

INTERACTING ASSOCIATIONS BETWEEN PLOIDY, BREEDING SYSTEM, AND LINEAGE  
DIVERSIFICATION

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## Abstract

If particular traits consistently affect rates of speciation and extinction, broad macroevolutionary patterns can be understood as consequences of selection at high levels of the biological hierarchy. Identifying traits associated with diversification rate differences is tricky, though, because of the many traits available to consider and the statistical challenge of testing for associations from phylogenetic data. Two traits that have been repeatedly suggested as drivers of differential diversification are whether a lineage is diploid or polyploid, and whether it is self-incompatible or self-compatible. We investigate the role of each of these traits, and their interaction, on speciation and extinction rates in Solanaceae. We find that the effect of ploidy can largely be explained by its correlation with breeding system, and that additional unknown factors work with breeding system to determine diversification rates. These results are largely robust to assumptions about whether diploidization occurs. Finally, we show that allowing for state-dependent diversification affects conclusions about the relative contribution of different evolutionary pathways to self-compatible polyploids.

## Introduction

Species accumulate within the tree of life at different rates. A possible explanation for this phenomenon is that species possess various traits or character states that may differentially affect rates of diversification. Dramatic increases in available phylogenetic and phenotypic data as well as methodological advances have greatly accelerated the search for such traits that influence diversification traits. Nevertheless, identifying focal traits associated with rates of speciation and extinction remains as a difficult statistical problem (see (Maddison and FitzJohn 2015; Rabosky and Goldberg 2015; Moore et al. 2016; FitzJohn et al. 2009; Goldberg and Igić 2012; Beaulieu and O'Meara 2016; Rabosky and Goldberg 2017)). In part, this is a difficult problem because speciation and extinction rarely depend on a single trait; the context in which traits occurs can lead to complex interactions resulting in a heterogeneous speciation and extinction rates (Beaulieu and O'Meara 2016; Caetano et al. 2018; Herrera-Alsina et al. 2018). Consequently, examining the association of only one trait when multiple complex interactions are changing diversification patterns can be misleading.

In flowering plants, both ploidy and breeding system are well-studied traits that have been linked to diversification but may also affect each other (Stebbins 1950). Whole genome duplication (WGD) is a remarkably common mutation in plants (Husband et al. 2013; Zenil-Ferguson et al. 2017), resulting in many plant species with a variety of ploidy levels. The prevalence of ploidy variation has been broadly considered as a salient feature of flowering plants for nearly a century (Stebbins 1938).

WGDs have the potential to affect many phenotypes and impact a variety of evolutionary (Ramsey and Schemske 2002) and ecological processes (Sessa 2019). Polyploids have long been thought to have lower rates of net diversification rate than diploids (Mayrose et al. 2011, 2015). Recent studies, however, find common and numerous paleo-polyploidizations, including some preceding the emergence of highly diverse plant clades (Soltis et al. 2014; Landis et al. 2018), suggesting that WGDS have played an important macroevolutionary role driving innovation and diversification in plants. Evidence of paleo-polyploidizations within the genomes of diploid plants also reveals the pervasiveness of diploidization, the return of a polyploid to a diploid state throughout the angiosperm phylogeny (Soltis et al. 2015; Dodsworth et al. 2015). Testing for the presence of diploidization in a diversification context is important because the presence of a positive net diversification rate for diploids can be the result of a polyploid lineage that has been diploidized. Under this scenario, polyploidy should no longer be considered to slow down diversification, but instead, polyploidy would be the initiator of the diversification process.

Breeding system shifts—changes in the collection of physiological and morphological traits that determine the likelihood that any two gametes unite—are remarkably common and affect the distribution and

45 amount of genetic variation in populations (Stebbins 1974; Barrett 2013). In particular, self-incompatibility systems cause a plant to reject its own pollen, and their loss, yielding self-compatibility, is one of the most replicated transitions in flowering plant evolution. Previous phylogenetic analyses have reported higher rates of diversification for self-incompatible than for self-compatible lineages (Goldberg et al. 2010; de Vos et al. 2014), even when allowing the possibility that hidden states could drive diversification patterns (Freyman  
50 and Höhna 2018). However, studying breeding system with other correlated traits and unobserved states driving diversification patterns simultaneously has not yet being explored.

Ploidy might also affect the propensity for self-fertilization (Stebbins 1950). The evidence for correlated shifts in mating system, from outcrossing to selfing, after polyploidization appears limited and sometimes contradictory (Barringer 2007; Barrett 2008; Husband et al. 2008). In other cases, however,  
55 polyploidy is not only suspected to correlate with breeding system but causes the transition from self-incompatibility (SI) to self-compatibility(SC) (Stout and Chandler 1942; Lewis 1947). Doubled number of alleles in pollen is thought to effect disruption of the genetic mechanisms in gametophytic SI systems, which prevent self-fertilization (Entani et al. 1999; Tsukamoto et al. 2005; Kubo et al. 2010). This creates a correlation between polyploidy and SC by precluding the existence of SI polyploids. In clades with these  
60 systems, it is thus natural to consider the simultaneous macroevolution of polyploidy and breeding system.

Given that changes in ploidy and breeding systems may be causally related and have profound affects on the fate of lineages, it seems particularly profitable to examine possible interactions in their macroevolutionary effects. This includes their joint influence on lineage diversification, and also potential patterns in the order of their transitions. For example, it is possible to ask whether losses of SI more  
65 commonly occurred tied to polyploidization, or without a ploidy shift. Or whether polyploids have arised more commonly from self-incompatible than from self-compatible diploids/ Robertson et al. (2011) found that the pathway from self-incompatible diploids to self-compatible polyploids is dominated by loss of self-incompatibility followed later by polyploidization over long timescales, but proceeds in one step via polyploidization of SI species over short timescales. Greatly improved phylogeny and methods that allow  
70 for diversification rate differences allow far more powerful approach to such problems.

Two methodological developments are critical for disentangling the complex interactions between two or more traits and their link to diversification. First, the model proposed by FitzJohn (2012) that extended the state-dependent diversification “\*SSE” modeling framework (Maddison et al. 2007), to allow estimating the effect of an arbitrary number of states (enabling multiple states, MuSSE, instead of binary  
75 states, BiSSE). This allows multiple states to represent trait combinations, particularly useful if one is interested in explicitly testing whether combinations of traits non-additively affect speciation or extinction.

Second, the model by Beaulieu and O'Meara (2016) HiSSE proposed a first solution to the problem of elevated false positive type I errors found often in the BiSSE framework (Goldberg and Igić 2012). The authors argued that it is unlikely that the variation in speciation and extinction rates is only due to a focal trait but by considering heterogeneity in the diversification process as the null hypothesis, it is possible to parse out the importance of the focal trait in diversification while other unobserved processes can happen. HiSSE models have effectively shown that some traits associated with diversification in BiSSE analyses are actually not associated with the diversification process, and instead a hidden state is linked to differences in the diversification rate. Such results raise the question of the identity of the hidden state, or whether it is an approximation of some unknowable heterogeneity, as well as whether interactions between the known and hidden trait were modeled appropriately. Third, one can simultaneously consider the effects of multiple known traits using MuSSE, especially in systems where multiple traits are suspected to influence diversification, and where interactions between those traits are well understood. However, MuSSE might also suffer from high type I errors if there are other processes shaping diversification other than the traits observed (Caetano et al. 2018). Therefore, a fourth option exists: considering both the presence of multiple known traits linked to the diversification process, but also, some unknown or hidden state changing speciation and extinction rates. These four modeling strategies are proposed here in the context of studying complex interactions among traits, linked diversification.

The focus of our study is the individual role of ploidy and breeding systems, as well as their interaction, on the diversification process in Solanaceae. Using ploidy and breeding system data from species in Solanaceae, and considering possible heterogeneity in the diversification process, we assess the influence of these traits on speciation and extinction rates. First, we present models in which the effect of ploidy and breeding system are considered separately, and compare the inferences to previously published results. Second, we consider the addition of a hidden trait to each of the models that examine traits individually, in order to investigate whether the focal trait directly affects differences in diversification, or if such differences are attributable to unobserved but correlated hidden states. Third, we ask whether the complex interaction between breeding system and ploidy may be explained by the hidden trait states. We evaluate a multi-state model to consider both traits linked to diversification, and then consider whether an additional unobserved trait may be driving differences in diversification. In all these models we also investigate the potential effects of diploidization to explore whether the reversion of polyploidy changes diversification interpretations. Our results highlight the importance of considering non-additive effects of traits on net diversification rates, especially when there are strong biologically driven correlations amongst them.

## Methods

### *Data*

Chromosome number data were obtained for all Solanaceae taxa in the Chromosome Counts Database (CCDB; Rice et al. 2015), and the ca. 14,000 records were curated semi-automatically using the CCD-Bcurator R package (Rivero et al. 2019). CCDB contains records from original sources that have multiple complex symbol patterns denoting multivalence, or irregularities of chromosome counts. After a first full round of curation via CCD-Bcurator, we contrasted results by hand and corrected chromosomes where CCD-Bcurator's output was not able to adequately correct records. Our hand-curated records were also contrasted against the ploidy dataset from Robertson et al. (2011) and against ploidy data in the C-value DNA dataset from Bennett and Leitch (2005). By comparing three different sources of information, we were able to code taxa as diploid (D) or polyploid (P). For the majority of species, ploidy was assigned according to information from the original publications included in the C-value DNA dataset (Bennett and Leitch 2005). For taxa without ploidy information but with information about chromosome number, we assigned ploidy based on the multiplicity of chromosomes within the genus or based on SI/SC classification. For example, if a species was incompatible without any chromosomal information it was assigned as diploid. This was the case of *Solanum betaceum* that did not include information about ploidy level but it has 24 chromosomes, since  $x = 12$  is the base chromosome number of the *Solanum* genus (Olmstead and Bohs 2007), we assigned *S. betaceum* as diploid. Species with more than one ploidy level were assigned the most frequent ploidy level recorded or the smallest ploidy in case of frequency ties. Breeding system was scored as self-incompatible (I) or self-compatible (C) based on results curated from the literature and original experimental crosses (as compiled in Igić et al. 2006; Goldberg et al. 2010; Robertson et al. 2011; Goldberg and Igić 2012). Many species could unambiguously be coded as either I or C (Raduski et al. 2012). But for diploids without breeding system information taxa was coded as both I or C. Following previous work, we coded as I any species with functional I systems, even if C or dioecy was also reported. Dioecious species without functional I were coded as C. To those existing data sets, we added some additional records for chromosome number and breeding system. The Supplementary Information contains citations for the numerous sources for the added data. Resolution of taxonomic synonymy followed the conventions provided in Solanaceae Source (PBI *Solanum* Project 2012). Hybrids and cultivars were excluded because ploidy and breeding system can be affected by artificial selection during domestication. Following the reasoning outlined in Robertson et al. (2011), we examined closely the few species for which the merged ploidy and breeding system data indicated the presence of self-incompatible polyploids. Although I populations frequently contain some C

individuals, and diploid populations frequently contain some polyploid individuals, in no case did we find a convincing case of a naturally occurring I and polyploid population. The single instance of an I and polyploid individual appears to be an allopolyploid hybrid of *Solanum oplocense* Hawkes x *Solanum gourlayii* Hawkes, reported by Camadro and Peloquin 1981. Under exceedingly rare circumstances, it is possible for polyploids containing multiple copies of S-loci to remain I, so long as they express a single allele at the S-locus (discussed in Robertson et al. 2011). Because of the resulting absence of I and polyploid populations, as well as the linked functional explanation for disabling of gametophytic self-incompatibility systems with non-self recognition, following whole genome duplication (reviewed in Ramsey and Schemske 1998; Stone 2002), we consider only three observed character states: self-incompatible diploids (ID), self-compatible diploids (CD), and polyploids which are always self-compatible (CP).

