# Introduction to Bioinformatics for Computer Scientists 

## Lecture 10

## Outline

- Last Time
- Bayesian statistics
- Monte-Carlo simulation \& integration
- Markov-Chain Monte-Carlo methods
- Metropolis-coupled MCMC-methods


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- Last Time
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- Today
- Bayesian Inference
- The curse of priors
- Some phylogenetic proposals
- Reversible jump MCMC
- Introduction to Population Genetics


## Bayes Theorem



## A few words about priors

- Prior probabilities convey the scientist's beliefs, before having seen the data
- Using uninformative prior probability distributions (e.g., uniform priors, also called flat priors)
$\rightarrow$ differences between prior and posterior distribution are attributable to likelihood differences only
- Priors can bias an analysis !
- For instance, we could chose an arbitrary prior distribution for branch lengths in the range [1.0,20.0]
$\rightarrow$ what happens if branch lengths are much shorter?


## An Analysis from Last Week

- We analyzed a couple of natural language datasets
- Under ML we found a very weird bi-modal distribution of the $\alpha$ shape parameter for the $\Gamma$ model of rate heterogeneity

| dataset | cognate classes | sound correspondences | combined |
| :--- | :---: | :---: | :---: |
| constenlachibchan | 0.592 | $\mathbf{9 9 . 8 7 1}$ | 4.178 |
| crossandean | 1.243 | 6.334 | 1.154 |
| dravlex | 0.702 | 4.301 | 2.234 |
| felekesemitic | 1.062 | 7.430 | 2.693 |
| hattorijaponic | $\mathbf{9 9 . 8 4 8}$ | $\mathbf{9 9 . 8 9 7}$ | $\mathbf{9 9 . 8 9 0}$ |
| houchinese | 2.357 | 6.120 | 4.195 |
| leekoreanic | 8.316 | 8.420 | 3.284 |
| robinsonap | $\mathbf{9 9 . 8 6 9}$ | 15.269 | 3.486 |
| walworthpolynesian | 1.333 | 4.233 | 1.624 |
| zhivlovobugrian | $\mathbf{9 9 . 8 5 0}$ | 4.244 | 3.134 |

Table 1: Alphas, Values indicating a very low rate heterogeneity are highlighted in bold

## What did the Bayesian Inference yield?

- With the default prior used for molecular Sequence Datasets



## Empirical Distribution of $\alpha$

- For tens of thousands of empirical molecular datasets this is how the ML estimate of $\alpha$ is distributed



## When specifying a uniform prior



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## Some Phylogenetic Proposal Mechanisms

- Branch Lengths
- Sliding Window Proposal
- Multiplier Proposal
- Topologies
- Local Proposal (the one with the bug in the Hastings ratio)
- Extending TBR (Tree Bisection Reconnection) Proposal
- Remember: We need to design proposals for which
- We either don't need to calculate the Hastings ratio
- Or for which we can calculate it
- That have a 'good' acceptance rate
$\rightarrow$ all sorts of tricks being used, e.g., parsimony-biased topological proposals


## Some Phylogenetic Proposal Mechanisms

- Univariate parameters \& branch lengths
- Sliding Window Proposal
- Branch lengths
- Node slider proposal
- Topologies
- Local Proposal (the one with the bug in the Hastings ratio!)
- Remember: We need to design proposals for which
- We either don't need to calculate the Hastings ratio
- Or for which we can calculate it
- That have an appropriate acceptance rate
$\rightarrow$ all sorts of tricks being used, e.g., parsimony-biased topological proposals
$\rightarrow$ acceptance rate should be around $25 \%$ (empirical observation)
$\rightarrow$ for sampling from a multivariate normal distribution it has been formally shown that an acceptance rate of $23.4 \%$ is optimal


## Sliding Window Proposal

Sliding window width $\delta$


Current parameter value
Parameter value range

## Sliding Window Proposal

Sliding window width $\delta$


Propose new value at random
Parameter value range within $\delta$

## Sliding Window Proposal



Allowed parameter value range

## Notes:

1. The hastings ratio of this move is 1
2. The edge cases can be handled by back-projection
3. The window size $\delta$ can be tuned itself (auto-tuning) to obtain an acceptance rate of $\approx 1 / 4$
4. This proposal can be used, e.g., for the $a$-shape parameter of the $\Gamma$ function in rate heterogeneity models

## The Node Slider Proposal

1. Pick 2 contiguous branches at random

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2. Multiply the 2 branches by the same random number

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3. Propose a new branch ratio $b_{1} / b_{2}$ at random

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The Hastings ratio of this move is not 1 !

## Moving through Tree Space

## Moving through Tree Space

1. Pick 3 contiguous branches at random

## Moving through Tree Space

1. Pick 3 contiguous branches at random that define 2
Subtrees $X$ and $Y$

## Moving through Tree Space



1. Pick 3 contiguous branches at random that define 2
Subtrees $X$ and $Y$
2. shrink or grow selected 3 branch segment by a random amount

## Moving through Tree Space



1. Pick 3 contiguous branches at random that define 2
Subtrees $X$ and $Y$
2. shrink or grow selected 3 branch segment by a random Amount
3. Chose either $X$ or $Y$ at random and prune it from the tree

## Moving through Tree Space



1. Pick 3 contiguous branches at random that define 2
Subtrees $X$ and $Y$
2. shrink or grow selected 3 branch segment by a random Amount
3. Chose either $X$ or $Y$ at random And prune it from the tree
4. Re-insert $Y$ at random into

The 3 branch segment

## Moving through Tree Space



Initial tree $\mathrm{t}_{\mathrm{i}}$


Proposed tree $\mathrm{t}_{\mathrm{i}+1}$

Proposed tree: 3 branch lengths changed and one NNI (Nearest Neighbor Interchange) move applied

## Moving through Tree Space



Initial tree $\mathrm{t}_{\mathrm{i}}$
LnL = -2900


Proposed tree $t_{i+1}$

The proposed tree has a better likelihood! Will the proposed tree always be accepted?

