

## Appendix 1 – Population Genetics and Colony Assignments

### 1 Assessing locus $F_{is}$ , $F_{st}$ and linkage disequilibrium

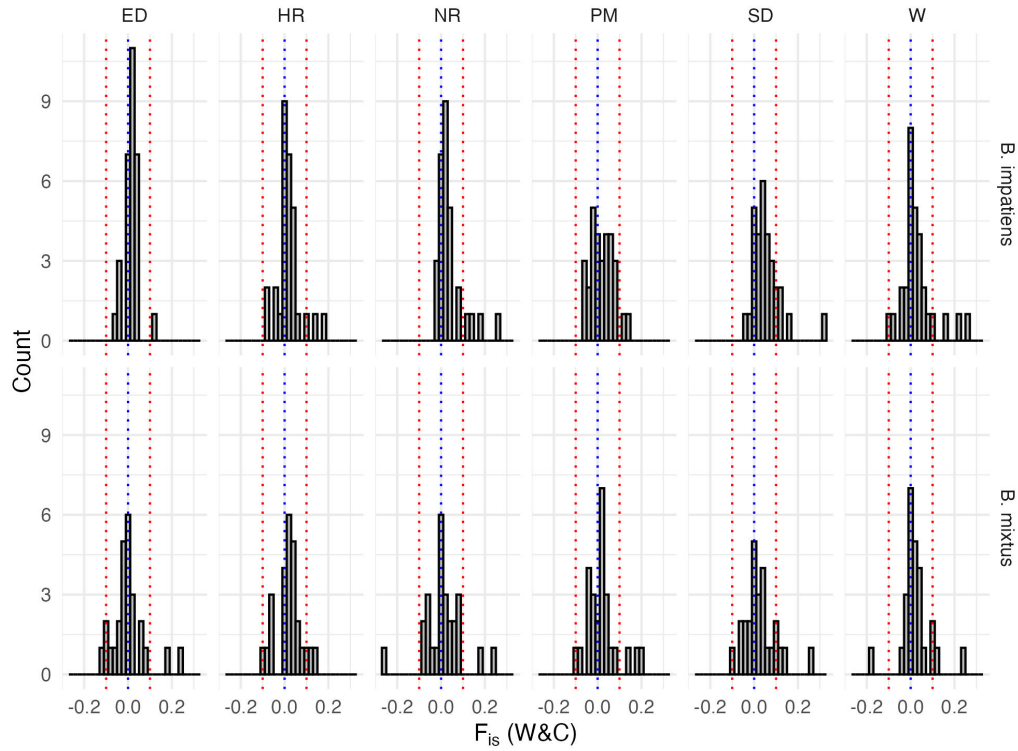


Figure A1: Estimates of  $F_{is}$  for each locus in each subpopulation. Estimates from 2022 and 2023 were calculated separately but are shown together for each site x species combination. Blue dotted lines indicates  $F_{is} = 0$  and red dotted lines indicate  $F_{is} = \pm 0.1$ .

