

Construction of coalescent trees on partially fixed pedigrees

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7 August 2018

Genealogies

- Φ : Phenotypes
 A : Affected, **N** : Not Affected
- \mathcal{H} : Haplotypes
 ■ : Mutant, □ : Non-Mutant
- θ : Unknown parameter

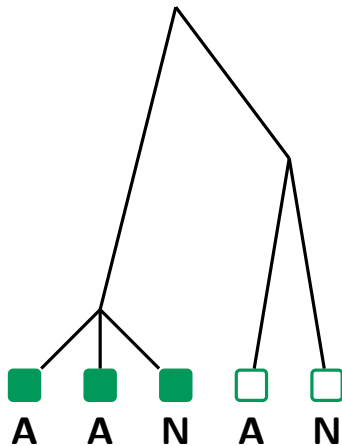
$$\mathcal{L}(\theta) = P(\Phi, \mathcal{H} | \theta).$$



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- \mathcal{G} : Genealogy

$$\begin{aligned}\mathcal{L}(\theta) &= P(\Phi, \mathcal{H} | \theta) \\ &= \int_{\mathcal{G}} P(\Phi, \mathcal{H} | \mathcal{G}, \theta) P(\mathcal{G} | \theta) d\mathcal{G}.\end{aligned}$$



Diploid, Two-Sex, Wright-Fisher Model

Population size $2N$; N males ; N females ; $4N$ genes.