## Moving through Tree Space



Initial tree $\mathrm{t}_{\mathrm{i}}$

LnL $=-2900$


Proposed tree $t_{i+1}$

The proposed tree has a better likelihood! Will the proposed tree always be accepted? $\rightarrow$ think about Priors and Hastings ratio!

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## How do we select models using MCMC?

- Example: Consider all possible time-reversible nucleotide substitution models ranging from Jukes Cantor (JC, 1 rate) to the General Time Reversible Model (GTR, 6 rates)
- We will denote rate configurations by strings, e.g.,
- 111111 is the JC model
- ...
- 123456 is the GTR model
- Let me explain this further ...


## Model Strings

111111

## Model Strings

111111
$\left.\left.\begin{array}{c} \\ A \\ \mathrm{~A} \\ \mathrm{G} \\ \mathrm{T}\end{array}\right] \begin{array}{cccc}\mathrm{A} & \mathrm{C} & \mathrm{G} & \mathrm{T} \\ * & \lambda & \lambda & \lambda \\ & * & \lambda & \lambda \\ & & * & \lambda \\ & & & *\end{array}\right)$

## Model Strings

112211

|  | A | C | G | T |
| :---: | :---: | :---: | :---: | :---: |
| A | * | $\lambda$ | $\lambda$ | V |
| - C |  | * | V | $\lambda$ |
| G |  |  | * | $\lambda$ |
| T |  |  |  | * |

## Model Strings

112121
$\left.\left.\begin{array}{c} \\ A \\ \mathrm{C} \\ \mathrm{G} \\ \mathrm{T}\end{array}\right] \begin{array}{cccc}\mathrm{A} & \mathrm{C} & \mathrm{G} & \mathrm{T} \\ * & \lambda & \lambda & V \\ & * & \lambda & V \\ & & * & \lambda \\ & & & *\end{array}\right)$

## Model Strings



## How many time-reversible DNA models are there?

- Number of ways a set with $n$ objects can be partitioned into disjoint nonempty sets
- Example: the set $\{a, b, c\}$ can be partitioned as follows:
\{ \{a\}, \{b\}, \{c\} \}
$\{\{a\},\{b, c\}\}$
\{ \{b\}, $\{a, c\}\}$
$\{\{c\},\{a, b\}\}$
$\{\{a, b, c\}\}$
- The number of combinations for $n$ (3 in our example) is given by the socalled Bell number, for details see https://en.wikipedia.org/wiki/Bell_number


## The Bell Numbers

- $n:=1 \rightarrow 1$
- $n:=2 \rightarrow 2$
- $n:=3 \rightarrow 5$
- $n:=4 \rightarrow 15$
- $n:=5 \rightarrow 52$
- $n:=6 \rightarrow 203$
- $n:=7 \rightarrow 877$
- etc...


## What do we need?

- Apart from our usual suspect parameters (tree topology, branch lengths, stationary frequencies, substitution rates, a), we also want to integrate over different models now ...
- What are the problems we need to solve?


## What do we need?

- Apart from our usual suspect parameters (tree topology, branch lengths, stationary frequencies, substitution rates, $\alpha$ ), we also want to integrate over different models now ...
- What are the problems we need to solve?
- Problem \#1: we need to design proposals for moving between different models
- Problem \#2: those models have different numbers of parameters, we can not directly compare likelihoods
- Here we use MCMC to not only sample model parameters, but also models


## Problem \#1 Model Proposals

- Any ideas?


## Problem \#1 Model Proposals

- Split move

Chose a set of substitution rates with > 1 member at random
111222 (two-parameter model)
and split it randomly into two rates
111223 (three-parameter model)

- Merge move

Chose two substitution rate sets at random
111223
and merge them into one substitution rate set
111222

## Problem \#1 Model Proposals

- Split move

Chose a set of substitution rates with > 1 member at random
111222 (two-parameter model)
and split it randomly in
Clear to everyone what the
111223 (three-param respective rate matrix looks like?