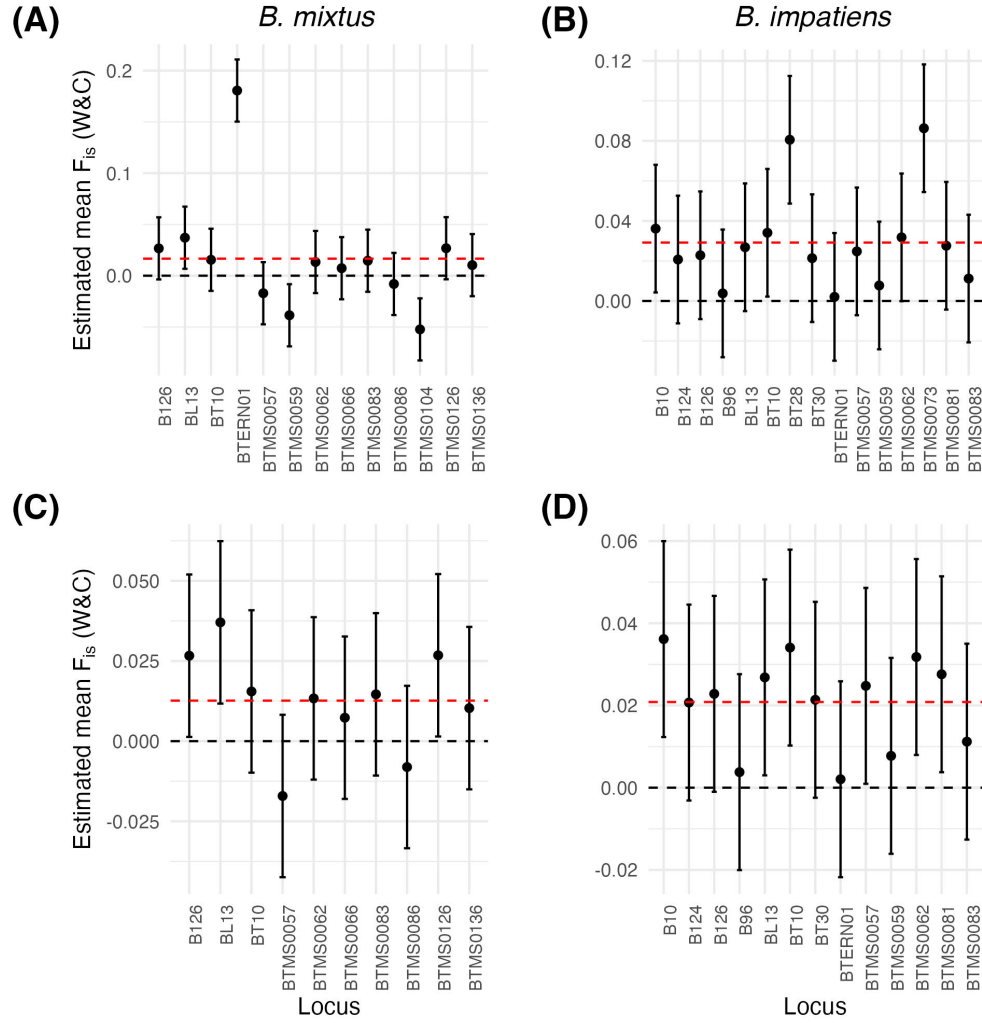


Figure A2: Locus-specific  $F_{is}$  marginal means. A) *B. mixtus* all loci; B) *B. impatiens* all loci; C) *B. mixtus* loci following iterative removal of loci which differed significantly from global mean  $F_{is}$ ; D) *B. impatiens* loci following iterative removal of loci which differed significantly from global mean  $F_{is}$ . Dashed black line denotes  $F_{is} = 0$ , dashed red line denotes global mean  $F_{is}$  for each species.

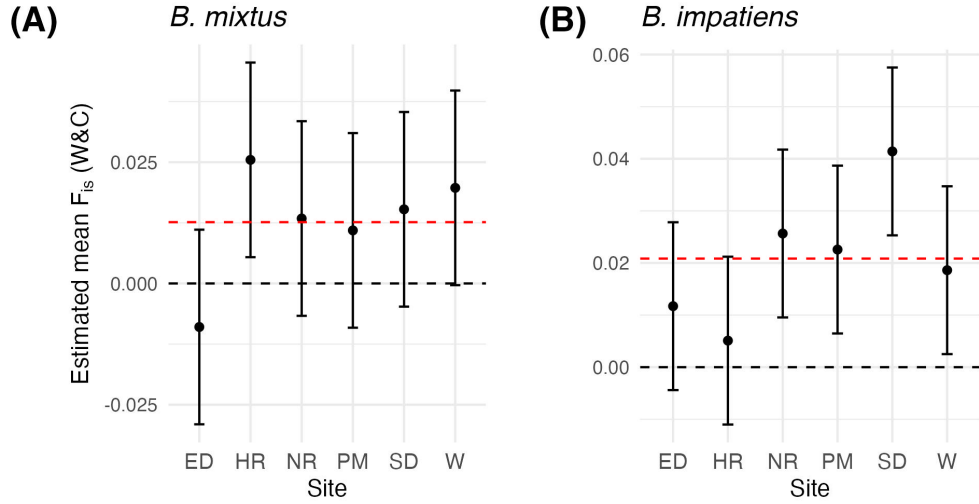


Figure A3: Site-specific  $F_{is}$  marginal means following removal of low-quality loci for A) *B. mixtus* and B) *B. impatiens*. Dashed black line denotes  $F_{is} = 0$ , dashed red line denotes global mean  $F_{is}$  for each species.

## 2 Testing COLONY on simulated data

To test the informativeness of our genetic loci and to validate the accuracy of COLONY2.0 (Jones & Wang, 2010) for accurately detecting siblingships amongst our samples, we performed simulations using realistic family size distributions and the allelic frequencies present in our real data.