Matching our character state data to the largest time-calibrated phylogeny of Solanaceae (Särkinen et al. 2013) yielded 595 species with ploidy and/or breeding system information on the tree. Binary or three-state classification of ploidy and breeding system for the 595 taxa is summarized in Fig. 1. We retained all of these species in each of the analyses below because pruning away tips lacking breeding system in the ploidy-only analyses (and vice versa) would discard data that could inform the diversification models. A total of 405 taxa without any information about breeding system or polyploidy were excluded. Including this many more species would have prohibitively slowed our analyses, especially those implementing the most complex models.

### ***Models for ploidy and diversification***

In order to investigate the association between ploidy level and diversification, we first defined a binary state speciation and extinction model (BiSSE, Maddison et al. 2007) in which taxa were classified as diploid (D) or polyploid (P) (Fig. 1). We call this the *D/P ploidy* model. This model is identical to the model proposed by Mayrose et al. (2011, 2015), but inferences were performed in a Bayesian framework. We calculated the posterior probability distributions of speciation rates ( $\lambda_D$ ,  $\lambda_P$ ), extinction rates ( $\mu_D$ ,  $\mu_P$ ), net diversification rates ( $r_D = \lambda_D - \mu_D$ ,  $r_P = \lambda_P - \mu_P$ ), and relative extinction rates ( $v_D = \mu_D/\lambda_D$ ,  $v_P = \mu_P/\lambda_P$ ) associated with each state. Parameter  $\rho$  represents the transition rate from *D* to *P*.

Our second model addressed the diversification due to ploidy differences while also parsing out the heterogeneity of diversification rates due a possible unobserved trait. Diversification rate differences explained by some trait other than ploidy were accommodated by adding a hidden state (HiSSE model; Beaulieu and O'Meara 2016). In this model, each of the observed diploid and polyploid states were subdivided by a binary hidden trait with states *A* and *B*. This model was called the *D/P+A/B ploidy and*

170 *hidden state* model. Posterior probability distributions of speciation rates ( $\lambda_{DA}$ ,  $\lambda_{DB}$ ,  $\lambda_{PA}$ ,  $\lambda_{PB}$ ), extinction rates ( $\mu_{DA}$ ,  $\mu_{DB}$ ,  $\mu_{PA}$ ,  $\mu_{PB}$ ), net diversification rates ( $r_{DA}$ ,  $r_{DB}$ ,  $r_{PA}$ ,  $r_{PB}$ ), and relative extinction rates ( $v_{DA}$ ,  $v_{DB}$ ,  $v_{PA}$ ,  $v_{PB}$ ) were estimated. In this model polyploidization rate is again represented by  $\rho$  and hidden state rates are symmetrical with parameter  $\alpha$ .

### ***Models for breeding system and diversification***

175 To assess the effects of breeding system in the diversification process, we first fit model in which the states are self-incompatible (I) or self-compatible (C). This is the same as the analysis of Goldberg et al. (2010), but with an updated phylogeny (Särkinen et al. 2013). We call this BiSSE model the *I/C breeding system* model. To parse out the effect of breeding system on diversification, while allowing for the possibility of heterogeneous diversification rates unrelated to breeding system, we subdivided each of those states into  
180 hidden states *A* and *B*. We call this HiSSE model the *I/C+A/B breeding system and hidden state model*. The I/C+A/B model is equivalent to the model Freyman and Höhna (2017) used to study diversification in Onagraceae, but we restricted the hidden state transition rates to be symmetrical.

Self-incompatibility is homologous in all species in which S-alleles are cloned, and controlled crosses are performed. All species sampled to date, possess a non-self recognition, RNase-based, ga-  
185 metophytic self-incompatibility (shared even with other euasterid families; Ramanauskas and Igić 2017). Furthermore, species that are distantly related within this family carry closely-related alleles, with deep trans-specific polymorphism, at the S-locus, which controls the SI response (Ioerger et al. 1990; Igić et al. 2006). This represents very strong evidence that the self-incompatible mechanism, and our *I* state is ancestral to the Solanaceae, and did not arise independently within the family. Therefore, for all breeding system  
190 models, we estimated a transition rate from *I* to *C* but not the reverse ( $q_{CI} = 0$ ).

### ***Models for ploidy, breeding system, and diversification***

Polyploidy and breeding system might have changed lineage diversification individually but these two traits also have complex interactions in Solanaceae species. Therefore, a multi-state model that investigates the contribution of both traits and the possible transitions between them was proposed. Self-incompatible  
195 diploids (ID), self-compatible diploids (CD), and polyploids, which are always self-compatible (CP) were used as input for a MuSSE model (FitzJohn 2012). We did not include a state for self-incompatible polyploids because they are not observed in the data, and that trait combination state is mechanistically predicted not to occur. The proposed model is called the *ID/CD/CP ploidy and breeding system* and has six parameters for diversification in each state: ( $\lambda_{ID}$ ,  $\lambda_{CD}$ ,  $\lambda_{CP}$  for speciation,  $\mu_{ID}$ ,  $\mu_{CD}$ ,  $\mu_{CP}$  for extinction), and three  
200 for transitions between states ( $\rho_I$ ,  $\rho_C$  for polyploidization transitions from *ID* and *CD* to *CP*, respectively;



$q_{IC}$  for loss of self-incompatibility without polyploidization, from *ID* to *CD*). The total rate of loss of self-incompatibility, i.e., transitions out of *ID*, is  $q_{IC} + \rho_I$ . Diploidization from *CP* to *ID* is not allowed because it would represent a simultaneous regain of SI.

The *ID/CD/CP* model could also suffer from a high type I error as the regular two state BiSSE model. A hidden factor might alter diversification beyond both ploidy and breeding system misleading inferences (Rabosky and Goldberg 2015). We therefore added a hidden trait layer on top of our three-state model (analogous to Caetano et al. 2018; Huang et al. 2018). We refer to this as the *ID/CD/CP+A/B* model. A fully parameterized version of this model would have 26 rate parameters. Because our goal was to look for diversification rate differences associated with ploidy and breeding system rather than the specific effects of the hidden states, we fitted a simplified version with 15 parameters. The reduction in parameter space is achieved by fixing the rates for transitions among hidden states to be equal with rate  $\alpha$ , and fixing the transition rates between observed states to be independent of the hidden state (rates  $\rho_I$ ,  $\rho_C$ ,  $\delta$ ,  $q_{IC}$  as defined for the *ID/CD/CP* model). There are additionally twelve diversification rate parameters ( $\lambda_{ID_A}$ ,  $\lambda_{ID_B}$ ,  $\lambda_{CD_A}$ ,  $\lambda_{CD_B}$ ,  $\lambda_{CP_A}$ ,  $\lambda_{CP_B}$ ,  $\mu_{ID_A}$ ,  $\mu_{ID_B}$ ,  $\mu_{CD_A}$ ,  $\mu_{CD_B}$ ,  $\mu_{CP_A}$ ,  $\mu_{CP_B}$ ).

### Pathways to polyploidy

Considering ploidy and breeding system together, there are two evolutionary pathways from SI diploid to SC polyploid (Brunet and Liston 2001; Robertson et al. 2011). In the one-step pathway, the *CP* state is produced directly from the *ID* state when whole genome duplication disables SI. In the two-step pathway, the *CD* state is an intermediate: SI is first lost, and later the *SC* diploid undergoes polyploidization. We quantify the relative contribution of these pathways to polyploidy in two ways, each using the maximum a posteriori (MAP) estimates of rates from the *ID/CD/CP* model.

Both of our methods are based on a propagation matrix that describes flow from *ID* to *CP*, as in Robertson et al. (2011). We insert an artificial division in the *CP* state, so that one substate contains the *CP* species that arrived via the one-step pathway and the other substate contains the *CP* species that arrived via the two-step pathway. We consider only unidirectional change along each step of the pathway in order to separate them into clear alternatives, and because in this family there is no support for regain of SI, and not strong support for diploidization. First, we consider only the rates of transitions between these states, placing them in the propagation matrix  $Q$ . The matrix  $P = \exp(Qt)$  then provides the probabilities of changing from one state to any other state after time  $t$ . Closed-form solutions for the two pathway probabilities are provided in Robertson et al. (2011). Our results will differ from those of Robertson et al. (2011) because our transition rate estimates come from a dated phylogeny and a model that allows for state-dependent diversification.

Second, we consider not only transitions between states but also diversification within each state. State-dependent diversification can change the relative contributions of the two pathways. In particular, if the net diversification rate is small for CD, the two-step pathway will contribute relatively less. We therefore include the difference between speciation and extinction along the diagonal elements of the propagation matrix. As before, matrix exponentiation provides the relative chance of changing from one state to any other state after time  $t$ . In this case, however, these are not probabilities because diversification changes the number of lineages as time passes. We can still use their ratios to consider the relative contribution of each pathway, though, analogous to the normalized age structure in a growing population.