- Merge move

Chose two substit an rate sets at random
111223
and merge them into one substitution rate set
111222

## Problem \#2 Sampling Different Models

- Use reversible jump MCMC (rjMCMC) to jump between models (posterior probability distributions) with different number of parameters (posterior distributions with different dimensions)
- The model proposal moves we designed are reversible jump moves!
- Evidently, we need to somehow modify our proposal ratio calculation ...
- In general terms, the acceptance ratio is calculated as:
$r=$ likelihood ratio * prior ratio * proposal ratio * Jacobian
A Jacobian defines
a linear map from $R^{n} \overrightarrow{R^{m}}$
at point $x$, if function $f(x)$
is differentiable at $x$


## Problem \#2 Sampling Different Models

- Use reversible jump MCMC (rjMCMC) to jump between models (posterior probability distributions) with different number of parameters (posterior distributions with different dimensions)
- The model proposal moves we designed are reversible jump moves!
- Evidently, we need to somehow modify our proposal ratio calculation ...
- In general terms, the acceptance ratio is calculated as:
$r=$ likelihood ratio * prior ratio * proposal ratio * Jacobian
I will not provide further
Details; see work by Peter Green
$(1995,2003)$ who developed the rjMCMC methods


## rjMCMC - summary

- Need to design moves that can jump back and forth between models of different dimensions (parameter counts)
- Need to extend acceptance ratio calculation to account for jumps between different models
- The posterior probability of a specific model (e.g., JC or GTR) is calculated as the fraction of time (fraction of samples) the MCMC chain visited/spent time/generations sampling within that model ...


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## Outline: Population Genetics

- What is biological evolution?
- Units \& Types of Evolution
- Good old G. Mendel (phenotypes)
- Alleles \& SNPs (genotypes)
- Models of evolution for infinite populations (Hardy)
- Models of evolution for finite populations (Wright-Fisher)


## What is Evolution?

- Change over time
- Languages evolve $\rightarrow$ languages change
- Galaxies evolve $\rightarrow$ galaxies change
- Political systems change $\rightarrow$ political systems evolve


## Biological Evolution

- In Biology one more condition, except for change, is required to characterize evolution
- Do you know which one?


## Biological Evolution



Is this evolution?

## Biological Evolution



Time
population individuals

## Biological Evolution



The frequency of black and white individuals in the population changes over time.
Is this evolution?

## Biological Evolution

Generation 0
Generation 1
Generation 2


The frequency of black
Is this evolution?


## Biological Evolution

Generation 0
Generation 1
Generation 2


Time

In population genetics, we are interested in how characteristics (e.g., ratio of black versus white individuals) of populations change over time.

## Another Example

- Population of 5 white and 5 black individuals
- frequency $(w h i t e)=0.5$
- frequency(black) $=0.5$
- Suddenly 7 out of 10 individuals die $\rightarrow 2$ white and 1 black left
- frequency(white) $=2 / 3$
- frequency(black) $=1 / 3$
- The population has changed!
- Is this evolution?


## Yet Another Example

- Population of 5 white and 5 black individuals
- frequency $(w h i t e)=0.5$
- frequency(black) $=0.5$
- 3 individuals (2 white \& 1 black) decide to leave and form a new colony
- frequency(white) $=2 / 3$
- frequency(black) $=1 / 3$
- The population of the new colony is different!
- Is this evolution?


## Biological Evolution

- The phenomenon of change is not sufficient for defining biological change/evolution
- For talking about biological evolution, change needs to be inherited
- The reasons for the change are not important for the definition of biological evolution
- ... but we are of course interested in them!


## Biological Evolution

- Given these examples, by biological evolution we refer to
- Change of the frequency of occurrence of features of individuals in the population
- Features can be, for instance, resistance to antibiotics, color, etc.
- These features should be inherited from generation to generation
- Key question: What are the mechanisms of feature inheritance?
- We distinguish between phenotype and genotype!


## The basic Unit of Biological Evolution

- Based on the previous examples, what is the biological unit of evolution:
- An individual?
- A population?
- Something else?


## Units of Evolution

- The population
- A gene
- The genome of an individual
- One needs to define first at which level evolutionary forces act
$\rightarrow$ what competes with what?


## Units of Evolution: The Population

- A Population evolves because the frequency of the features of its individuals changes
- Features frequency can change due to

1. Genetic Drift: Chance (other than a random mutation)
2. Migration
3. Mutation
4. Natural Selection: Response to some pressure (e.g., antibiotics, climate change)

- Features can be:
- Genotype
- Phenotype


## Genetic Drift



Composition of population changes by some random event

## Migration



## Mutation



A random mutation may occur that changes the color of the offspring and hence the frequency of brown beetles in the population

## Natural Selection



Green Beetles may be easier to spot for birds $\rightarrow$ they will have less offsprings in the following generations

Time

## Units of Evolution: The Gene

- Genes encode information
- Assume that gene A encodes eye color
- In reality a total of about 15 genes encode eye color
- If $\mathbf{A}$ has the form $\mathrm{A} \rightarrow$ color $=$ blue
- If $\mathbf{A}$ has the form $\mathrm{a} \rightarrow$ color $=$ brown
- What does form mean?


## Units of Evolution: The Gene

- Genes are inherited from generation $\rightarrow$ generation
- Inheritance take places via Alleles
- An Allele is a specific form (slightly different DNA sequence): a or A of gene A
- Most multi-cellular organisms are diploid $\rightarrow$ they have two sets of corresponding chromosomes that are called homologous
- Diploid organisms have one copy of each gene/allele in each of the homologous chromosome pairs
- If the Allele sequences in the two chromosomes are identical: homozygous
- If the Allele sequences in the two chromosomes are different:
heterozygous