We approached this simulations with four objectives:

- (i) To determine false positive and false negative siblingship assignment rates, given the informativeness of our microsatellite datasets,
- (ii) To inform an appropriate strategy (probability threshold, number of runs of the software) for maintaining or rejecting each sib-pair;
- (iii) To select suitable software parameters, and in particular to evaluate the usefulness of siblingship size priors and exclusion of across-site siblingships for reducing false positive rates as sample size increases;
- (iv) To assess whether modelling female polygamy would improve family reconstruction in the case of sibling genotypes simulated under varying rates of multiple paternity.

## 2.1 Simulation strategy

### 2.1.1 Spatially explicit siblingships

We first simulated spatially explicit siblingships following Pope and Jha (2017). We began by simulating six  $5 \times 5$  trapping grids (locations  $k \in \kappa$ ) on a single raster surface comprised of cells  $j \in \mathbb{J}$ . Colonies ( $i \in \mathbb{C}$ ) were distributed uniformly at random throughout the “landscape.”

We sampled individuals from colonies  $i \in \mathbb{C}$  captured at traps  $k \in \mathbb{K}$  from the joint distribution  $\Pr(s, c \mid s \in \kappa)$ , where  $\{s, c\}$  are the indices of a random visitation event of an individual from colony  $c \in \mathbb{C}$  to grid cell  $s \in \mathbb{J}$ .

To do this, we first sampled a trap ( $k$ ) from

$$\Pr(s = k \mid s \in \kappa) = \frac{\Pr(s = k)}{\Pr(s \in \kappa)} \quad (1)$$

where

$$\Pr(s = k) = \sum_{i \in \mathbb{C}} \Pr(s = k \mid c = i) \Pr(c = i)$$

and

$$\Pr(s \in \kappa) = \sum_{i \in \mathbb{C}} \Pr(s \in \kappa \mid c = i) \Pr(c = i) = \sum_{i \in \mathbb{C}} \sum_{k \in \kappa} \Pr(s = k \mid c = i) \Pr(c = i)$$

Combining these statements gives a probability of sampling from trap  $k$  of:

$$\Pr(s = k \mid s \in \kappa) = \frac{\sum_{i \in \mathbb{C}} \Pr(s = k \mid c = i) \Pr(c = i)}{\sum_{i \in \mathbb{C}} \sum_{k \in \kappa} \Pr(s = k \mid c = i) \Pr(c = i)} \quad (2)$$

We then sampled a colony ( $i$ ) from

$$\Pr(c = i \mid s = k) = \frac{\Pr(s = k \mid c = i) \Pr(c = i)}{\Pr(s = k)} = \frac{\Pr(s = k \mid c = i) \Pr(c = i)}{\sum_{i \in \mathcal{C}} \Pr(s = k \mid c = i) \Pr(c = i)} \quad (3)$$

We define the foraging kernel of workers from colony  $i$  as:

$$\Pr(s = k \mid c = i) = \frac{\lambda_i(k)}{\sum_{j \in \mathcal{J}} \lambda_i(j)} \quad (4)$$

The visitation intensity of individuals from colony  $i$  to location  $j$  is defined as:

$$\ln(\lambda_i(j)) = \frac{-\|x_j - \delta_i\|}{\rho} \quad (5)$$

where  $x_j$  are the spatial coordinates of any grid cell in the raster, and  $\delta_i$  are the spatial coordinates of colony  $i$ . The foraging kernel in this example is therefore assumed to be symmetrical and exponentially decaying as a function of distance from the colony location. This means that the total visitation of each colony across the landscape ( $\sum_{j \in \mathcal{J}} \lambda_i(j)$ ) is the same for all colonies, and can be represented using the constant  $\mathbb{D}$ .  $\Pr(c = i)$  is the proportion of all bees in the landscape originating from colony  $i$ , e.g.,  $\Pr(c = i) = \frac{n_i}{N}$  where  $n_i$  is the number of bees from colony  $i$ , and  $N = \sum_{i \in \mathcal{C}} n_i$  is the total number of bees in the landscape.