#### *Diploidization as an exploratory hypothesis*

For all four models that consider ploidy changes, we allowed diploidization by adding a parameter  $\delta$  connecting polyploidy state back to diploid or to self-compatible diploid in the multi-state models (denoted by  $+/delta$  in the models of Table 1). Previous modeling approaches (Mayrose et al. 2011) have argued against inferring diploidization rates when using ploidy data that comes from classifications based on chromosome number multiplicity or chromosome number change models like chromEvol (Mayrose et al. 2010; Glick and Mayrose 2014; Mayrose et al. 2015; Freyman and Höhna 2017). These types of classifications do not allow for a ploidy reversion. Where indicated, the classification of ploidy for the data used in our models was based on chromosome multiplicity at the genus level (data available at <https://www.github.com/roszenil/solploidy/datasets>). However, the majority of the ploidy classifications were adopted from original studies with alternative sources of information (e.g., geographic distribution, genus ploidy distribution) where ploidy was defined by authors that found evidence for it. Since it is not clear whether diploidization can be detected under alternative ploidy classifications or even classifications based on chromosome number multiplicity at the genus level, we also fit the models without diploidization in order to test whether the conclusions about diversification are sensitive to including diploidization. As discussed by Servedio et al. (2014), the presence or absence of a hypothesis can have an exploratory goal. In our case the diploidization parameter (or its absence,  $\delta = 0$ ) in our models is an opportunity to explore an assumption that might be important but that is not the single definitive process to understand the interactions among polyploidy, breeding system, and diversification.

#### *Statistical inference under the models*

Parameters for each of the 10 diversification models were estimated using RevBayes (Höhna et al. 2016). Code for analyses and key results are available at <https://github.com/roszenil/solploidy>. We accounted for incomplete sampling in all analyses by fixing the probability of sampling a species at the present

s to 595/3000 since the Solanaceae family has approximately 3,000 species as estimated by the Solanaceae Source project (PBI *Solanum* Project 2012). For all 10 models, we assumed that speciation and extinction parameters had log-normal prior distributions with means equal to the expected net diversification rate (number of taxa/[2 × root age]) and standard deviation 0.5. Priors for parameters defining trait changes were assumed to be gamma distributed with parameters  $k = 0.5$  and  $\theta = 1$ . For each model, a Markov chain Monte Carlo (MCMC; Metropolis et al. 1953; Hastings 1970) was run for 96 hours in the high-performance computational cluster at the Minnesota Supercomputing Institute, which allowed for 5,000 generations of burn-in and a minimum of 200,000 generations of MCMC for each of the 10 models. Convergence and mixing of each MCMC was determined by ensuring the effective sample size of each parameter was over 200.

For each model, ancestral states at each node in the phylogeny were sampled jointly during the MCMC analyses. Ancestral state reconstructions for all models can be found in the supplementary information and show the maximum a posteriori of the marginal probability distributions for each of the 594 internal nodes for each of the 10 models.

### **Model selection**

We calculated the marginal likelihood for each of the 10 models in RevBayes (Höhna et al. 2016). Marginal likelihoods were calculated using 50 stepping stone steps under the methodology of Xie et al. (2010). Each stepping stone step was found by calculating 500 generations of burn-in followed by a total of 1,000 MCMC steps (Table 1). The calculation of each marginal likelihood ran for 24 hours on a high-performance computational cluster.

Using the marginal likelihood values, we calculated thirteen different Bayes factors. Six compared the models of ploidy against one other (D/P and D/P+A/B, each with or without diploidization), one compared the breeding system models (I/C and I/C+A/B), and six compared the models with both traits (ID/CD/CP and ID/CD/CP+A/B, each with or without diploidization) (Table 2). Other comparisons between these models are not valid because the input data are different under the different state space codings (Fig. 1). In mathematical terms, the D/P, I/C, and ID/CD/CP state spaces are not ‘lumpable’ with respect to one another (Tarasov 2018). Each model comparison is reported with a Bayes factor on the natural log scale: the comparison between models  $M_0$  and  $M_1$  is  $BF(M_0, M_1) = \ln[P(\mathbf{X}|M_0) - P(\mathbf{X}|M_1)]$ . There is ‘positive’ support for  $M_0$  when this value is more than 2, ‘strong’ support when it is more than 6, and ‘very strong’ support when it is more than 10 (Kass and Raftery 1995).

## Results

### *Traits and diversification*

Individual trait-dependent diversification inference, of ploidy and breeding system considered separately, each showed significant associations with net diversification differences. When considered together, however, the effect of breeding system dominated the effect of ploidy, although hidden factors played an important role as well.

When considering ploidy alone (D/P model), we found a greater net diversification rate for diploids than for polyploids, in agreement with (Mayrose et al. 2011, 2015). This result holds with (Fig. 2A) or without (Fig. 3A) the diploidization parameter, though including diploidization shifts the net diversification rate of polyploids to be non-negative. Incorporating a hidden state, however, removes the clear separation in diversification between diploids and polyploids (D/P+A/B model; Fig. 2B, Fig. 3B). Thus, differences in net diversification are better explained by an unknown factor than by ploidy. Statistical model comparisons show very strong support for including the hidden state and strong support for including diploidization (Table 2).

When considering breeding system alone (I/C model, Fig. 2C), we found a larger net diversification rate for SI than for SC species, in agreement with Goldberg et al. (2010). When a hidden state is included (I/C+A/B model), the large net diversification difference persists for one of the hidden states but is removed for the other (Fig. 2D). Thus, differences in net diversification are best explained by both breeding system and an unknown factor, similar to results found by Freyman and Höhna (2018). The statistical model comparison shows very strong support for including the hidden state (Table 2).

When considering ploidy and breeding system together (ID/CD/CP model), the net diversification rate for SI diploids was greater than for either SC diploids or SC polyploids, with or without diploidization (Fig. 2E, Fig. 3C). It thus appears that the difference in net diversification with breeding system persists when ploidy is included in the model, but not the reverse. The association of ploidy with net diversification in the D/P model (Fig. 2A, Fig. 3A) appears to be driven by the subset of diploids that are SI; among SC species, net diversification rates for diploids and polyploids are similar. When a hidden state is included (IC/CD/CP+A/B model), the same general pattern remains when diploidization is prevented (Fig. 3D), although the higher net diversification rate of ID is less clear within one of the hidden states. With diploidization, the net diversification rate of ID is still greater than CD within each hidden state, but diversification for P is highly uncertain and perhaps bimodal. Statistical model comparisons show very strong support for including the hidden state and at most positive support for including diploidization (Table 2).

### ***Pathways to polyploidy***

325 Evolution from self-incompatible diploids to self-compatible polyploids can proceed through two different pathways. Determining the relative contribution of these pathways based on our estimated transition rates from the ID/CD/CP model, we find that the one-step pathway is more likely on short timescales and the two-step pathway is more likely on long timescales (Fig. 4, left panels). When not much time has elapsed, the one-step pathway is more likely because only one event need happen. When more time has elapsed, the  
330 two-step pathway is more likely because the rate of loss of SI within diploids,  $q_{IC}$ , is greater than the rate of polyploidization for SI species,  $\rho_I$  (Fig. S7). These conclusions are the same as those of Robertson et al. (2011) under one of their branch length approximations.

When rates of lineage diversification are considered, however, the conclusions change. Even over long timescales, the two-step pathway contributes less (Fig. 4, right panels). The lower rate of net diver-  
335 sification in the CD state, relative to ID, means that relatively fewer lineages are available to complete the second step of the two-step pathway.

### ***Diploidization as an exploratory hypothesis***

We considered models both with or without diploidization in order to explore the effects of this process on the estimates of state-dependent diversification. For the two models that only include diploid and poly-  
340 ploid states (D/P and D/P+A/B), the net diversification rate of polyploid lineages is likely negative when diploidization (parameter  $\delta$ ) is excluded but positive when  $\delta$  is included (Fig. 2A versus Fig. 3A). For the two models that also included breeding system (ID/CD/CP and ID/CD/CP+A/B), the main effect of including diploidization is increasing the uncertainty of the estimate of polyploid net diversification rate (Fig. 2EF versus Fig. 3CD). For all models, there is greater uncertainty in the estimate of diploidization rate than  
345 polyploidization rate, as judged by the width of the credibility intervals (see supplementary information figures).

## **Discussion**

The present work shows the importance of considering the trait linked diversification patterns under a multivariate approach. Species are created and go extinct based on multiple and often highly correlated phe-  
350 notypes, understanding the speciation and extinction processes requires understanding of the evolutionary consequences that those trait correlations produce in organisms. In the present work we show how considering both polyploidy and breeding system can disentangle the importance (or the lack of) of polyploidy when confronted with the evidence brought by breeding system.

Using the most complete dataset for polyploidy in a phylogenetic tree in Solanaceae, we were able  
355 to replicate the results found by Mayrose et al. (2011), polyploids have a slower net diversification compare  
to diploids. Furthermore, we also found polyploids had a high probability of having negative diversification  
which implies that polyploids can become a macroevolutionary dead-end, a result that was also found in  
the two large angiosperm diversification studies Mayrose et al. (2011) and Mayrose et al. (2015). However,  
we expanded this study to accommodate background heterogeneity in the diversification process. When  
360 adding heterogeneity we found that it was more likely that an unobserved trait linked to diploid state was  
the one leading the net diversification patterns, and that there were some “second-class” diploids that were  
not different from polyploids in diversification terms (Figure 2A). This result lead us to our central question:  
*what is that other trait linked diploids that makes them different in the diversification process?* .

In Solanaceae, our immediate intuition was to look into breeding system. Previous studies shown  
365 that self-incompatible Solanaceae species have also higher rates of diversification compared to their self-  
compatible counterparts (Goldberg and Igić 2012). Self-incompatible species are diploid in our sample and  
also expected to be diploid due to ... (citation?). By considering both polyploidy and breeding system  
simultaneously for every species in our sample, it was possible to disentangle why some diploids were  
quantitatively different than polyploids. In the three-state diversification model ID/CD/CP, we found that  
370 self-incompatible diploids have faster and positive rates than self-compatible diploids and polyploids, and  
that the difference between the rates of net diversification of self-compatible diploids is not as large (Figures  
2C) as first found by binary trait diversification models (Figures 2A). This result is important, since it aligns  
with the net diversification results of the D/P+A/B model, where a ‘ ‘hidden-trait” seem to be dictating the  
diversification pattern. By adding breeding system, we were able to hint at which that hidden-trait possibly  
375 is. Therefore, we consider that finding a heterogenous result in the hidden trait approaches should be be  
treated as evidence of a second trait that is necessary to consider. Pursuing knowledge of such trait can  
result on a clearer picture of the importance of trait linked diversification patterns, but also on a better  
reconstruction on past events in phylogenies.

