(T_0\) at time \(t=0\), and where the rate of moving from one node to any particular neighboring node is \(\frac{1}{2}\lambda\). At time \(t=s\), the length of two pendant edges of the \(t\)-tree will be \(t_{a,b,c} -s\) while the other two pendant edges have a larger length of \(t_{a,b,c} -5\), so we stop the process when the length of the shorter pair of pendant edges reaches zero (i.e. at \(t=t_{a,b,c}\)). Note that the length of the pendant edges does not affect the transition process under the standard LGG model, so we can indeed view it as a discrete-state random walk on six states (rather than on a continuum of states).

**Lemma 2** now ensures that if \(\epsilon^*\) is the sequence of \(A\)-moving transfers between \(t=0\) and \(t=t_{a,b,c}\) then \(T(\epsilon^*)\) resolves \(a, b, c\) in the same way as the tree \(T_{b,c}\) does, where \(T_{b,c}\) is the state of the random walk on the 6-cycle at time \(t_{a,b,c}\). At time \(t=t_{a,b,c}\) the random walk on the 6-cycle is at one of the following nodes:

- \(\epsilon^*_a\), in which case \(T(\sigma_0|A = a|bc)\) (with probability 1); this does not depend on any transfer events that may occur after \(t_{a,b,c}\);
- \(\epsilon_{a,b}\), in which case \(T(\sigma_0|A = a|bc)
\]
- with probability 1 if there is no transfer event between \(t_{a,b,c}\) and \(t_{b,c}\), or
- with probability \(\frac{1}{2}\) if there is at least one transfer event between \(t_{a,b,c}\) and \(t_{b,c}\);
- \(\epsilon^*_c\), in which case \(T(\sigma_0|A = a|bc)\) regardless of any further transfers.

In this case analysis, the probability factor \(\frac{1}{2}\) arises from Lemma 1, part (b)(ii). Note also that the probability that there is at least one transfer event between \(t_{a,b,c}\) and \(t_{b,c}\) has probability \(1 - e^{-\lambda}\), and this event is independent of the random walk on the 6-cycle.

Consequently, by independence, and by combining these cases we obtain
\[
P(\epsilon|J = 0) = p_3(\mu) \cdot \left(1 + p_0(\mu) \cdot (e^{-\lambda} + \frac{1}{2}(1 - e^{-\lambda})) + p_2(\mu) \cdot \frac{1}{2}(1 - e^{-\lambda}) + p_1(\mu) \cdot 0ight),
\]
which, combined with Eq. (7) and Lemma 6, establishes Eq. (1). The remainder of part (i) is justified by Eq. (3) and straightforward algebra to determine when the expression in Eq. (1) is lower than the probability of a specific mismatch topology on \(A\).

For the proof of part (ii), an analogous argument shows that for \(T(a,b,c)\) of type \(\epsilon^*_a\) and \(\mu = \frac{1}{2}\lambda t_{b,c}\), and \(B = 3\lambda t_{a,b,c} - t_{b,c}\), Eq. (7) still holds, that is
\[
P(\epsilon|J = 0) = \frac{1}{4} (1 - e^{-6\mu}) + e^{-6\mu} P(\epsilon|J = 0).
\]
For the last term, we find that
\[
P(\epsilon|J = 0) = p_0(\mu) \cdot (p_1(\mu) + p_2(\mu) \cdot \frac{1 - e^{-\lambda}}{3} + p_3(\mu) \cdot e^{-\lambda} + p_2(\mu) \cdot 0),
\]
from which Eq. (2) now follows, by Lemma 6. The remainder of part (ii) is justified by Eq. (3), and straightforward algebra shows that the expression in Eq. (2) is never lower than the probability of a specific mismatch topology on \(A\).