Combining (4) with (2) and (3) gives the probability of sampling an individual from trap  $k$

$$\Pr(s = k \mid s \in \kappa) = \frac{\sum_{i \in \mathcal{C}} \lambda_i(k) \frac{n_i}{N}}{\sum_{k \in \kappa} \sum_{i \in \mathcal{C}} \lambda_i(k) \frac{n_i}{N}}$$

and the probability that the individual originates from colony  $i$

$$\Pr(c = i \mid s = k) = \frac{\lambda_i(k) \frac{n_i}{N}}{\sum_{i \in \mathcal{C}} \lambda_i(k) \frac{n_i}{N}}$$

For each simulation, samples are drawn from  $\Pr(s, c \mid s \in \kappa)$  until a stopping point (desired number of samples) is reached.  $n_i$  is updated after each "sampling event" to prevent oversampling from colonies located very close to traps.

To verify that the size of sampled sibships (e.g., number of siblings per sibling group) accurately mirrors the distribution of sibship sizes in real data, we compared our simulated distributions to the distribution of sibship sizes in our real data (??). For this simulation strategy, we found that moderating the background density of colonies (i.e., the total number of colonies simulated on the landscape) was the most effective strategy for moderating average sibship size. A larger number of simulated colonies resulted in a higher proportion of singleton colonies (colonies represented by only a single individual).

### 2.1.2 Multilocus genotypes

We simulated multilocus genotypes for each sampled individual under several mating scenarios. In the simplest case, we assume monogamy for both males and queens. The majority of the simulation results presented below follow this assumption. In a second set of simulations, we assumed varying rates of multiple paternity (e.g., queen polygamy) to assess the impact of this assumption on sibship inferences. For each simulation we used the following heuristic:

- (i) Simulate parental genotypes for each sibship based on the allele frequencies present in our real data;
- (ii) Randomly draw offspring genotypes from the set of possible parental alleles at each locus.

We performed simulations based on allele frequencies for both species (*B. mixtus* and *B. impatiens*) because variation in marker number and/or polymorphic information content could lead to differing results. We used inferred allele frequencies from an earlier run of COLONY2.0, which accounts for heightened frequency of alleles present in large families; although raw allele frequencies would have likely been sufficient, given that average family size was small (<2 individuals) and families with > 4 individuals were rare for both datasets.

In the case of monogamous matings, male genotypes were assigned di-

rectly to all offspring in the sibship; in families which were assigned multiple paternity, we assumed two fathers and drew alleles from  $\text{Pr}(father_1, father_2) = (0.7, 0.3)$  following the proportions observed for *B. impatiens* in **birdMatingFrequencyEstimation**

After assigning a multilocus genotype to each individual, we introduced errors and data missingness based on observed error and missingness rates in our real datasets. To introduce errors, we mutated each allele with a probability equal to the rate of errors for that locus and species; we assumed that most errors would be due to contamination, rather than allele dropout, and drew new (erroneous) alleles from the same allele frequencies described above. In the case of data missingness, we observed that individuals which were missing data for *one* copy of a locus were more likely to be missing data for *both* copies than if missingness were distributed uniformly at random. This is likely because there were two primary missingness-generating processes in real data: amplification failure (both alleles missing for an individual) and binning failure (one or both alleles missing for an individual). (In cases where only one copy of a locus failed to amplify, heterozygous individuals would be falsely classified as homozygous—an error, rather than missing data). To account for the structure of missingness, we first calculated the proportion of missing data for each marker ( $P_{missing}$ ) and then removed data for (i) both alleles of each individual with probability  $1/3 * P_{missing}$  and for (ii) a single allele per individual with probability  $1/3 * P_{missing}$ .

## 2.2 Determining an appropriate heuristic for maintaining or rejecting inferred sibships

It has been previously noted (or speculated) in the literature that COLONY is prone to inferring sibships between non-siblings, referred to hereafter as *false positive sibships*. Our own preliminary data analyses suggested that this was the case (e.g., a high number of inferred sibships between individuals separated by >20 km, when individuals from all study sites were permitted to form sibships). While the biology of bumblebees does not unilaterally exclude the possibility of such distant relationships, the likelihood of observing such separation distances is extremely small (see discussion below, section "Observing colony mates at multiple sites").

A commonly used strategy to deal with false positives is to repeat multiple "runs" (usually 2-5) of the COLONY software on the same dataset, and

maintain family groups which are inferred in all runs at or above some confidence threshold (usually  $P \geq 0.95$ , but sometimes  $P \geq 0.8$ ). See, for example, **carvellMolecularSpatialAnalyses2012; raoBumbleBeeHymenoptera2012; dreierFinescaleSpatialGenetic2014a; geibBumbleBeeNest2015a; carvellBumblebeeFamilyLin molaWildfireRevealsTransient2020a**<empty citation>. However, we are not aware of any studies which give support for a particular threshold probability or number of runs necessary to reach a particular confidence level in assignments, or to achieve a satisfactory balance between false positive sibships and missed (false negative) sibships. Indeed, the desirable threshold is likely to vary as a function of the number and informativeness of markers for a given population.