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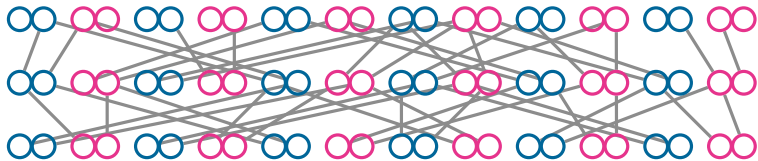
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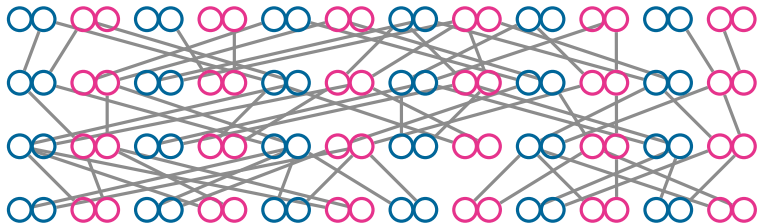
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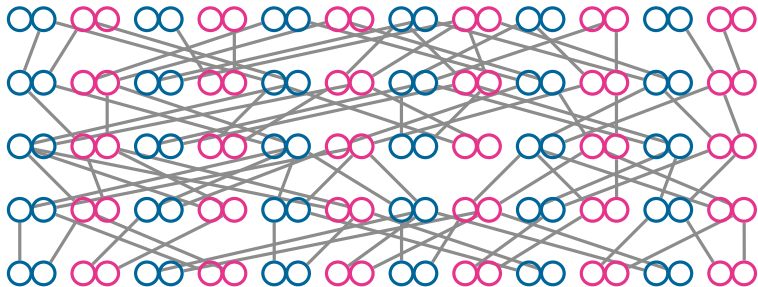
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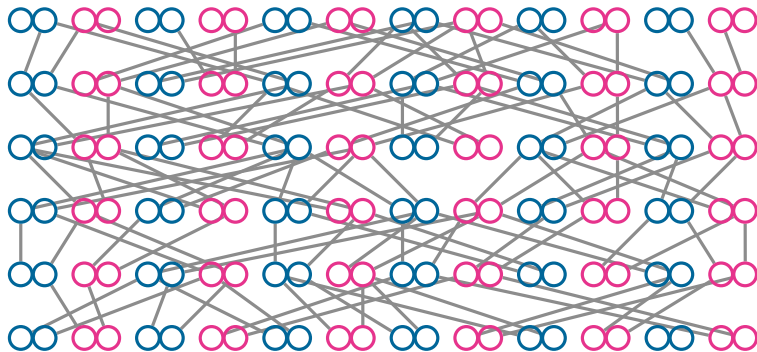
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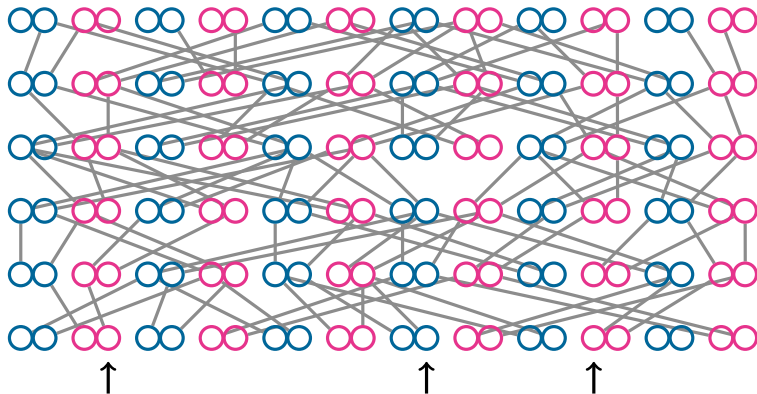
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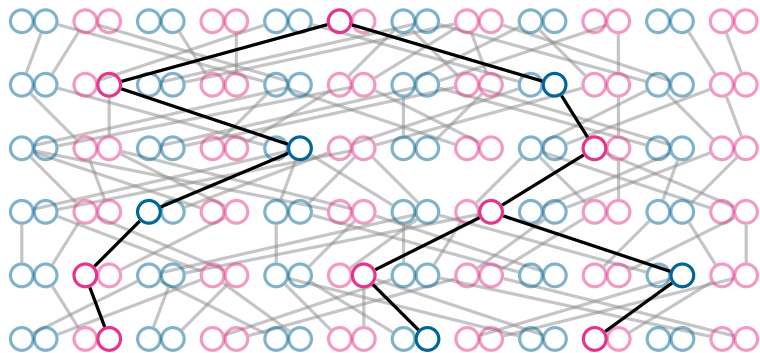
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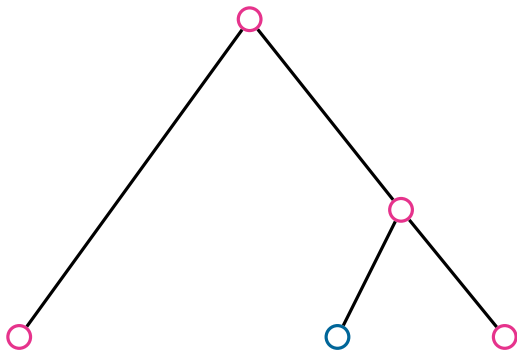
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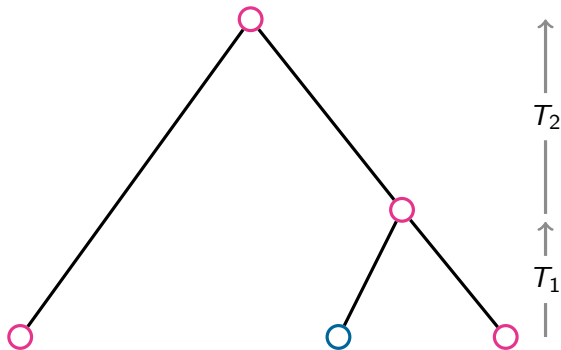
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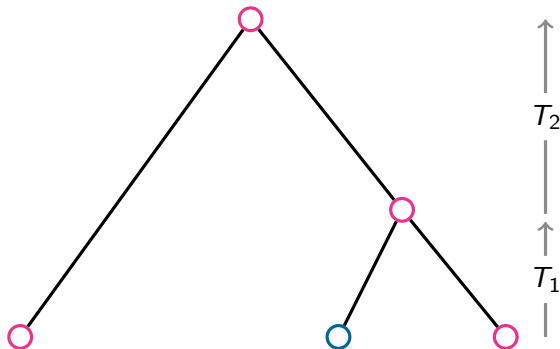