Our work shows that a diploid Solanaceae lineage is much more likely to take a one-step ID/CP  
380 pathway to polyploidy rather than a two-step ID/CD/CP pathway. We modeled both of these pathways  
as anagenetic character changes occurring within a species. However Goldberg and Igić (2012) showed  
that I/C transitions are often associated with speciation events, and similarly Freyman and Höhna (2017)  
demonstrated that D/P transitions may also be associated with speciation events. Models that fail to consider  
transitions that occur at speciation (cladogenetic changes) may make misleading estimates of anagenetic  
385 transition rates. The HiSSE-based models introduced in our work here could be extended to incorporate

cladogenetic transitions. It is possible that allowing for cladogenetic transitions may enable the one-step pathway to even further dominate over the two-step pathway to polyploidy, but this remains to be tested in future work.

Diploidization needs to be considered in diversification models because higher rates of net diversification (speciation minus extinction) in diploids can be obtained from models that ignore the possibility of a polyploid lineage that has diploidized being the diversification enhancer. Hence, in a diploidization lacking model, diploids will show higher rates of diversification when in fact it is a polyploidy event and its subsequent diploidization that generated higher speciation and less extinction. By adding diploidization to models of polyploidy linked to diversification it is possible to recover this complicated scenario and to reconcile genomic evidence with stochastic models.

Diploidization is widely considered to be relatively rare, compared with polyploidization, but flowering plant lineages are thought to have experienced at least one round of polyploidization in their evolutionary history. As a consequence, it is possible that nearly all extant species classified as “diploid” in our analyses are possibly secondary diploids, having undergone both polyploidization and re-diploidization. Genomic evidence indicates that such an event may have occurred prior to the origin of the family Solanaceae. Ku et al. (2000) and Blanc and Wolfe (2004) posited that the lineage leading to tomato, *Solanum lycopersicum*, may have experienced one or more paleopolyploid events. A subsequent analysis of synteny between grape and *Solanum* genomes, as well as the distribution between inferred paralogs within *Solanum* (tomato and potato) genomes both suggest that this lineage experienced a likely round of ancient genome duplication or triplication (Tomato Genome Consortium 2012). The age of the peak of paralog Ks distances, is approximately 71 million years (Tomato Genome Consortium 2012). If this is the case, then all of the genomes may have been subsequently re-diploidized, because common base chromosome numbers this and related families are  $n=11-12$ . On the other hand, studies comparing map-based genome synteny within the family find no evidence for recent polyploidization-diploidization (Wu and Tanksley 2010). Simple genome re-arrangements appear sufficient to explain chromosomal evolution between a number of species, including all of those in the relatively cytogenetically conserved ‘ $x=12$ ’ group, which includes tomato, potato, eggplant, pepper, and tobacco.

### Acknowledgements

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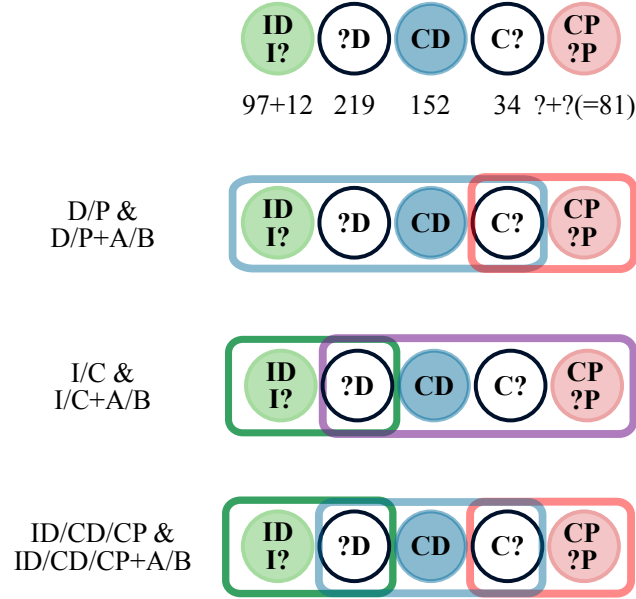


Figure 1: Character states used for each of the models. Each species retained on the tree belonged to one of five possible categories, depending on whether ploidy and/or breeding system were known. The number of species in each is shown under the corresponding circles in the top row. These categories were then grouped in a manner appropriate to the states of each model. For example, there are 34 species that are self-compatible and of unknown ploidy; these are coded as either *D* or *P* in the D/P models (uncertain, or consistent with either state), as *C* in the I/C models, and as either *CD* or *CP* in the ID/CD/CP models. In all cases, species were coded as either *A* or *B* in the hidden state models. The models depicted in the different rows use state spaces that are not comparable with one another. For example, we cannot test whether the D/P model fits better than the I/C model because they use states that are not the same and are not ‘lumpable’ (Tarasov 2018).

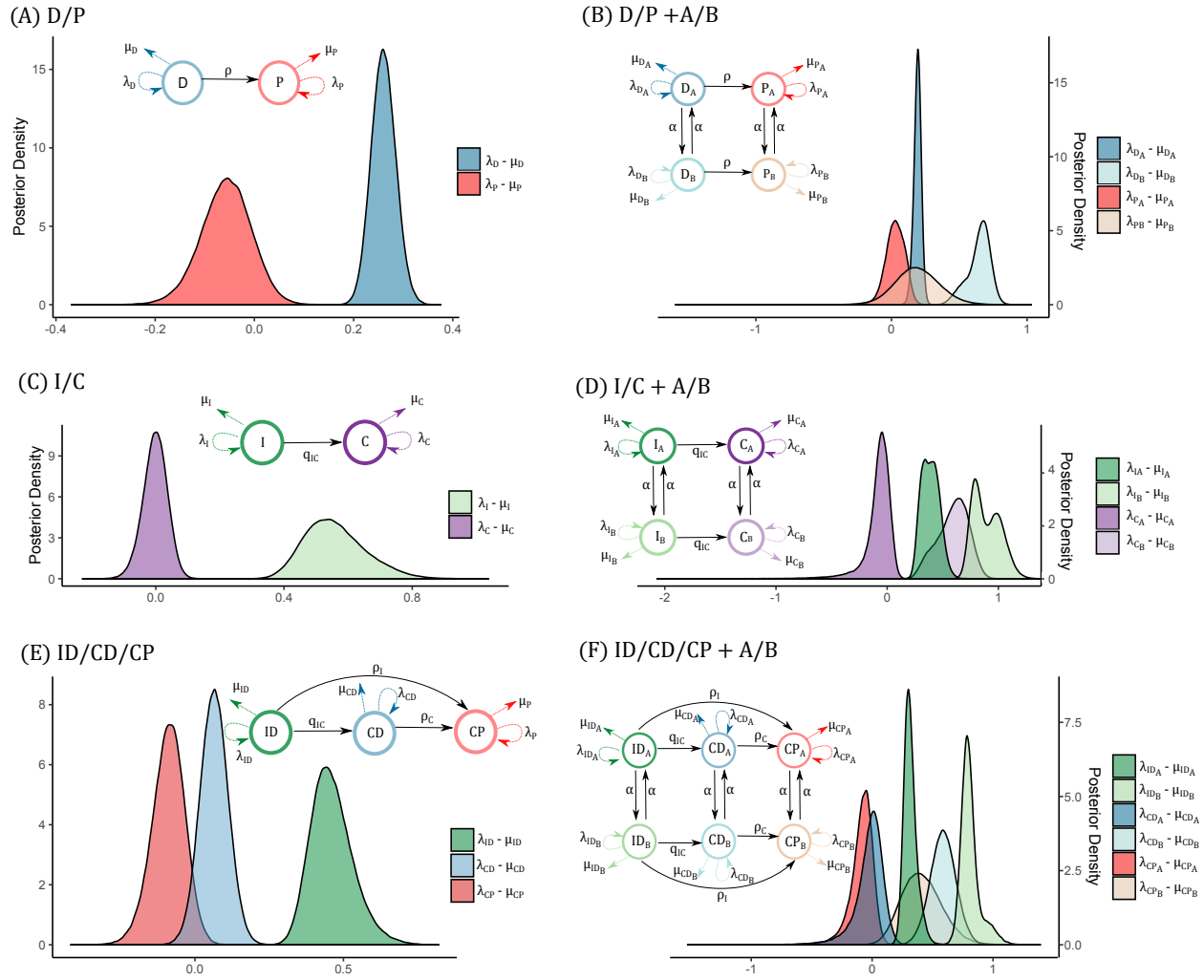


Figure 2: Net diversification rates for all models.

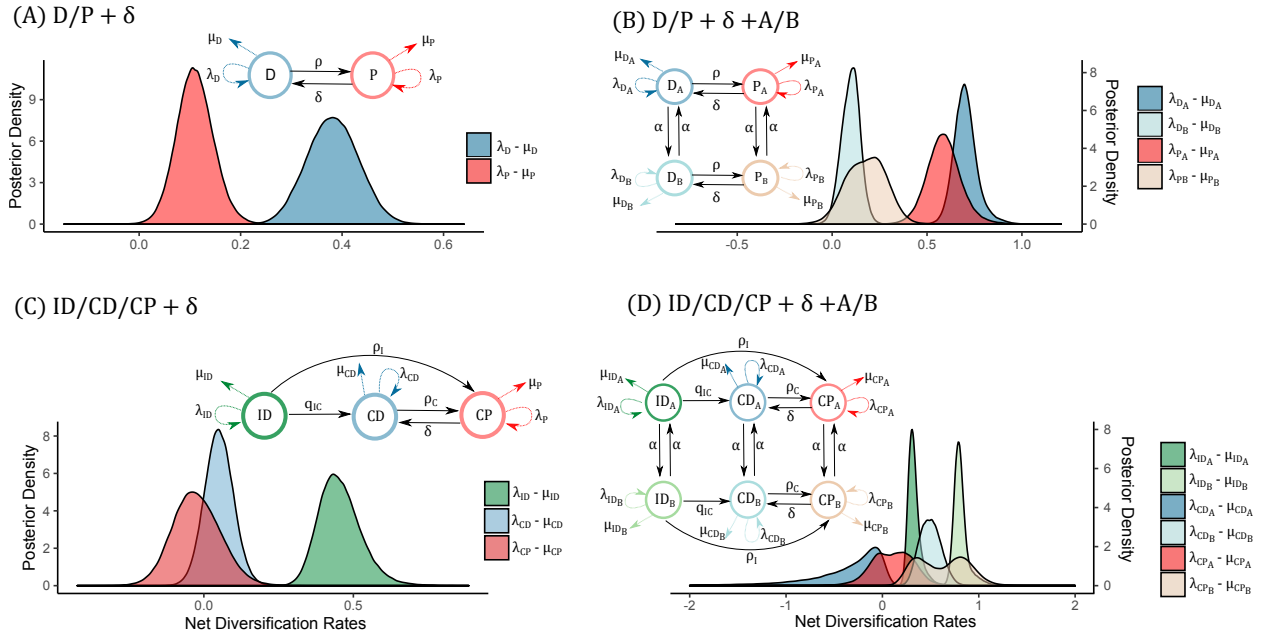


Figure 3: Net diversification rates for all models that include diploidization.