5.2. Statistical inconsistency?

Part (i) of Theorem 5 shows that the species tree topology for three taxa inside a larger tree can have the lowest probability among the three possible gene tree topologies on those three taxa, under the standard LGG model (see Fig. 6). This is in sharp contrast to what occurs with incomplete lineage sorting, where the most probable gene tree for three taxa matches the species tree topology for those three taxa, regardless of what other taxa are present, and how they are arranged in the species tree. Thus, in the setting of Theorem 5(i), estimating the species tree for a set
A of three taxa from the frequency of triplet gene trees will be statistically inconsistent (it will converge on an incorrect tree). However, this does not imply that one cannot estimate the species tree from the probability distribution of all gene trees topologies (and for all subsets A of size three from X). Moreover, if we use the R* tree reconstruction method (which infers clades that unanimously support majority-rule triplet topologies (Bryant, 2003) for four-taxon tree having the clades \((a, b), (c, d)\), Theorem 5 (parts (i) and (ii)) shows that we will always return a tree that includes at least one of these clades, and no contradictory clades. Thus, the R* method would, in this case, be only weakly inconsistent (i.e. it would return a tree that is either equal to or is resolved by the species tree, rather than being positively misleading).

5.3. The other tree shape on four taxa

One could perform a similar analysis for the 12 rooted binary trees that have the four leaves \((a, b, c, d)\) and just a single cherry (rather than two cherries as above). In this case, the associated transition graph consists of a 12-cycle, together with three additional edges—obtained by placing an edge between \((xy)z)w\) and \((xy)w)z\) for each of the three choices of \((x, y)\) from \((a, b, c)\). This graph is shown in Fig. 7. The analysis of the probability of a matching topology for \(a, b, c\) under the random LGT model (depending on the position of the * lineage) could be carried out by a similar, albeit more complex, analysis to that for the simpler 6-cycle graph, but this is beyond the scope of the current paper.

6. General case, trees with \(n\)-taxa

We now include the three- and four-taxon results into a more comprehensive statement concerning the statistical consistency of species tree reconstruction under the LGT model, for an arbitrary number of taxa.

**Theorem 7.** Consider the standard or extended LGT model on a rooted binary phylogenetic \(X\)-tree.

(i) If \(T\) has just three taxa, then under either model the probability that a transfer sequence induces a match for the three taxa is strictly greater than the probability that it induces either one of the two mismatch topologies (which have equal probability).

(ii) A four-taxon tree and branch lengths exist for which the model can assign higher probability to a particular mismatch topology for some triplet, \((a, b, c)\), than for a match, even under the standard LGT model of Linz et al. (2007).

(iii) Regardless of the number of taxa in the tree and the branch lengths if \(N_A\), for some subset \(A\) of taxa of size 3, the expected total number of transfers into an \(A\)-lineage (for the particular gene) is no more than 0.69 in the extended model, and no more than 1.14 in the standard LGT model of Linz et al. (2007), then the probability of a topology match is strictly greater than the probability of either of the mismatch topologies.

(iv) When LGT rates ensure that condition (iii) holds for every subset of \(A\) of taxa of size 3, there is a polynomial-time method for reconstructing the species tree from the gene trees which is statistically consistent under the model, as the number of independently generated gene trees tends to infinity.

**Proof.** Parts (i) and (ii) are established by Proposition 4 and Theorem 5 respectively.

**Proof of part (iii).** Let \(N_A\) denote the total number of transfer events of the gene in the tree into an \(A\)-lineage. By Lemma 3, \(N_A\) has a Poisson distribution with some mean \(m\). Then for any triple

![Fig. 6. Plot of Eq. (1) in the statement of Theorem 5. Note that the probability \(P(a|bc)\) for a match (shown in the z-axis) is less than \(\frac{1}{3}\) in the bottom right-hand side of the figure.](image-url)
Suppose that $T_{a,b,c} = a/bc$. Then if $N_A = 0$ then there is a match with probability 1. Thus, in the extended model, if $m < \ln(2) \approx 0.69$ then the probability of a match is at least $P(N_A = 0) = e^{-m} > 0.5$. This establishes the first claim in part (iii).

For the second claim in part (iii), consider the standard LGT model. In Appendix we establish the following claim by means of a coupling-style argument.