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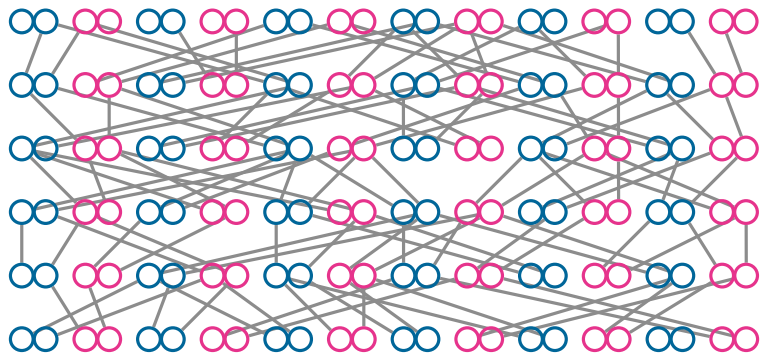
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Convergence to the n -coalescent (Möhle [1998])

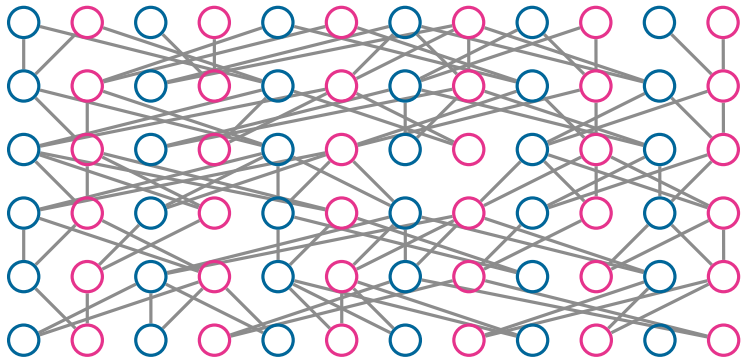
$$T_1/4N \sim \text{Exp}(3), T_2/4N \sim \text{Exp}(1); \quad N \text{ large.}$$



Population Pedigree



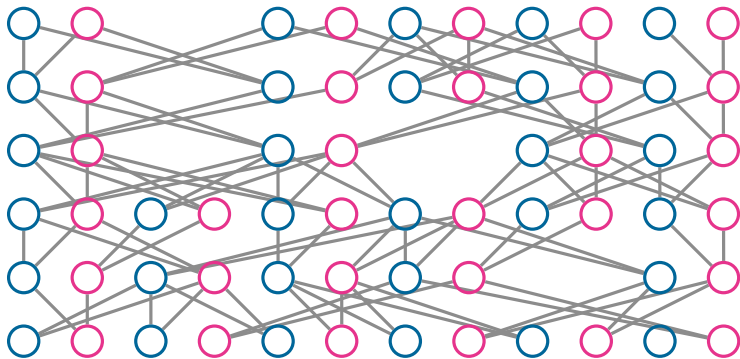
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Population Pedigree (Wakeley et al. [2012])

The set of all family relationships among members of the population for every generation.