To overcome these limitations, we tested probability exclusion criteria from  $P = 0.95$  to  $P = 1$ , for 1 or 5 runs of COLONY version 2.0.6.5. Further, we compared the use of family cluster probabilities (COLONY output file .BestCluster—hereafter referred to as the family method) to full sibling dyad probabilities (COLONY output .FullSibDyad—hereafter referred to as the dyad method).

We began by simulating 5 datasets (e.g., different sibship arrangements with unique parental genotypes) consisting of  $n = 1200$  individuals each, which was roughly the midpoint of population sizes for our real data. For each dataset we performed 5 runs of COLONY (see A1 for a summary of COLONY software settings).

We found that the family method was less reliable than dyad method for excluding erroneous sibships; while false negative rates were similar for both strategies, 17-36% of all inferred pairwise relationships were false positives when using the family method, even



Table A1: Description of software settings (COLONY 2.0.6.5 (Jones & Wang, 2010)) for simulations.

Simulation	Comparison	Number of Individuals	Siblingship Size Prior	Cross-site Exclusion	Number of Runs
Subsection 1	Number of COLONY runs	1200	yes	yes	1-5
Subsection 1	Probability threshold	1200	yes	yes	1-5
Subsection 1	Families vs dyads	1200	yes	yes	1-5
Subsection 2	Siblingship size prior	400-2000	yes/no	yes/no	1
Subsection 2	Cross-site exclusion	400-2000	yes/no	yes/no	1
Subsection 3	Mating system	1000	no	no	1
Subsection 3	Mating system (augmented data)	1000	yes	no	1

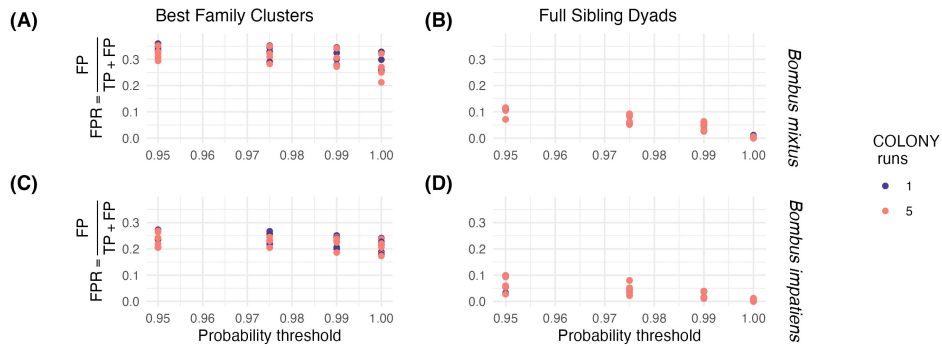


Figure A4: caption

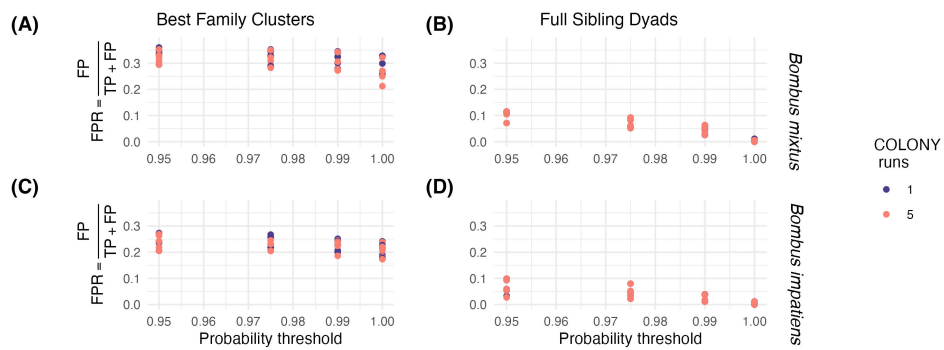


Figure A5: caption

- 2.3 Assessing the use of sibblingship size priors and cross-site sibling exclusion for reducing false positive rates
- 2.4 Evaluating the effects of multiple paternity on sibblingship inference
- 3 Observing colonymates at multiple sites