**Claim 1.** Consider a sequence of transfer events under the standard LGT model. Then, conditional on the event that $N_A = 1$, the probability $p$ that this sequence of transfers induces a match for $A$ is greater or equal to the probability $q$ of inducing a specific mismatch topology (say $c|ab$) for $A$.

Now, under the standard LGT model, the probability of a match is at least

$$P(N_A = 0) + p \cdot P(N_A = 1),$$

while the probability of a specific mismatch topology (say $c|ab$) is at most

$$q \cdot P(N_A = 1) + P(N_A > 1),$$

and from Claim 1, $p \geq q$ so the difference obtained by subtracting the mismatch topology probability (10) from the matching topology probability (9) is at least

$$P(N_A = 0) - P(N_A > 1) = e^{-m} - (1 - e^{-m} - me^{-m}) = e^{-m}(2 + m) - 1,$$

and the term on the right is strictly positive for $m \leq 1.14$, as claimed.

**Proof of part (iv).** We can apply the same argument used by Degnan et al. (2009), who showed that the (polynomial time) $R^*$ tree reconstruction method (based on triplet topologies) is statistically consistent under models of incomplete lineage sorting. Here, we are dealing with LGT rather than incomplete lineage sorting, but the only property required of either model in order to ensure the statistical consistency of the $R^*$ method is that for each triplet $A$, the probability that the gene tree matches the species tree topology restricted to $A$ has a probability that is greater (by some fixed $\epsilon > 0$) than either of the other two topologies (in the case of incomplete lineage sorting, the two alternative topologies have equal probability, but this may not be the case under LGT; however, this is not essential to prove consistency).

### 6.1. Rates of LGT

Theorem 7 (part (iii)) requires a small expected number of transfers into an $A$-lineage for any subset $A$. The question arises as...
to how this expected number would compare with the total expected number of transfers in the tree. The total number of LGT transfers in the tree \( N_{\text{tot}} \) has a Poisson distribution; under the standard LGT model this distribution has mean \( r \cdot L(T) \), where \( r \) is the rate of LGT transfer out of any given lineage at any given time and \( L(T) \) is the phylogenetic diversity of \( T \) (the sum of the lengths of all its branches). For a Yule (pure-birth) tree with \( n \) taxa and timespan \( t \), if the speciation rate is set equal to its expected value, then from Steel and Mooers (2010), \( L(T) \) has expected value

\[
L_n(t) = \frac{(n-2)t}{\ln(n/2)}. \tag{11}
\]

On the other hand, the rate of transfers into an \( A \)-lineage at any time is at most \( 3r \) and so the expected number of transfers into an \( A \)-lineage is, at most, \( 3rt \). Thus, if the expected number of LGT transfers in the entire tree is \( G \) then the expected number of transfers into an \( A \)-lineage in a tree with phylogenetic diversity equal to its expected value under the Yule model is, at most

\[
3\ln(n/2)/(n-2) G. \tag{12}
\]

For example, if \( n=200 \) and \( G \leq 10 \) (on average every gene is transferred at most 10 times on the tree) then (12) takes the value 0.7, which is within the 1.14 bound of Theorem 7 (part (iii)). We note that in one study, it was suggested that the average number of times each gene has been transferred might be around 1.1 (Dagan and Martin, 2007), so the condition imposed in Theorem 7 (iii) may not be unreasonable. In the recent paper by Abby et al. (2012), an average rate across bacterial genomes was estimated at between 0.02 and 0.04 LGTs per branch of the tree. Thus, for \( n=200 \), \( G \) is about 8–16 events, where the upper end of these results include some of extremely high rates of LGT in bacteria.

The discussion raises the question of the possible impact on tree reconstruction accuracy of LGT transfers from other lineages, additional of those connecting the 200 taxa, assuming the standard LGT model applies across this larger tree. Notice that if we consider the enlarged set of taxa of size \( n' > n \) then all triplets will be consistently inferred for the same value of \( G \) (or even one that grows sufficiently slowly as to satisfy inequality (12) with \( n \) replaced by \( n' \)). It follows that the original subtree of \( n=200 \) will still be consistently inferred. This is reassuring since Theorem 5(i) established a case of statistically inconsistent tree reconstruction on triplets within a larger (four-taxon) tree, even though Proposition 4 had established statistical consistency for triplets (without other lineages contributing LGT transfers) under the standard LGT model. However, this earlier inconsistency result requires the LGT rate \( 3r \) to be sufficiently large in relation to certain branch lengths, and here we place an upper bound on the LGT rate.