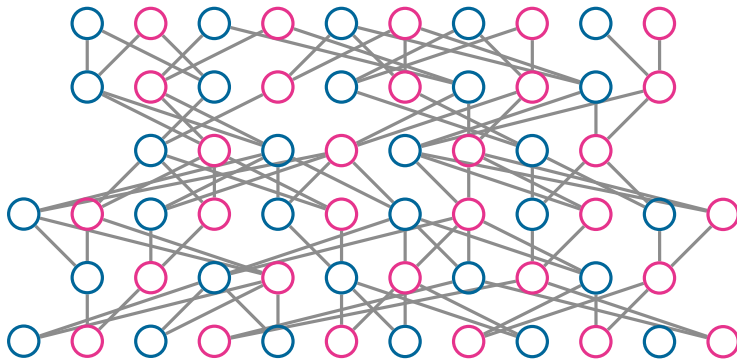
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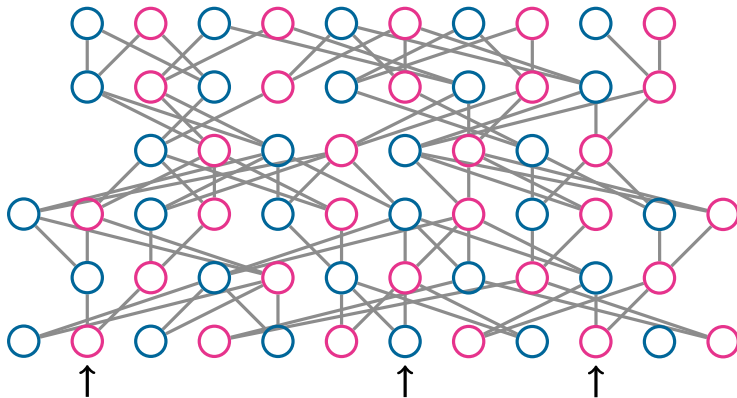
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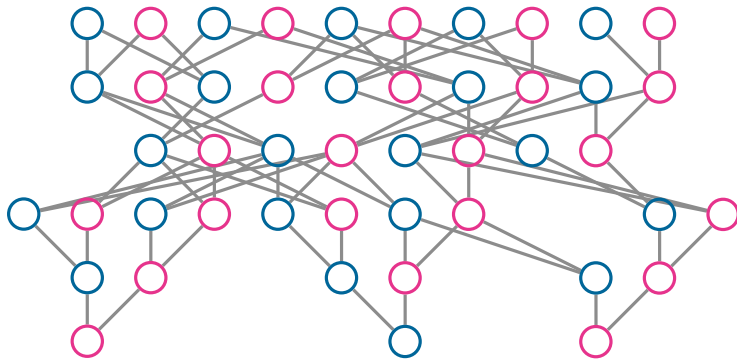
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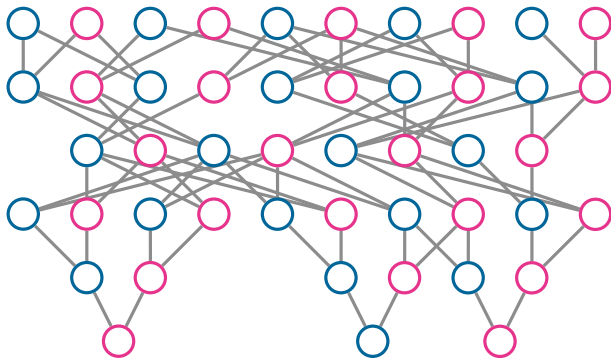
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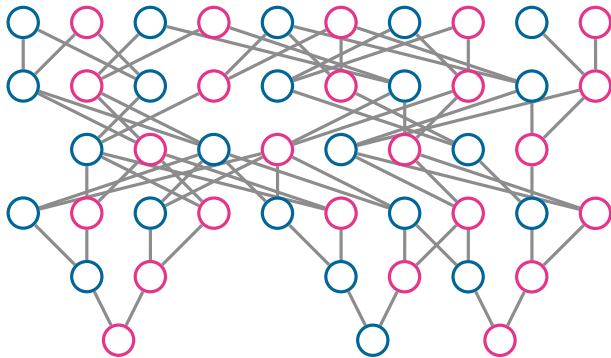
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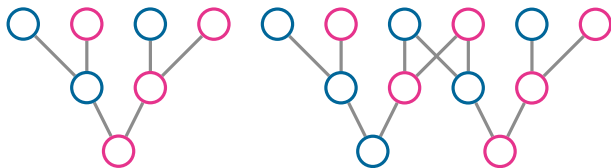


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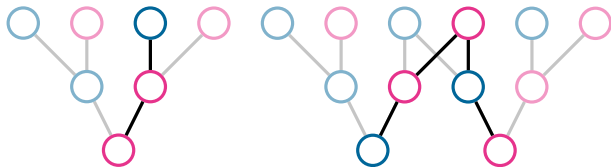