## Units of Evolution: The Gene

Diploid Chromosome



Homologous pair of chromosomes

## Units of Evolution: The Gene



Homozygous Allele $\rightarrow$ identical DNA


Heterozygous Allele $\rightarrow$ different DNA sequence

## Fraction of heterozygous Alleles

|  |  | Individual |
| :--- | :--- | :---: |
| Ancient DNA Siberia | Heterozygosity estimate (\%) |  |
| http://en.wikipedia.org/wiki/Denisova_Cave | San | 0.0165 |
|  | Mandenka | 0.0721 |
|  | Yoruba | 0.0686 |
|  | Mbuti | 0.0649 |
|  | Dinka | 0.0657 |
|  | Sardinian | 0.0635 |
|  | French | 0.0490 |
|  | Dai | 0.0473 |
|  | Han | 0.0465 |
|  | Papuan | 0.0454 |
|  | Karitiana | 0.0386 |
|  |  | 0.0353 |

Table from:
http://genetics.med.harvard.edu/reich/Reich_Lab/Welcome_files/2013_Bryc_Genetics.pdf

## Units of Evolution: The Gene

- Why are we interested in heterozygous versus homozygous Alleles?
- Inheritance $\rightarrow$ Humans inherit one allele from the father and one from the mother
- Some more terminology:
- Genotype of a gene: the set of corresponding alleles in a diploid organism
- Phenotype of a gene: observation for the trait/property that the gene controls (e.g. brown eye color) $\rightarrow$ in reality more complex genes interact on traits


## Mendelian Inheritance



Pea plant traits (phenotype!) studied by G. Mendel

## Dominance

- In Mendel's experiment
- An individual with the Round-Wrinkled genotype had the Round phenotype, i.e., $R W \rightarrow R$
- We say that the round allele is dominant and the wrinkled allele is recessive
- What are the phenotypes of:
- RR $\rightarrow$ ?
- RW $\rightarrow$ ?
- WR $\rightarrow$ ?
- WW $\rightarrow$ ?
- If there is no dominance-recession relationship the phenotype is intermediate!


## Mendel

Homozygous round seed: RR

cross

## Mendel

Homozygous round seed: RR


Generation 1

## Mendel

Homozygous round seed: RR

## Generation 1


selffertilize
cross

Homozygous wrinkled seed: WW
$>$

## Mendel

Homozygous round seed: RR

Homozygous wrinkled seed: WW


Generation 1

What do you expect?

## Mendel

Homozygous round seed: RR

Generation 1
selffertilize Homozygous wrinkled seed: WW


Generation 2
5474


1850

## Mendel

Homozygous round seed: RR


## Mendel's $1^{\text {st }}$ law The principle of Segregation

Each physical trait of a diploid organism is determined by two factors (alleles). These two factors separate between the generations and re-unite in the next generation.

- Observation: the $2^{\text {nd }}$ generation shows all traits from the initial generation 0 even though the parents in generation 1 do not show all traits.
- Conclusion: Generation 1 must receive some information that causes this "hidden" trait to be revealed in generation 2, in addition to the traits of generation 1 .


## Allele Inheritance

- As we know, a diploid organism has 2 alleles per gene
- Alleles can either be heterozygous or homozygous
- One allele is inherited from the mother and one from the father
$\rightarrow$ each parent will pass only one of his - possibly heterozygous
- alleles to the offspring
- For a certain, single allele, there is a $50 \%$ chance to have obtained it either from the mother or from the father


## Allele Inheritance Terminology

- We denote a gene with the capital bold-font letter $\mathbf{A}$
- We denote corresponding Alleles by $A$ and $a$ if two alleles exist or as $A_{1}, A_{2}, A_{3}, \ldots$ if more than two alleles exist
- A denotes both, an allele, and the corresponding gene which may sometimes lead to confusion
- I use bold font $\mathbf{A}$ to denote the gene and italic $\mathrm{a}, A$ to denote the corresponding Allele


## Why do we care about Alleles?

- In population genetics we study the evolution of populations, that is:
- How does the frequency of alleles change over time?
- Why does the frequency change?
- As a consequence we are interested in the evolution of socalled Polymorphisms
- Polymorphism (Greek): many shapes


## Polymorphism

- Polymorphic gene
- A gene $\mathbf{A}$ in the population is polymorphic when there exist multiple alleles (e.g. A, a)
- Polymorphic site
- Today, we can sequence the entire DNA of several individuals of a population
- After multiple sequence alignment we can observe sites in certain genes with more than one state
- Such sites are called polymorphic!


## Population genetics versus Phylogenetics

- Evolution at very different scales
- In an alignment of individuals of a single population (species) there will be far less mutations than in the phylogeny of mammals, for instance!
- Since in population genetics there are so few mutations and each mutation is much more important we need to absolutely get the alignment right!


## An Alignment Of Individuals



## An Alignment of Species



CACCCCCACTTGGAACCGCTACCTTCGGCTTCTGCGTTTTACCAGACACACTGGCAGGTTGT | CGTCTTCAACTGAACCGCTACCCCCAGCTCCGAAGTCTCACCAGACGC???????????? |
| :--- |
| $C A T C T T C A C C T G A A C C G C T A C C C C C A G C T C C G A A G T C T C A C C A G A C G C A T C A T T G C C ? ? ? ~$ |