6.2. LGT and incomplete lineage sorting

We have described sufficient conditions for the \( R^* \) tree-reconstruction method to be a statistically consistent estimator of a species tree topology under various LGT models. Moreover, it is known that the \( R^* \) method is also a statistically consistent estimator of the species tree topology under lineage sorting and without any non-trivial restrictions on branch lengths (Degnan et al., 2009). It follows that if each topology of each gene tree is determined by either LGT acting on the species tree topology or by incomplete lineage sorting (but not by both processes) then the \( R^* \) tree reconstruction method can be a statistically consistent estimator of species tree topology (under conditions where it would be for the genes undergoing LGT).

One could also ask what happens when the two processes are combined—that is, if we allow the ancestry of a gene to follow transfers and to coalesce within lineages, under the usual coalescent process. While it is possible to extend the earlier results a little in this direction (results not shown), the applicability of the results is somewhat limited for the following reason.

LGT is especially prevalent in haploid, largely asexual taxa with limited recombination, such as bacteria and archaea and, in this case, although incomplete lineage sorting may apply in considering how genetic lineages coalesce in the species tree, there is an important difference to diploid sexual taxa. Namely, in this latter case, the coalescent history of each gene sampled from an extant individual in each species represents an essentially independent sample from the same (multi-species coalescent) process, while in taxa with little or no recombination, the lineages of all genes follow the same ancestral trajectory, apart from LGT events. This complicates any statistical analysis based on assuming that the genes are independent samples from a common process and leads to some delicate statistical issues in attempting to analyse data with such a mixed mode. We defer this issue for future consideration.

7. Missing taxa: primordial tree consensus

One obstacle to applying the \( R^* \) construction is that many genes may not be present across all taxa (Sanderson et al., 2010). This may be due to a variety of factors, including gene loss or gene conversion, or simply because certain genes have not been sequenced yet.

Consequently, we describe a slight extension of the \( R^* \) consensus approach to handle this situation (our approach is similar to the ‘Quartet Plurality’ reconstruction method described in Roch and Snir, 2013, but our approach always returns a tree on any input). For any three taxa \( a,b,c \in X \), let \( G(a,b,c) \) denote the set of genes that are present in all three taxa \( a,b,c \). We will assume that the pattern of taxon coverage is sufficiently dense that the following condition holds:

\[
G(a,b,c) > 0 \quad \text{for all } a,b,c \in X. \tag{13}
\]

To give some indication of how much coverage this requires, if \( n_i \) denotes the number of taxa containing gene \( i \in \{1, \ldots, k\} \), and \( n \) is the total number of taxa, then we require

\[
\sum_{i=1}^{k} n_i(n_i-1)(n_i-2) \geq n(n-1)(n-2)
\]

(from the proof of Theorem 1 of Sanderson et al., 2010), which, in turn, implies the weaker but simpler, inequality

\[
\sum_{i=1}^{k} f_i^3 \geq 1 \quad \text{for } f_i = n_i/n.
\]

We consider the following simple extension of the \( R^* \) consensus method to the setting of partial taxon coverage. For \( a,b,c \in X \), let \( c(a/b/c) \) denote the proportion of genes in \( G(a,b,c) \) which resolve \( a,b,c \) as the triplet topology \( a/b/c \).

Lemma 8. The set

\[
C = \{ b \in X : c(f(b)/aa') > \max\{c(f(a/a'b), c(f(a'/ab)) \}
\]

for all \( a,a' \in A, a \neq a' \) and \( b \in X - A \}

forms a hierarchy, and can be constructed in time that is polynomial in \( n \).