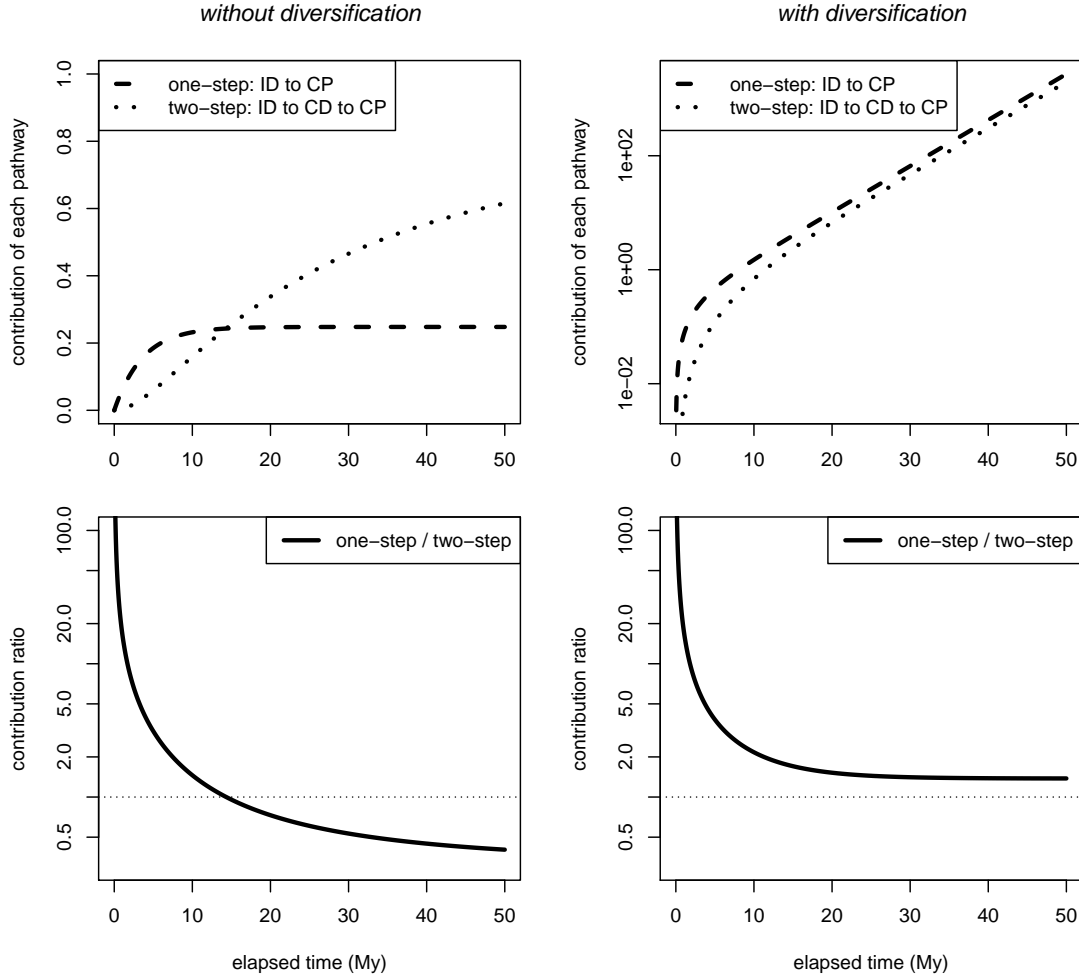


Figure 4: Contributions of the two pathways to polyploidy, the one-step ID→CP transitions, and the two-step, ID→CD→CP transitions. Panel (a) shows the contributions when considering only transition rates among the states (ignoring the diversification rate parameters). The one-step pathway dominates on short timescales and the two-step on long timescales. Panels (b) shows the contributions when including diversification of lineages in each state. The one-step pathway, in which polyploidy breaks down SI, dominates over any timescale. Panels (c) and (d) plot the ratio of the pathway contributions in the upper panels, (a) and (b), respectively.

| Model                       | Ploidy | Diploidization | Breeding System | Hidden State | Num Parameters | Marginal Log-Likelihood |
|-----------------------------|--------|----------------|-----------------|--------------|----------------|-------------------------|
| 1. D/P                      | Yes    | No             | No              | No           | 5              | -1193.66                |
| 2. D/P + $\delta$           | Yes    | Yes            | No              | No           | 6              | -1182.93                |
| 3. D/P+A/B                  | Yes    | No             | No              | Yes          | 10             | -1150.99                |
| 4. D/P+ $\delta$ +A/B       | Yes    | Yes            | No              | Yes          | 11             | <b>-1145.69</b>         |
| 5. I/C                      | No     | No             | Yes             | No           | 5              | -1194.80                |
| 6. I/C+A/B                  | No     | No             | Yes             | Yes          | 10             | <b>-1155.37</b>         |
| 7. ID/CD/CP                 | Yes    | No             | Yes             | No           | 9              | -1345.87                |
| 8. ID/CD/CP + $\delta$      | Yes    | Yes            | Yes             | No           | 10             | -1344.50                |
| 9. ID/CD/CP+A/B             | Yes    | No             | Yes             | Yes          | 15             | -1303.55                |
| 10. ID/CD/CP+ $\delta$ +A/B | Yes    | Yes            | Yes             | Yes          | 16             | <b>-1300.35</b>         |

Table 1: The ten models and their marginal likelihoods. Values in bold are for the best models within each class that are comparable (see Table 2). Abbreviations are D: diploid, P: polyploid, I: self-incompatible, C: self-compatible, A: one state of hidden trait, B: other state of hidden trait,  $\delta$ : diploidization.



| Ploidy Models          |   |       |        |        | Breeding System Models |   | Ploidy and Breeding System Models |                              |   |      |        |        |
|------------------------|---|-------|--------|--------|------------------------|---|-----------------------------------|------------------------------|---|------|--------|--------|
|                        | 1 | 2     | 3      | 4      |                        | 5 | 6                                 |                              | 7 | 8    | 9      | 10     |
| 1. D/P                 | · | 10.72 | -37.24 | -31.94 | 5. I/C                 | · | -39.43                            | 7. ID/CD/CP                  | · | 1.36 | -44.15 | -40.95 |
| 2. D/P no $\delta$     | · | ·     | -47.97 | -42.66 | <b>6. I/C+A/B</b>      | · | ·                                 | 8. ID/CD/CP no $\delta$      | · | ·    | -45.51 | -42.31 |
| <b>3. D/P+A/B</b>      | · | ·     | ·      | 5.30   |                        |   |                                   | <b>9. ID/CD/CP+A/B</b>       | · | ·    | ·      | 3.2    |
| 4. D/P+A/B no $\delta$ | · | ·     | ·      | ·      |                        |   |                                   | 10. ID/P/CD no $\delta$ -A/B | · | ·    | ·      | ·      |

Table 2: Bayes factors for model comparisons. Each of the three boxes contains models that can be compared with one another, based on the character states they include (see Fig. 1). Models are numbered in as Table 1. Bayes factors are reported on the natural log scale, so numbers greater than +2 mean that the model in the row label has ‘positive’ support relative to the model in the column label; numbers less than −2 mean that model in the column label is the preferred one. Conventional thresholds for ‘strong’ and ‘very strong’ support are 6 and 10, respectively. The best model in each set is written in bold. In each case, it is the most complex model of the set.

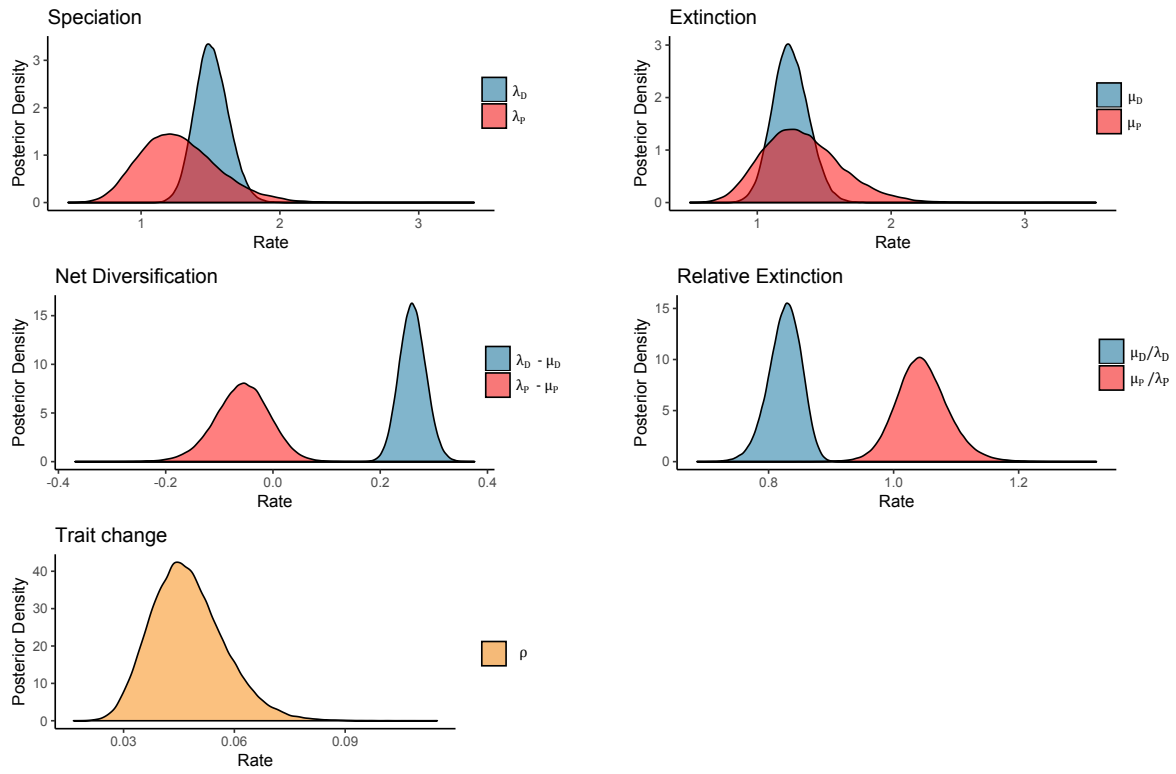


Figure S1: Posterior distribution for each of the parameters in the D/P, polyploidy model