 TATCTTCACCTGAACAGCTACCCCCAGCTCCGAAGTCTCACCAGACGCATCATTGCCITCT TATCTTCACCTGAACAGCTACCCCCAGCTCCGAAGTCTCACCAGACGCATCATTG? ? ? ? ? CACTCTCACTTGAGCCGCTACCCTCGGTTTC CACTCTCACTTGAGCCGCTACCCTCGGTTTC ???TCTAATTCCCGTCGTTACCCTTGGTTTA CACTCTCATTTGAGCCGCTACCCTTGGTTTC CACTCTCATTTGAGCCGCTACCCTTGGTTTC CACT CTCATTTGGAGCCGCTACCCTTTGGTTTC
 CACTCTCATTTGAGCCGCTACCCTTGGTTTC ? ACTCTCATTTGAGCCGCTAYCCTTGGTTTC CACTCTCATTTGAGCCGCTAYCCTTGGTTTC
 CACTCTCATTTGAGCCGCTATCCTTGGTTTC CACTCTCATTTGAGCCGCTATCC?TGGTTTC CACTCTCATTTGAGCCGCTATCCTTGGTTTC CACTCTCACTTGAGCCGCTACCCTCGGTTTC CACTCCCACTTGAGCCACTACCCTCGGTTTC CAACTCCCCACTTGGAGCCACTTACCCTTCGGTTTTC CACTCTCAACTTGAGCCGCTACCCTCGGTTTC CACTCYCACTTGAGCCRCTRCCCTCGGTTTC CACTCYCACTTGAGCCRCTACCCTCGGTTTC CACTCYCACTTGAGCCRCTACCCTCGGTTTC CACTCTCACTTGAGCCGCTACCCTCGGTTTC CACTTCTCACTTG A GCCCGCTACCCTCGGTTTTC CACTCTCACTTGAGCCGCTACCCTCGGTTTC CACTCTCACTTGAGCCGCTACCCTCGGTTTC ?? ? CTCACTTGAGCCGCTACCCTCGGTTTC CACTTCTCACTTGAGCCGCTACCCTCGGTTTC
 CACTCTTCACTITGAGCCGCAACCCTCGGTTTTC CACTCTTCACTTGAGCCGCAACCCTCGGTTTC CACTCTTCACTTGAGCCGCTGCCCTCGATTTC CACTCTCGCTTGAGCCGCTACCCTCGGTTTC

GTCCTACCAGACGCATCGCTGCCTGT
 GTCCTACCAGACGCATCGC??????


 $\begin{array}{ll}\mathrm{GT} C & C T A C C A G A C G C A T C G C T G C \\ G T C C T A C C A G A C G C A T C G C T G C C T G T\end{array}$ GTCCTACCAGACGCATCGCTGCCTG? GTCCTACCAGACGCATCGCTGCCTGT GTCCTACCAGACGCATCGCTGCCT GT GTCCTACCAGACGCATCGCTGCCTGT GTCC?ACCAGACGCATCGCTGCCTGT GTCCTACCAGACGCATCGCTGCCTGT GCCTTACCGGACGCATCGC:GCCTAT GTCTTACCGGACGCATCGCTGCCTAT GTCTTAACCGGACGCT TCGCTGCC TAAT
 GTCTYACCGGACGCTTCGCTGCCTAT GTCTTACCGGACGCTTCGCTGCCTA? STCTTACCGGACGCTTCGCTGCCTAT CTCTTACCGGACG????????????? CTCTTACCGGACGCTTCGCTGCCTAT GTCTTACCGGACGCATCGCTGCC A A GTCTTACCGGACGCATCGCTGCCTAT GTCTTACCGGACGCATCGCTGCCTAT GTCTTACCGGACGCATCGCTGCCTAT GTCTTTATGGCCGCATCGCTGCCTAC GTCTTACCGGACGCATCGCTGCCTAT GTCTTACCGGATGCATCGCTGCCTAT


Boletus is a Fungus

## Polymorphic Sites - SNPs

- In the MSA of the individuals, we observe some sites, that have more than one nucleotide state
- Such sites are called Polymorphic sites or more commonly SNPs = Single Nucleotide Polymorphisms
- SNPs is pronounced: Snips
- Modern population genetic analyses mostly operate on SNPs


## Modern Population Genetics

- Study of polymorphisms in a population
- Which processes introduce polymorphisms into the population?
- If a polymorphisms exists in a population will it be there for ever?
- Is there some process that removes polymorphisms from the population?
- Do polymorphisms exhibit patterns?


## A simple Hypothesis \& Model

- Question: Does dominance affect the frequency of alleles?
- First tested by the famous mathematician G. Hardy at the beginning of the $20^{\text {th }}$ century
- Assume
- infinite population size
- random mating


## Random Mating



## A simple Hypothesis \& Model

- Question: Does dominance affect the frequency of alleles?
- First tested by the famous mathematician G. Hardy at the beginning of the $20^{\text {th }}$ century
- Assume
- infinite population size
- random mating
- A gene $\mathbf{A}$ with 2 alleles: $A$ and $a$
- Current frequencies (at generation 0 ) of allele pairs defining the genotype
- $f_{0}(A)=p$
- $f_{0}(\mathrm{a})=q$
- Evidently, $p+q=1$
- Does the frequency of occurrence of $A$ change over generations?
- Does the proportion of genotypes AA:Aa:aa change over generations?