This procedure has been implemented in the phylogenetic software package Dendroscope 3 (version 3.2.2) (Huson and Scornavacca, 2012) as the ‘primordial tree’ consensus method.
Now, under a model in which the pattern of gene presence and absence is a random process (as in Sanderson et al., 2010, where the presence or absence of a gene for each taxon is an independent stochastic process) and provided this process is independent of the LGT process, the results on the statistical consistency of species tree reconstruction will carry over. The same also holds if we were to consider incomplete lineage sorting rather than LGT.

Fig. 8 shows a tree constructed in this way from the recent bacterial data set of Abby et al. (2012), for which the authors estimated that the rate of LGT was fairly high (but not too high to erase all phylogenetic signal). The input was 1338 rooted gene trees on variable label sets for the Actinobacteria phylum of their study (the clade at the upper left in Fig. 3 of Abby et al., 2012).

The ‘unrooted’ gene trees (from the website of the authors) were midpoint-rooted using the phylogenetic program ‘Phylip’. Abby et al. (2012) implemented a complex procedure as part of their ‘Prunier’ software, to look at all possible rootings of the input gene trees, selecting the ones that minimized the number of LGT events. Rooting them in a new way here provides an independent analysis of these data.

The output tree is fairly close to the tree suggested by Abby et al. (2012). The biggest difference is that the primordial tree roots it at *R. xylanophilus*, which is a very long branch in their Fig. 3. The authors of Abby et al. (2012) were concerned about long branch attraction in these data.

As a second application to a quite different data set (Eukaryotes) and a different process (incomplete lineage sorting rather than LGT), Fig. 9 shows a tree constructed by the same method, from 1000 rooted gene trees from chromosome 3 of 11 taxa of the genus *Oryza* (rice and its relatives). Some trees contained all 11 taxa but most did not; however the pattern of taxon coverage is sufficiently dense that condition (13) does hold. While this data set is unlikely to exhibit LGT at anywhere near the rate of the previous one, gene flow persisting for some time after speciation would essentially show the same pattern as LGT. Moreover, incomplete lineage sorting is likely quite extensive across the genome at several nodes in the tree (Zou et al., 2008; Zwickl et al., 2012).

We note that our method requires estimates of rooted gene trees. We used different rooting methods in the rice and bacterial data sets, which represent a range of methods available, each with strengths and weaknesses. For the rice data set, the species tree is well understood from many other studies (Zou et al., 2008) and there is little doubt that *Oryza brachyantha* is the outgroup to the other taxa. We therefore used this to root all gene trees. Although rooting by reference to a carefully chosen outgroup has long been advocated, the method was designed in the context of morphological data and species tree inference (Watrous and Wheeler, 1981), before extensive collections of gene trees became available—and their frequent discordance uncovered (Pollard et al., 2006). Moreover, inclusion of an outgroup can actually disrupt correct inference in the ingroup under conditions likely found in the bacterial data, a combination of long and short branches (Holland et al., 2003). In cases in which the wider species tree is no longer considered a reliable guide for selecting gene tree outgroups, alternative methods have been increasingly used, including those that rely on midpoint rooting (Farris, 1972), assuming a molecular clock (Huelsenbeck et al., 2002), or rooting with gene duplications (Brady et al., 2011). Despite its simplicity, midpoint rooting performs surprisingly well compared to outgroup rooting (Hess and De Moraes Russo, 2007). We therefore used midpoint rooting on the bacterial gene trees. This had one clear advantage: we could root all gene trees this way, including those with differing label sets (matching differing label sets to a fixed collection of outgroups is problematic), and thus bring to bear the largest quantity of data to compare with the published results.

7.1. Questions for future work

It would be interesting to extend the scope of Theorem 5 to include the other four-taxon tree-shape (under the standard LGT model), as well as to analyze both tree shapes under the extended LGT model. Our analysis also raises some intriguing statistical questions: Is strong statistical inconsistency possible (for R* or perhaps other methods)? Is the species tree identifiable from the probability distribution on gene trees, regardless of the LGT rate? If the rate of LGT decreases sufficiently fast with phylogenetic distance in the tree, then is statistical consistency restored for the R* method? And what can be said regarding statistical issues arising when we combine LGT and incomplete lineage sorting and phylogenetic sampling error? Further research on these questions would help us better understand the extent to which signal for a species tree can be recovered above the ‘noise’ of random processes that can cause gene trees to conflict with the species tree.