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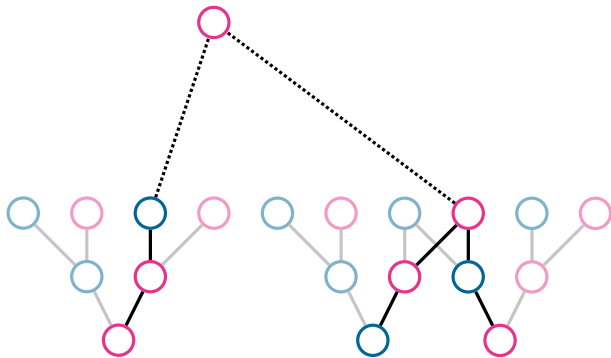


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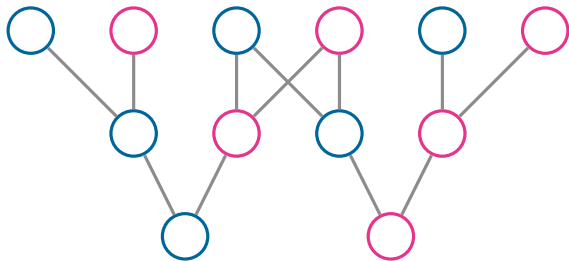


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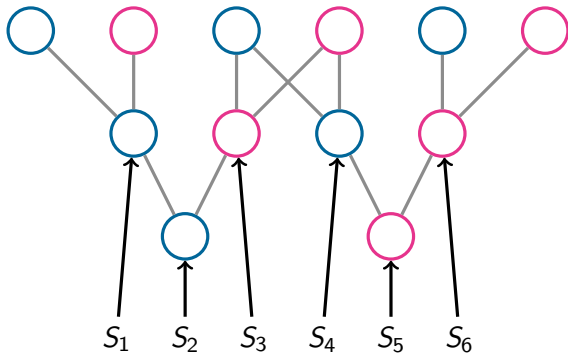
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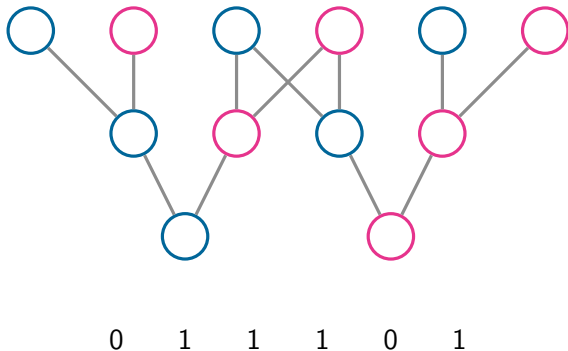


Meiosis Indicator (Thompson [2000])

$S_i \sim i.i.d. \text{Bernoulli}(0.5).$

Mendel's 1st Law

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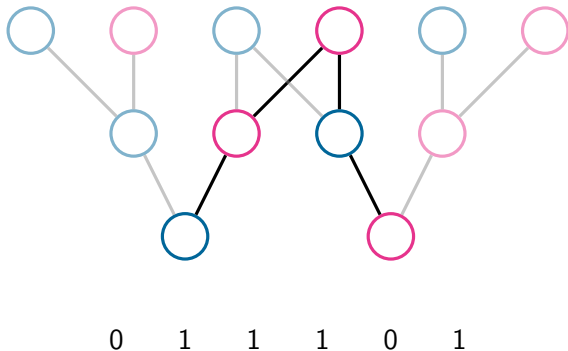


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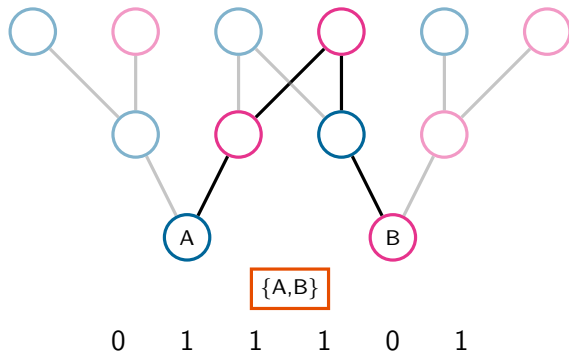