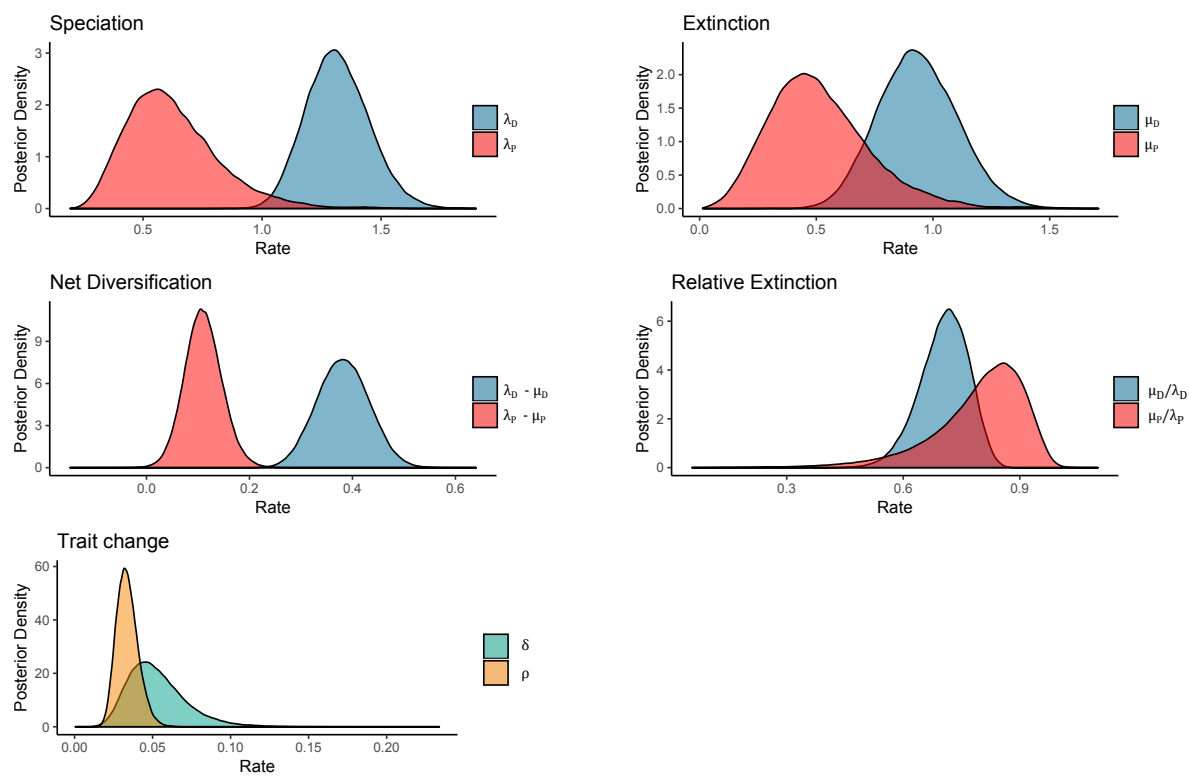


Figure S2: Posterior distribution for each of the parameters in the D/P+  $\delta$ ,polyploidy model

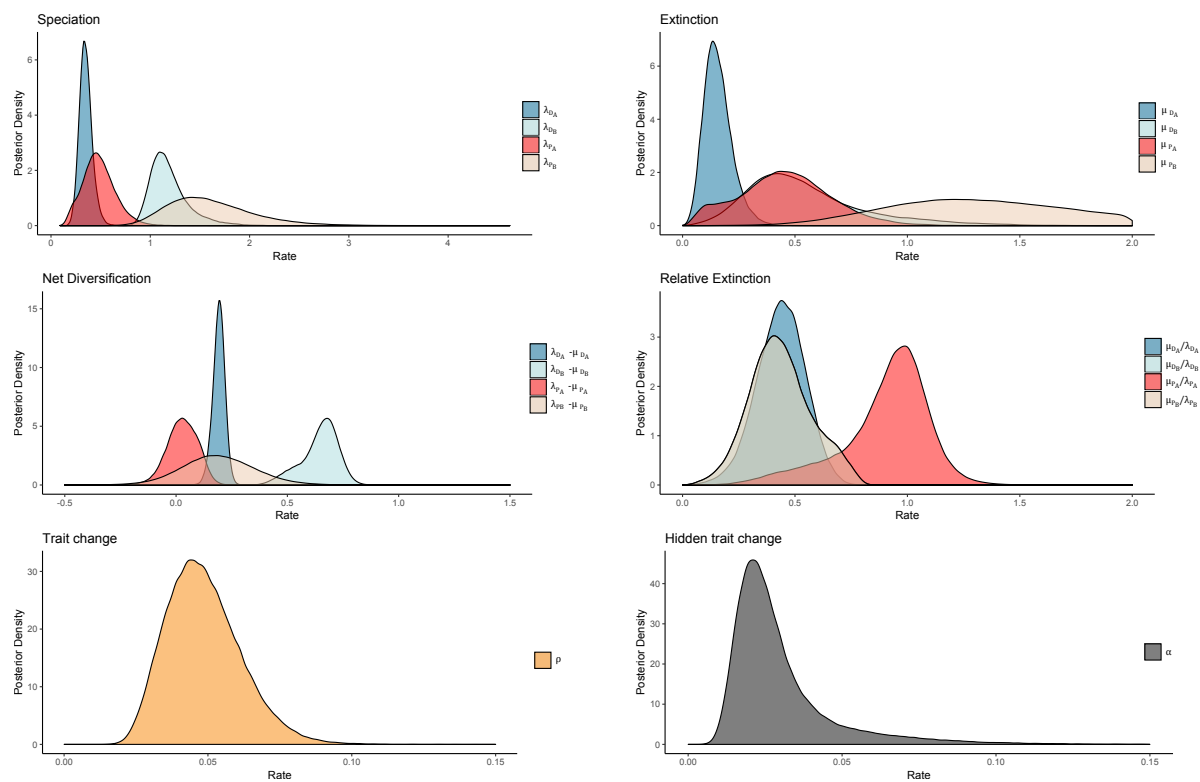


Figure S3: Posterior distribution for each of the parameters in the D/P+A/B, polyploidy model

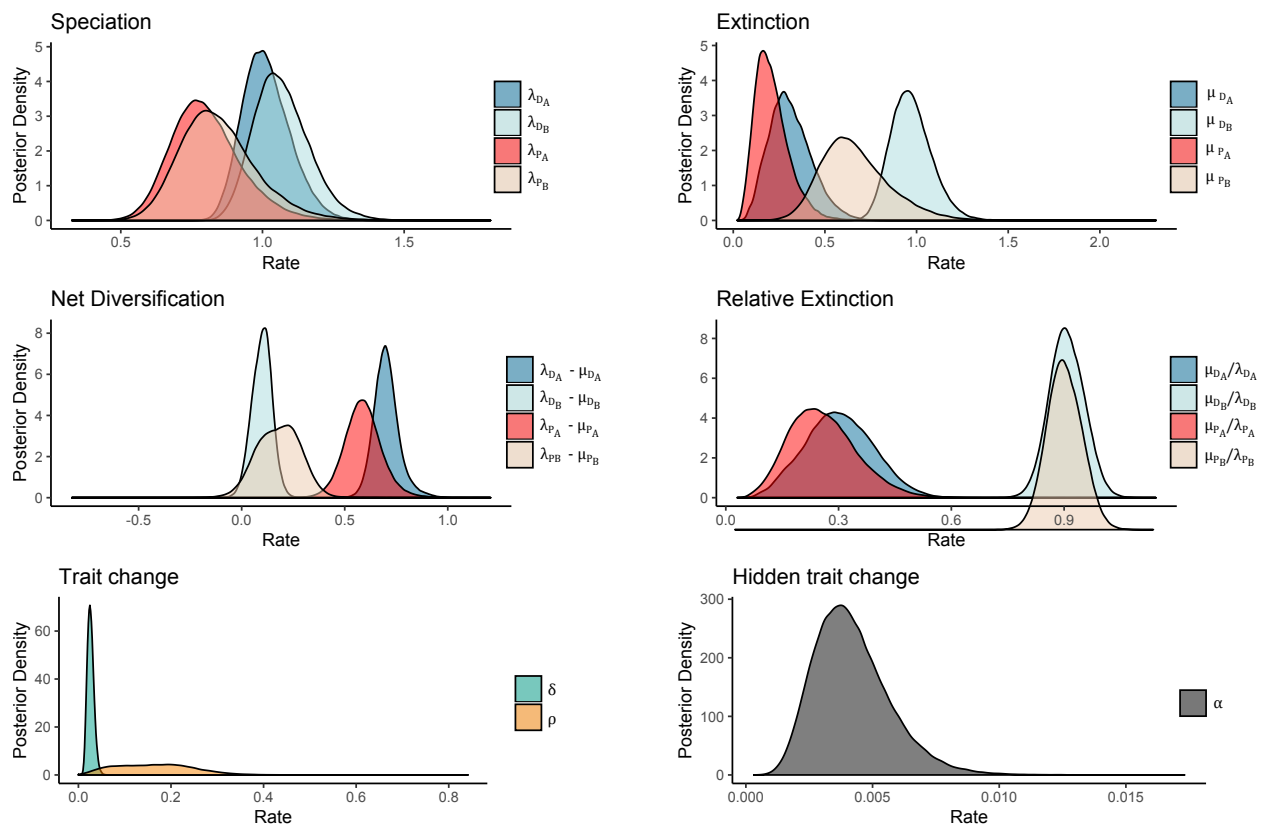


Figure S4: Posterior distribution for each of the parameters in the D/P+ $\delta$ +A/B, polyploidy model. The axis is offset in one location so that the two overlapping distributions can be seen.