## Hardy-Weinberg Equilibrium

- What happens to the frequencies of two alleles at a single gene when the four evolutionary forces (Natural selection, mutation, migration, genetic drift) are not acting on a population, and where mating is random?
- If allele frequencies are the same between a parental and offspring generation $\rightarrow$ no evolution has occurred at that gene
- Serves as null hypothesis in evolutionary biology \& population genetics


## Hardy Weinberg - the Maths

- Assumptions/Definitions (again):
- Population with 2 alleles: $A$, a
- $A$ is dominant and $a$ is recessive
- Mating is random
- Population is infinitely large
- Sexes are evenly distributed between 3 genotypes $A A$, aa and [Aa or aA]
- The ratio of frequencies for the three genotypes
$f(A A): f(A a): f(a a)=x: 2 y: z$


## Hardy Weinberg - the Maths

- We want to find out how the frequencies of the genotypes and the gametes (individual alleles) evolve
- Let's start with the mating behavior assuming random mating via the Punnett square

|  | $A A$ | $A a$ | aa |
| :---: | :---: | :---: | :---: |
| $A A$ |  |  |  |
| Aa |  |  |  |
| aa |  |  |  |

$$
f(A A): f(A a): f(a a)=x: 2 y: z
$$

## Hardy Weinberg - the Maths

- We want to find out how the frequencies of the genotypes and the gametes (individual alleles) evolve
- Let's start with the mating behavior assuming random mating via the Punnett square

|  | $A A$ | $A a$ | $a a$ |
| :---: | :---: | :---: | :---: |
| $A A$ | $x^{2}$ |  |  |
| $A a$ |  |  |  |
| $a a$ |  |  |  |

$$
f(A A): f(A a): f(a a)=x: 2 y: z
$$

## Hardy Weinberg - the Maths

- We want to find out how the frequencies of the genotypes and the gametes (individual alleles) evolve
- Let's start with the mating behavior assuming random mating via the Punnett square
$f(A A)$ * $f(A A)=x * x$

|  | $A A$ | $A a$ | aa |
| :---: | :---: | :---: | :---: |
| $A A$ | $\mathrm{x}^{2}$ |  |  |
| $A a$ |  |  |  |
| aa |  |  |  |

$$
f(A A): f(A a): f(a a)=x: 2 y: z
$$

## Hardy Weinberg - the Maths

- We want to find out how the frequencies of the genotypes and the gametes (individual alleles) evolve
- Let's start with the mating behavior assuming random mating via the Punnett square

|  | $A A$ | $A a$ | aa |
| :---: | :---: | :---: | :---: |
| $A A$ | $x^{2}$ |  |  |
| Aa | $2 x y$ |  |  |
| aa |  |  |  |

$$
f(A A): f(A a): f(a a)=x: 2 y: z
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| $A a$ | $2 x y$ | $4 y^{2}$ |  |
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## Hardy Weinberg - the Maths

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| $A a$ | $2 x y$ | $4 y^{2}$ | $2 y z$ |
| $a a$ | $z x$ | $2 y z$ | $z^{2}$ |

- Then, at the next generation $f^{\prime}()$

$$
f^{\prime}(A A)=x^{2}+x y+x y+y^{2}
$$

## Hardy Weinberg - the Maths

|  | $A A$ | $A a$ | aa |
| :---: | :---: | :---: | :---: |
| $A A$ | $x^{2}$ | $2 \times y$ | $x z$ |
| $A a$ | $2 x y$ | $4 y^{2}$ | $2 y z$ |
| $a a$ | $z x$ | $2 y z$ | $z^{2}$ |

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| $a a$ | $z x$ | $2 y z$ | $z^{2}$ |

Crossing $A A$ with Aa will yield $A A$ or $A a$ with equal probability!

- Then, at the next generation $f^{\prime}()$

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$$

## Hardy Weinberg - the Maths

|  | $A A$ | $A a$ | $a a$ |  |
| :---: | :---: | :---: | :---: | :---: |
| $A A$ | $x^{2}$ | $2 \times y$ | $x z$ | Crossing Aa with Aa will yield <br> $A A$ or aa or Aa or aA with <br> equal probability! |
| $A a$ | $2 \times y$ | $4 y^{2}$ | $2 y z$ |  |
| $a a$ | $z x$ | $2 y z$ | $z^{2}$ |  |

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$$
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- Then, at the next generation $f^{\prime}()$

$$
\begin{aligned}
& f^{\prime}(A A)=x^{2}+x y+x y+y^{2} \\
& f^{\prime}(A a)=\ldots=2(y+z)(x+y) \\
& f^{\prime}(a a)=\ldots=(y+z)^{2}
\end{aligned}
$$

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- Thus, the ratio now is:

$$
f^{\prime}(A A): f^{\prime}(A a): f^{\prime}(a a)=(x+y)^{2}: 2(x+y)(y+z):(y+z)^{2}=x_{1}: 2 y_{1}: z_{1}
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$$

- The ratio between the genotypes remains unaltered between generations when the following holds:
- $(x+y)^{2}=x$ and $2(x+y)(y+z)=2 y$
- Remember that the original ratio at generation 0 was defined as $x: 2 y: z$ and that the equality of $z$ to $(y+z)^{2}$ follows because $x+2 y+z=1$ !


## Hardy Weinberg - the Maths

- Let's look at deducing $(x+y)^{2}=x$

$$
\begin{aligned}
& x^{2}+2 x y+y^{2}=x \rightarrow x(x+y)+y(x+y)=x \rightarrow y(x+y)=x(1-x-y) \\
& \rightarrow y(x+y)=x(y+z)[\text { remember } x+2 y+z=1!]
\end{aligned}
$$

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& y y+y^{2}=x y y+x z \rightarrow y^{2}=x z
\end{aligned}
$$

- Obviously, this holds for the frequencies after the first generation.