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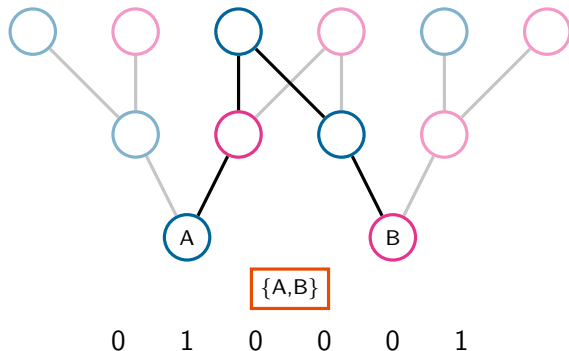
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IBD Partition

Partition of genes into subsets that are IBD

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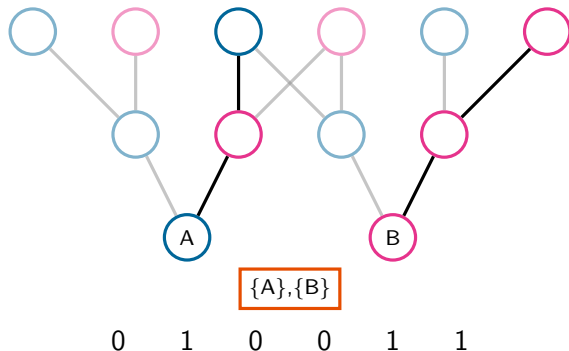
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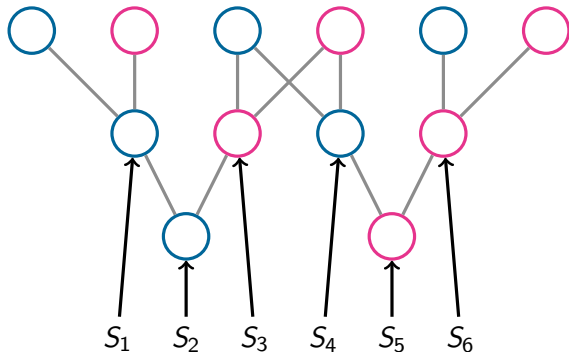
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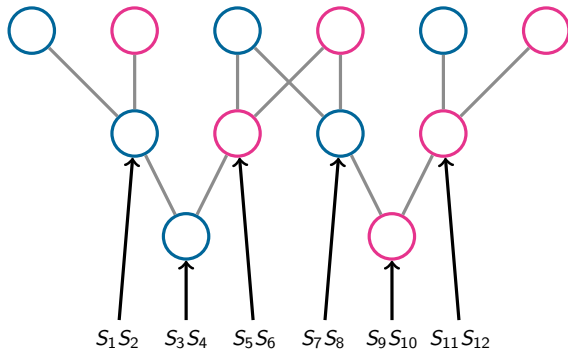
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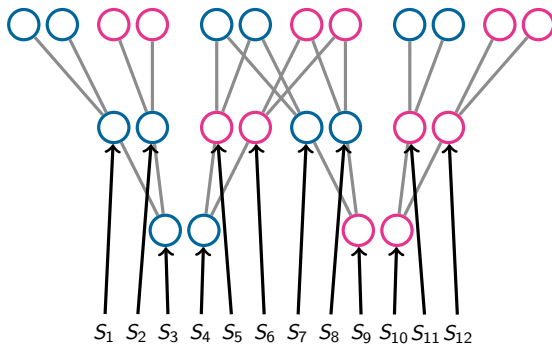
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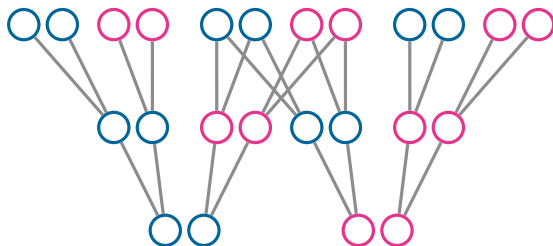
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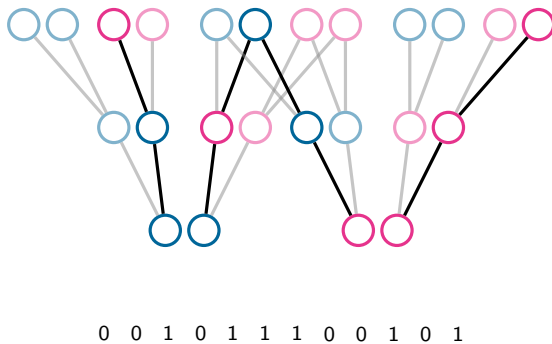
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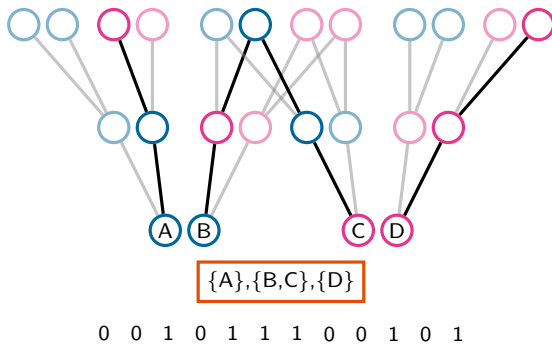
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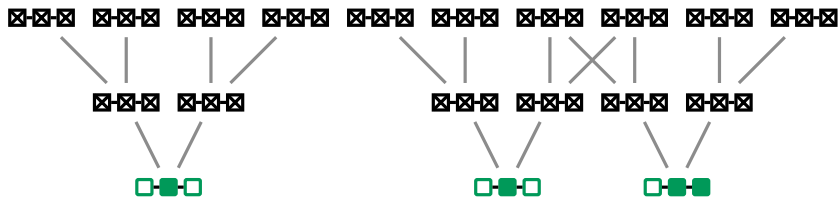
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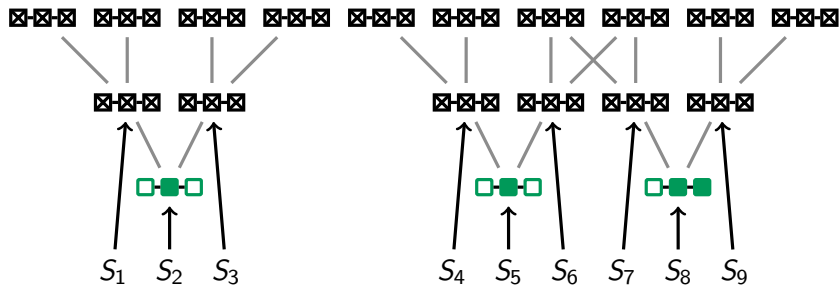
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SNP Data