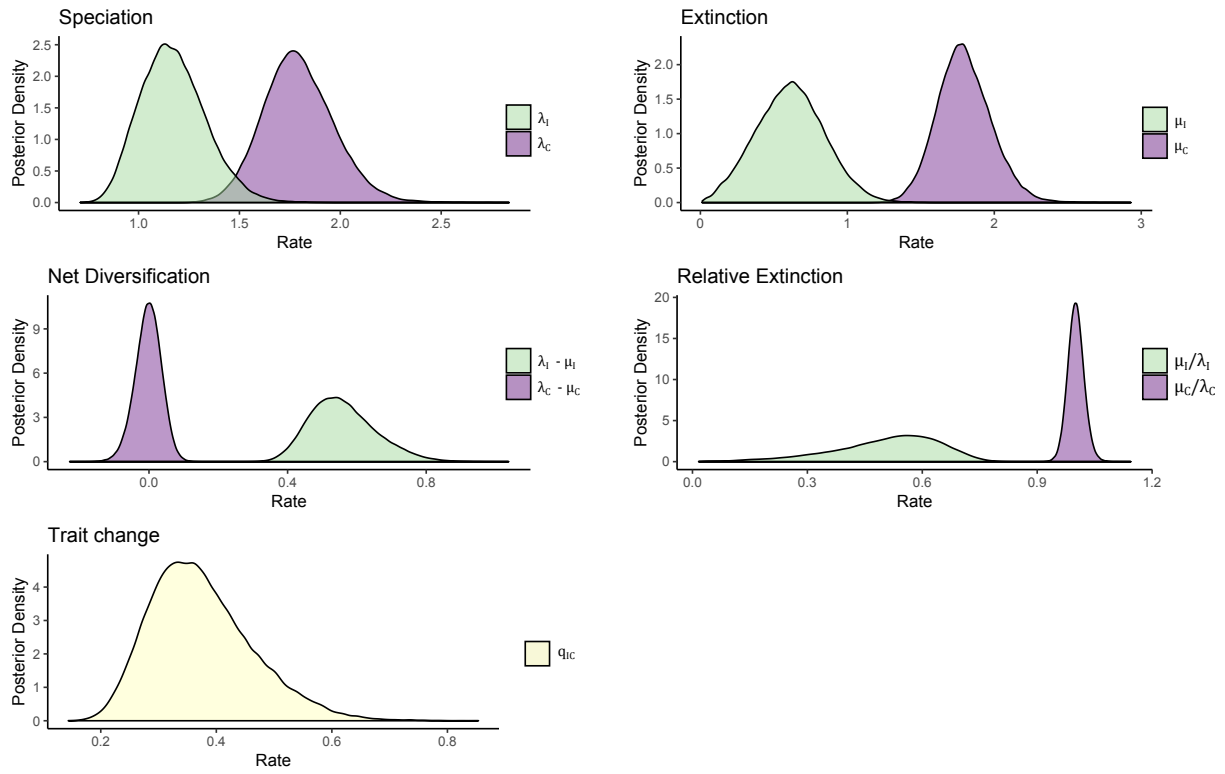


Figure S5: Posterior distribution for each of the parameters in the I/C, breeding system model

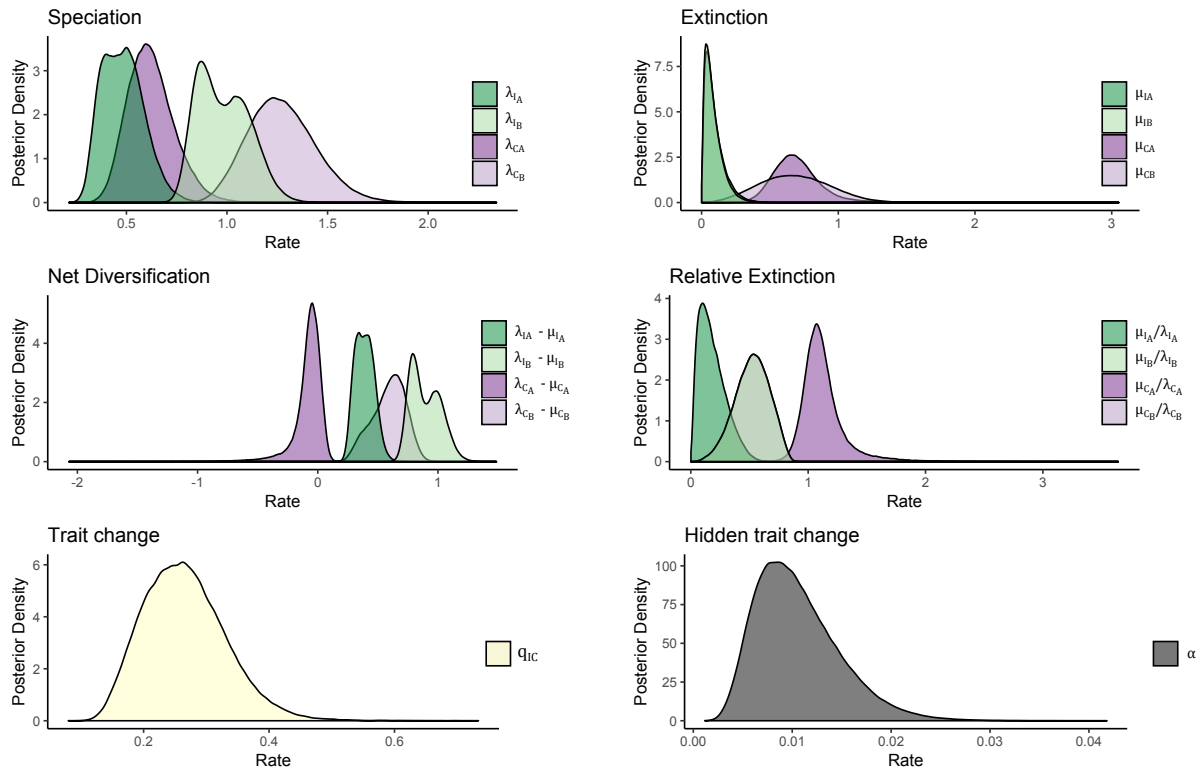


Figure S6: Posterior distribution for each of the parameters in the I/C+A/B, breeding system model

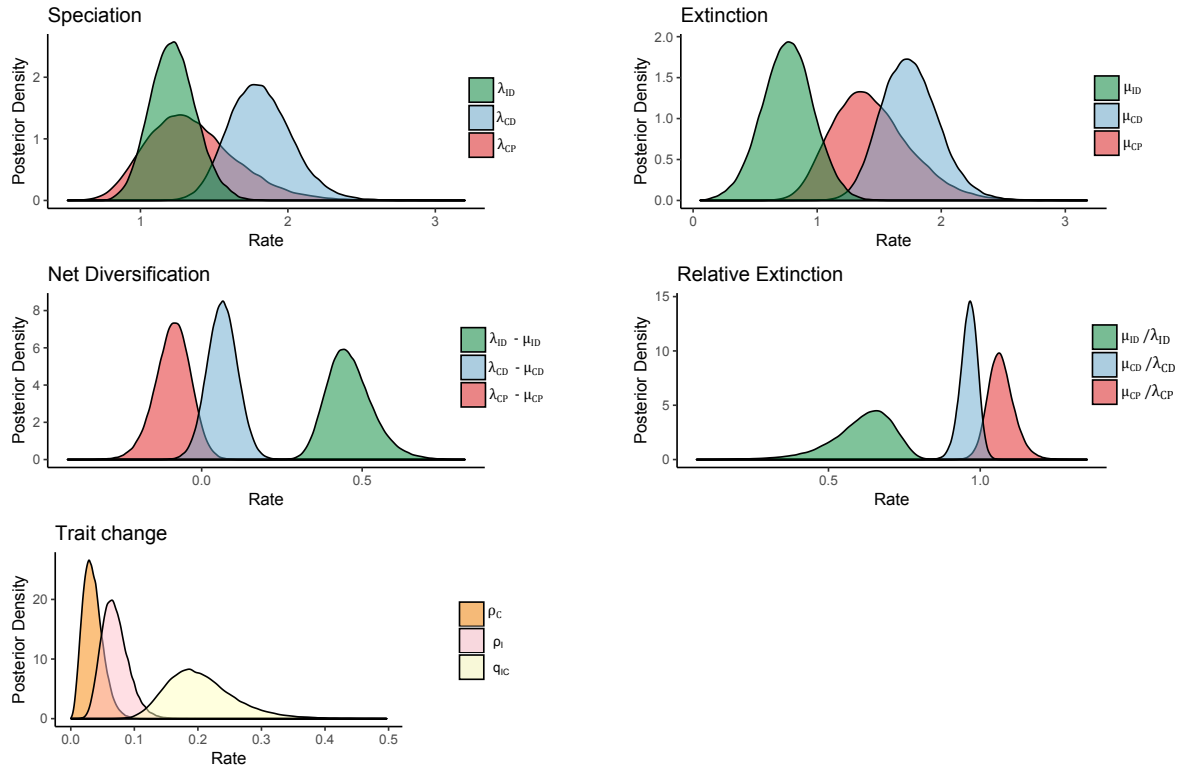


Figure S7: Posterior distribution for each of the parameters in the ID/CD/CP, polyploidy and breeding system model



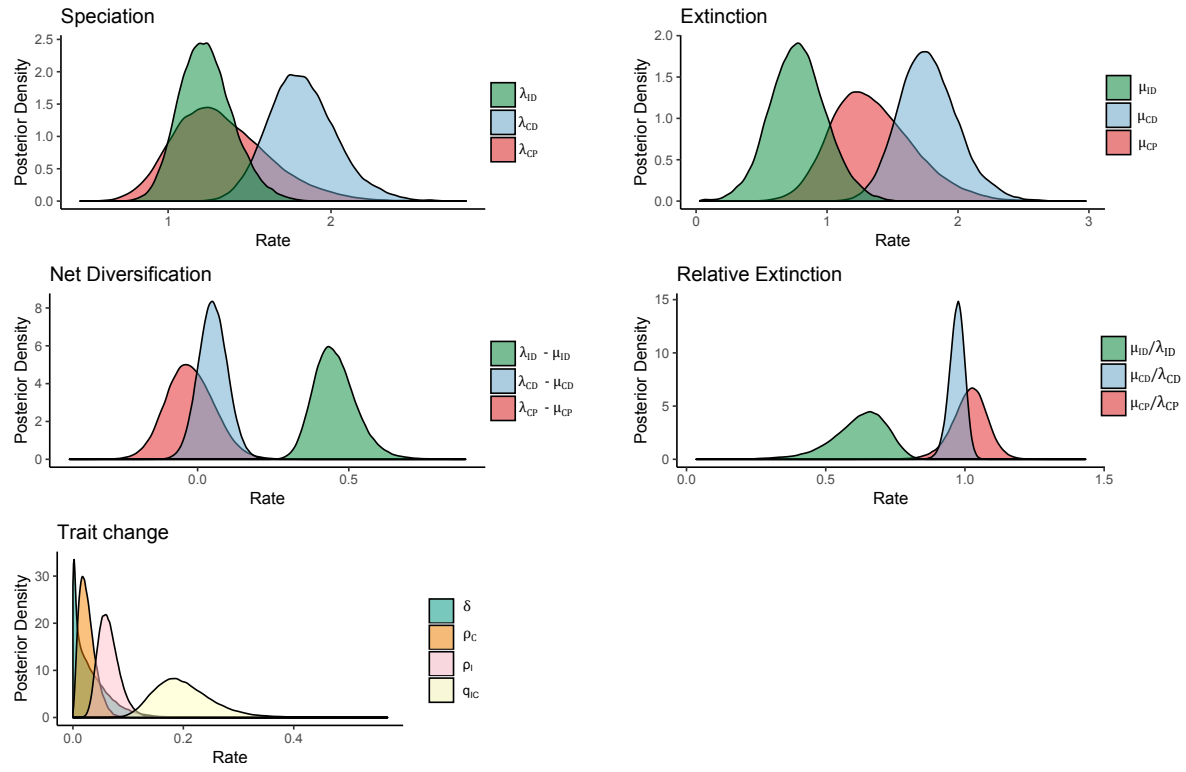


Figure S8: Posterior distribution for each of the parameters in the ID/CD/CP+ $\delta$ , polyploidy and breeding system model