## Remember

$$
\begin{aligned}
& f^{\prime}(A A)=(x+y)^{2} \leftarrow \text { this is our } x \text { above } \\
& f^{\prime}(A a)=2(y+z)(x+y) \leftarrow \text { this is our } y \text { above } \\
& f^{\prime}(a a)=(y+z)^{2} \leftarrow \text { this is our } z \text { above }
\end{aligned}
$$

- Thus, genotypic frequencies will remain constant FROM the first generation.


## Hardy Weinberg - the Maths

- Let's look at the gametic frequencies now
- At generation 0 (remember: $f(A A): f(A a): f(a a)=x: 2 y: z)$ they are:

$$
\begin{aligned}
& f(A)=x+y \\
& f(a)=y+z
\end{aligned}
$$

- At generation 1 they are:

$$
\begin{aligned}
& f^{\prime}(A)=(x+y)^{2}+(x+y)(y+z)=x+y \\
& f^{\prime}(a)=. \boldsymbol{A}=y+z
\end{aligned}
$$

$$
\begin{aligned}
& \text { remember } \\
& f^{\prime}(A A)=x^{2}+x y+x y+y^{2} \\
& f^{\prime}(A a)=\ldots=2(y+z)(x+y) \\
& f^{\prime}(a a)=\ldots=(y+z)^{2}
\end{aligned}
$$

## Hardy Weinberg - the Maths

- Let's look at the gametic frequencies now
- At generation 0 (remember: $f(A A): f(A a): f(a a)=x: 2 y: z)$ they are:
$f(A)=x+y$
$f(\mathrm{a})=y+z$
- At generation 1 they are:

$$
\begin{aligned}
& f^{\prime}(A)=(x+y)^{2}+(x+y)(y+z)=x+y \\
& f^{\prime}(a)=\ldots=y+z
\end{aligned}
$$

- Thus, allelic frequencies will remain constant even from generation 0 onwards!


## Effects of finite Population Size Random Genetic Drift

- Populations are of finite size!
- Does this affect the evolution of allele frequencies over generations?
- Assume:
- there are $N$ individuals in a diploid population $\rightarrow 2 N$ chromosomes
- Frequency of $A$ allele is $p$
- What will be the frequency of $A$ in the next generation?


## Random Genetic Drift

- Definition:

Genetic drift is a random process that causes changes in allele frequencies from one generation to the next. Some alleles will be passed on to the next generation disproportionally without being advantageous or harmful. Especially in small populations genetic drift is strong due to sampling errors. Alleles can be fixed or get lost by chance.

## The Wright-Fisher Model for finite populations

- Assume a diploid population:
- Population size: $N$ ( 2 N chromosomes)
- Random mating
- Non-overlapping generations $\rightarrow$ something like discrete time steps from generation to generation (e.g., annual plants)
- No natural selection
- Equal distribution of sexes
- The Wright-Fisher model is the simplest model of evolution for a population of finite size


## Wright-Fisher Rules/Simulation Example

- We assume a constant population $\rightarrow$ say 10 individuals (or 5 diploid individuals) per generation
- Each individual from the offspring generation picks a parent at random from the previous generation
$\rightarrow$ all parents have equal probability to be picked
$\rightarrow$ a parent can be picked more than once
- Each offspring inherits the genetic information of the parent
- The process and maths are easier to understand if we forget about alleles for a second and just think about individuals


## Wright-Fisher

Gen 0

## Wright-Fisher

time


## Wright-Fisher

time

Gen 1

offsprings

## Wright-Fisher

time


## Wright-Fisher Binomial Random Sampling

- The probability to pick an individual $X$ as ancestor of an individual in the next generation is $p=1 / 2 \mathrm{~N}$
- If the population remains constant then you have to sample $2 N(2 N=10$ in our example) times from the current generation to construct the next generation with 2 N offsprings
- For every sample, the probability to pick $X$ remains constant at $p \rightarrow$ by definition of our model
- The number of offsprings for $X$ follows a binomial distribution, thus the probability to pick $X$ as an ancestor $k$ times is

$$
\mathbf{P}[X=k]=\binom{n}{k} p^{k}(1-p)^{n-k}
$$

- Where $p:=1 / 2 \mathrm{~N}$ and $n:=2 \mathrm{~N}$


## Binomiat Ranoonsing

- The probability to pick an allele $A$ as ancestor of an individual in the next generation is $p=\# A / 2 N$
- If the population remains constant then you have to sample $2 N(2 N=10$ in our example) times from the current generation to construct the next generation with 2 N offsprings
- For every sample, the probability to pick $A$ remains constant at $p \rightarrow$ by definition of our model
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$$

- Where $p:=\# A / 2 N$ and $n:=2 N$


## Binomial Sampling of Alleles




Binomial distributions for frequency of allele $A$ in the next generation for $p=f(A)=0.4$ and $p=f(A)=0.2$ and a population size of $2 N=10$

# Mean and Variance of Allelic Frequency due to drift 

- From the properties of the binomial distribution we obtain
- $E(\# A)=2 N * p$
- $\operatorname{Var}(\# A)=2 N * p *(1-p)$