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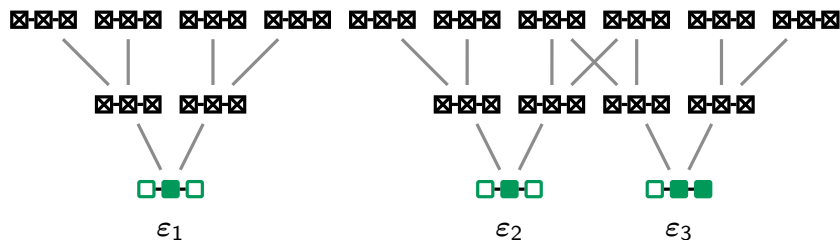
k -locus Meiosis Indicator

$$S_i \in \{0, 1\}^k.$$

e.g. No interference (Allen and Darwiche [2008]) :

$$P(S_i = 011) = P(S_i = 100) = 0.5p_{12}(1 - p_{23}).$$

SNP Data

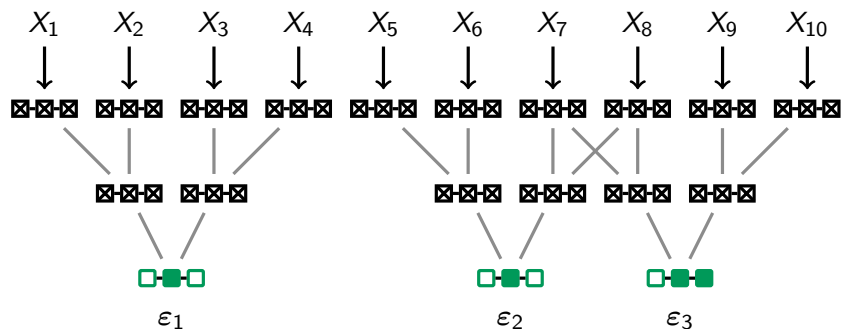


Evidence : ϵ (Koller and Friedman [2009])