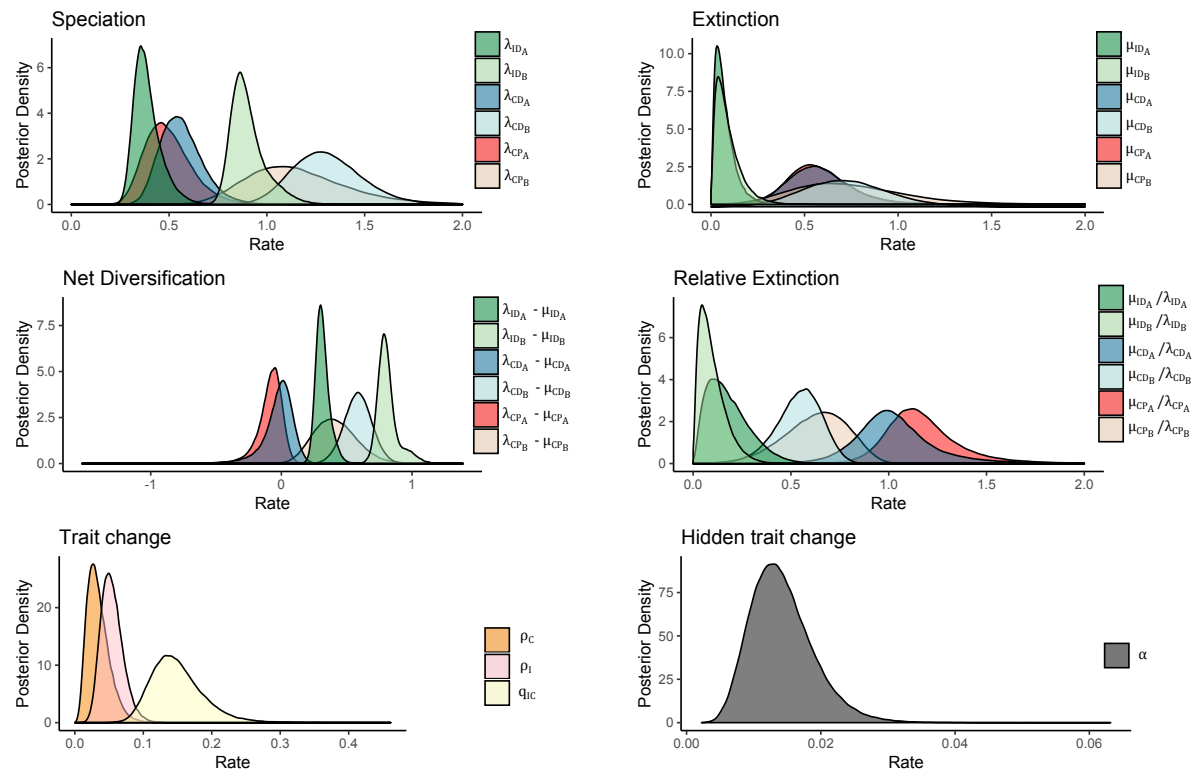


Figure S9: Posterior distribution for each of the parameters in the ID/CD/CP+A/B polyploidy and breeding system model

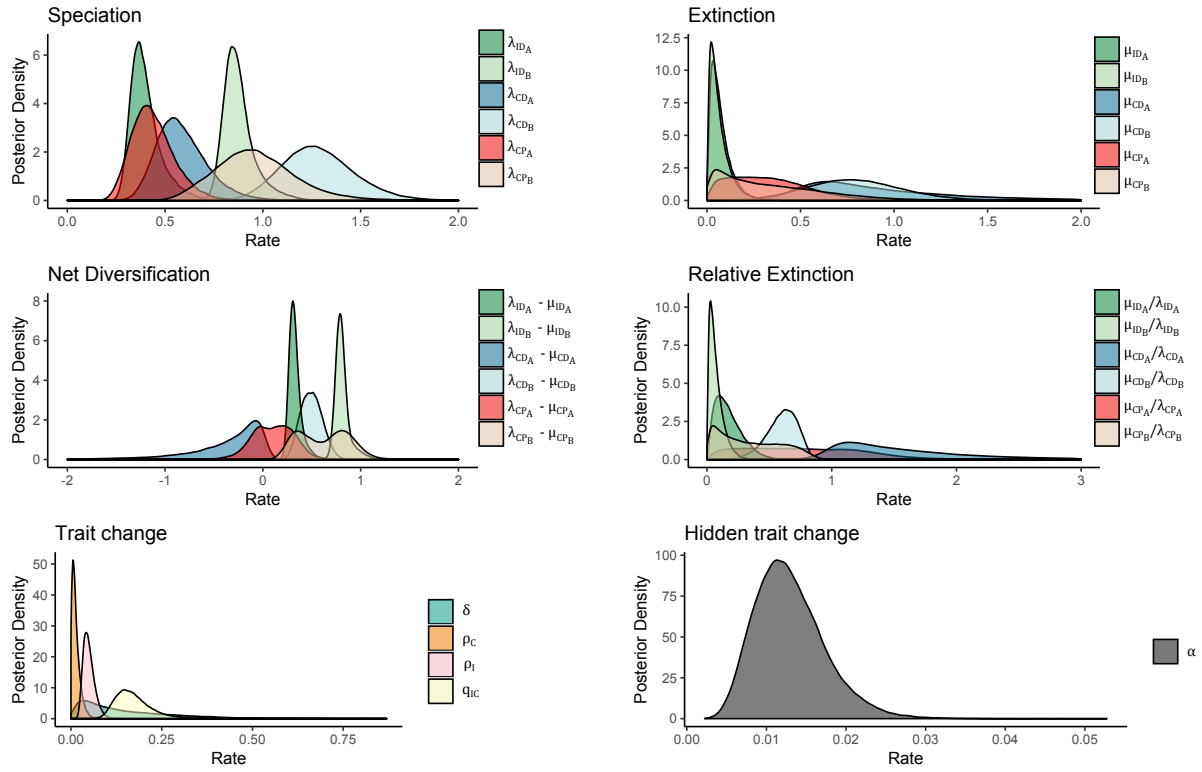


Figure S10: Posterior distribution for each of the parameters in the ID/CD/CP+ $\delta$ +A/B, polyploidy and breeding system model

|          | D/P    | D/P+A/B      | ID/CD/CP | ID/CD/CP+A/B  |
|----------|--------|--------------|----------|---------------|
| $r_D$    | 0.260  | 0.193, 0.658 | —        | —             |
| $r_P$    | -0.056 | 0.030, 0.187 | —        | —             |
| $r_I$    | —      | —            | —        | —             |
| $r_C$    | —      | —            | —        | —             |
| $r_{ID}$ | —      | —            | 0.455    | 0.309, 0.797  |
| $r_{CD}$ | —      | —            | 0.065    | -0.006, 0.587 |
| $r_{CP}$ | —      | —            | -0.088   | -0.074, 0.403 |
| $\rho$   | 0.047  | 0.047        | —        | —             |
| $\rho_I$ | —      | —            | 0.067    | 0.053         |
| $\rho_C$ | —      | —            | 0.033    | 0.032         |
| $q_{IC}$ | —      | —            | 0.198    | 0.145         |

|          | D/P   | D/P+A/B      | I/C    | I/C+A/B       | ID/CD/CP | ID/CD/CP+A/B  |
|----------|-------|--------------|--------|---------------|----------|---------------|
| $r_D$    | 0.382 | 0.698, 0.100 | —      | —             | —        | —             |
| $r_P$    | 0.109 | 0.587, 0.182 | —      | —             | —        | —             |
| $r_I$    | —     | —            | 0.550  | 0.386, 0.877  | —        | —             |
| $r_C$    | —     | —            | -0.001 | -0.059, 0.606 | —        | —             |
| $r_{ID}$ | —     | —            | —      | —             | 0.449    | 0.318, 0.789  |
| $r_{CD}$ | —     | —            | —      | —             | 0.050    | -0.248, 0.494 |
| $r_{CP}$ | —     | —            | —      | —             | -0.027   | 0.110, 0.634  |
| $\rho$   | 0.033 | 0.026        | —      | —             | —        | —             |
| $\rho_I$ | —     | —            | —      | —             | 0.063    | 0.047         |
| $\rho_C$ | —     | —            | —      | —             | 0.024    | 0.011         |
| $\delta$ | 0.050 | 0.162        | —      | —             | 0.022    | 0.107         |
| $q_{IC}$ | —     | —            | 0.364  | —             | 0.194    | 0.164         |

Table S1: Median rate estimates for all fitted models. Units are per million years. Two comma-separated numbers refer to the *A* and *B* hidden states, and — means the parameter was not present in the model. Net diversification rates ( $r$ ) are subscripted with trait state initials (Diploid, Polyploid, Incompatible, Compatible). Transition rates are  $\rho$  (polyploidization), subscripted with background breeding system state;  $\delta$  (diploidization); and  $q_{IC}$  (loss of self-incompatibility). The upper section is for models without diploidization, and the lower section is for models with diploidization. The supplemental figures show the corresponding distributions of parameter estimates.