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Fig. 9. A tree constructed from 1000 rooted gene trees from chromosome 3 for taxa of the genus *Oryza* for which incomplete lineage sorting (rather than LGT) is a likely cause of gene tree discordance.
manuscript. We also thank Derrick Zwickl and the Oryza Genome Evolution Project.

**Appendix A. Proof of Claim 1 (from proof of Theorem 7)**

Consider a sequence $\sigma$ of one or more transfer events, which contains exactly one transfer $\sigma_i = (p, p')$ that is into an $A$-lineage and for which $T[\sigma_i] = A \rightarrow ab$. We will associate with $\sigma$ another sequence of transfer events $\sigma'$ which induces a match for $\overline{A}$, and which is identical to $\sigma$ except that $\sigma$ is substituted by a particular alternative transfer $\sigma''$.

In case $\sigma$ is an $A$-joining transfer (in which case it joins $a$ to $b$, or $b$ to $a$ in order for $T[\sigma] = A \rightarrow c(ab)$ we replace $\sigma$ with the transfer $\sigma'$ (at the same time-instant) that:

1. $\sigma$ joins $b$ to $c$ if $\sigma$ joins $a$ to $b$;
2. $\sigma$ joins $c$ to $b$ if $\sigma$ joins $b$ to $a$.

These two cases are illustrated in Fig. 10.

Otherwise, in case $\sigma = (p,p')$ is an $A$-moving transfer, either:

(iii) $\text{des}_A(T, p') = (a)$, or
(iv) $\text{des}_A(T, p') = (b)$, or
(v) $\text{des}_A(T, p') = (c)$.

Consider case (iii). Recalling that $\sigma$ is the $r$-th transfer in $\sigma$, in order for $\sigma$ to induce the topology $c|ab$, there must be a vertex $v$ in the derived trees $T_{r-1}$ which $\text{des}_A(T_{r-1}, v) = (a,b)$; moreover, since $\sigma$ is the only transfer into an $A$-lineage, $v$ must lie on the path in $T$ between $b$ and the MRCA of $b, c$. In this case, we take $\sigma'' = (p, p_b)$ where $p_b$ is the unique point in $T$ with $t(p_b) = t(p)$ and which has $\text{des}_A(T, p_b) = (b)$.

In case (iv), in order for $\sigma$ to induce the topology $c|ab$ there must be a vertex $v$ in the derived trees $T_{r-1}$ for which $\text{des}_A(T_{r-1}, v) = (a,b)$; moreover, since $\sigma$ is the only transfer into an $A$-lineage, $v$ must lie on the path in $T$ between the $a$ and the MRCA of $A$. In this case, we take $\sigma'' = (p, p_a)$ where $p_a$ is the unique point in $T$ with $t(p_a) = t(p)$ and which has $\text{des}_A(T, p_a) = (a)$.

Case (v) is similar to case (ii) except that we can take $v$ to be the MRCA of $\{a,b\}$, and (as in case (ii)) we take $\sigma'' = (p, p_a)$ where $p_a$ is the unique point in $T$ with $t(p_a) = t(p)$ and which has $\text{des}_A(T, p_a) = (a)$.

Cases (iii)–(v) are also shown in Fig. 10. In all five cases (i)–(v), replacing $\sigma$ by $\sigma''$ in the sequence $\sigma$ results in the modified sequence $\sigma''$ that induces a match for $\overline{A}$; moreover the association $\sigma''$ is one-to-one on the set of transfer sequences with $N_{\text{rate}} = 1$ and which induce $c|ab$ and $a|bc$ respectively. It then follows that $p \geq q$ under the standard LGT model, as claimed.

**References**