An instantiation of a subset of random variables.

$$P(S_1, S_2, \dots, S_n | \epsilon) \neq \prod_{i=1}^n P(S_i | \epsilon).$$

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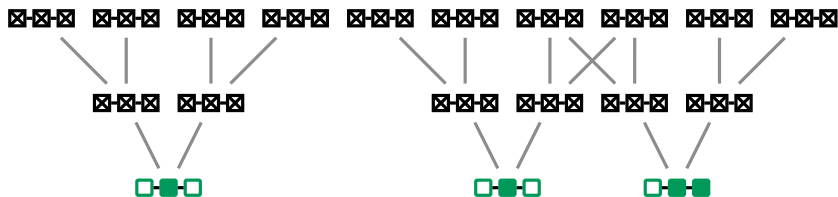


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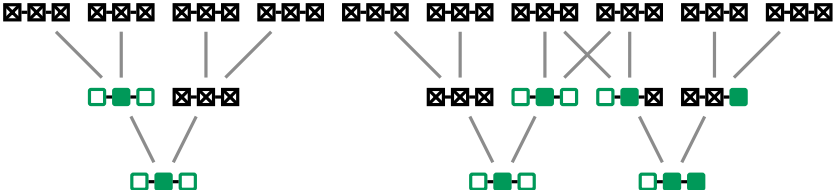
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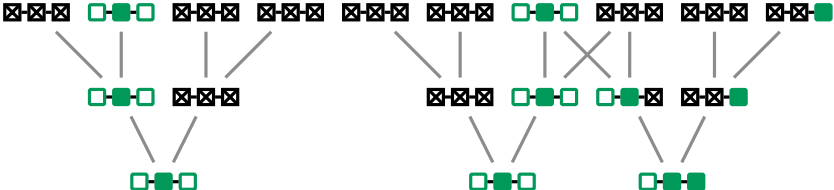
P-Coalescent



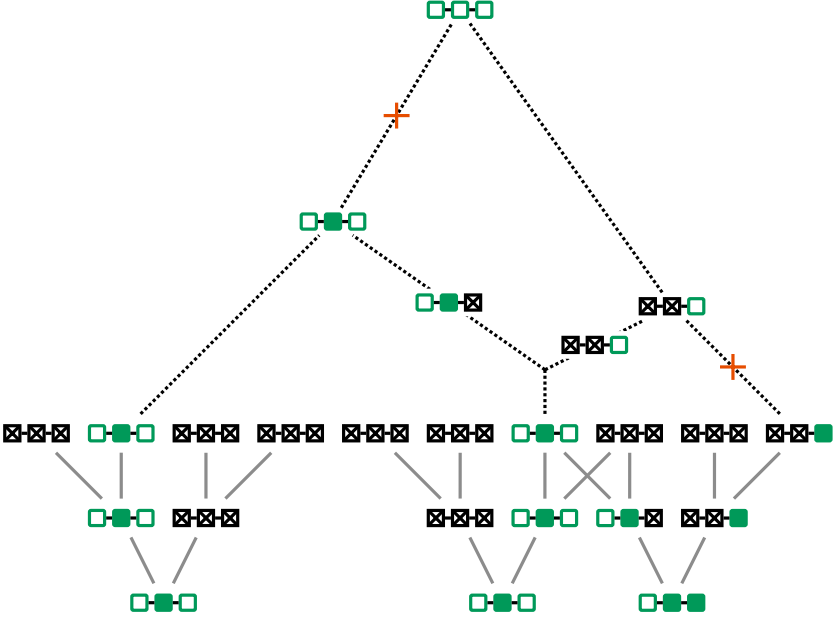
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Perspectives

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Big Pedigree \Leftrightarrow High Complexity.

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Founder Prior

- Uniform prior
- Linkage equilibrium
- Linkage disequilibrium

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