## The evolution of the frequency of $A$ as a Markov Chain

- The evolution of the frequency of $A$ over generations is a stochastic process!
- Even if we know everything about the population we cannot predict the state at the next generation with certainty
- One important property of the process: the next state depends only on the current state
$\rightarrow$ The process can be modeled as a Markov Chain


# Transition Probabilities Wright-Fisher 

Frequency in next generation $t+1$
Frequency in current generation $t$
Probability of changing from $i$ alleles in generation $t$ to $j$ alleles in generation $t+1$

$$
\begin{aligned}
& \operatorname{Prob}\{X(t+1)=j \mid X(t)=i\} \\
& =p_{i j}=\binom{2 N}{j}\left\{\frac{i}{2 N}\right\}^{j}\left\{1-\left(\frac{i}{2 N}\right)\right\}^{2 N-j} \\
& i, j=0,1,2, \ldots, 2 N
\end{aligned}
$$

## Example

- Prob of change from $i=4 \rightarrow j=8$ Alleles of same type for a population of size $2 \mathrm{~N}:=10$ from one generation to the next

$$
p_{4,8}=\binom{10}{8}\left(\frac{4}{10}\right)^{8}\left(1-\left(\frac{4}{10}\right)\right)^{10-8}=0.0106168
$$

## Wright-Fisher Model

- A state of a Markov process is called absorbing when the probability to exit this state once we have entered it is 0.
- Are there absorbing states in the Wright-Fisher model?


## Probability to enter an absorbing state

- Useful to study the evolution in a Wright-Fisher model as a Markov Chain because you can answer a lot of questions via standard Markov Chain theory.
- For instance: What is the probability that the population will end up (after how many generations?) in the absorbing state where $f(A)=1$ ?
$\rightarrow$ this is also called fixation
- Given that the frequency of $A$ is \#A/2N, the probability that $A$ will become fixed is \#A/2N
- For details, see:
http://people.sc.fsu.edu/~pbeerli/isc5317-notes/pdfs/01-populationmo des.pdf


## Random genetic Drift

- The change in allele frequencies over generations in finite populations due to stochasticity (re-sampling) is called random genetic drift
- What is the effect of random genetic drift on the polymorphism level?
- Since our human population is finite, why do we still observe polymorphisms?


## Heterozygosity and Genetic Drift

- Reduction of polymorphism is quantified by the degree of homozygosity $\rightarrow$ The probability that two alleles are identical
$\rightarrow$ heterozygosity $=(1$ - homozigosity $)$ at generation $t$ is defined as: Het $_{t}$
- Assume a population of size 2 N
- We can define the heterozygosity recursively as $\operatorname{Het}_{t}=\operatorname{Het}_{(t-1)}(1-1 / 2 N)$
- Thereby we obtain: $\operatorname{Het}_{t}=\operatorname{Het}_{0}(1-1 / 2 N)^{t}$

Probability that two randomly chosen Alleles are different




Initial allele frequency $f(A)=0.5$

## Mutation-Drift Balance

- Genetic drift removes polymorphisms (SNPs) from the population
- Mutations introduce polymorphism (SNPs) into the population
- Is there some balance?


## Heterozygosity at mutation - drift balance

- Define:
- Het: heterozygosity
- $-1 / 2 N$ * Het: Loss of heterozygosity per generation due to genetic drift
- $\mu$ : mutation rate per gene (remember two alleles per gene!) and per generation
- $2 \mu(1-H e t):$ gain of heterozygosity due to mutation
- Pick two alleles
- Consider transition from generation $t \rightarrow t+1$
- The probability that they are identical is: (1-Het)
- If they are identical, the probability that one out of the two will mutate is $2 \mu$
$\rightarrow 2 \mu(1-H e t)$ gain in heterozygosity due to mutation
- Overall: Het $_{t+1}=$ Het $_{t}-1 / 2 N *$ Het $_{t}+2 \mu(1-$ Het $)$
$\Delta H e t=-1 / 2 N * H e t_{t}+2 \mu\left(1-\right.$ Het $\left._{t}\right)$
- $\Delta H e t=0 \rightarrow$ Het $=(4 \mu N) /(1+4 \mu N)$


## Rate of Evolution by mutation and genetic drift

- Rate of Evolution = The probability of a new mutation to arise in the population and to eventually become fixed
- Assume
- $\mu$ is the probability of mutation per generation and per individual
- $2 N$ individuals $\rightarrow 2 N \mu$ mutations per generation
- The probability that a particular mutation will be fixed is $1 / 2 \mathrm{~N}$
- Thus, the rate at which a mutation will arise and fix in the population is $1 / 2 \mathrm{~N} * 2 \mathrm{~N} \mu=\mu$
- Why is this result remarkable?


## Natural Selection

- So far, we have assumed that the probabilities of fitness and reproduction are the same for each individual, independently of its genotype
- Consequently, a random individual at generation $t+1$ descends from any individual in generation $t$, with the same probability
- We denote the ability of an individual "to survive and reproduce" as fitness
- We assume that fitness depends on the genotype


## Natural Selection

- The term selection means that a genotype reproduces more frequently than others
- If a certain genotype, e.g., $A A$ has better/higher fitness
$\rightarrow$ it will fix in the population after several generations
$\rightarrow$ consequently, the allele $A$ will also fix
- We say: Natural selection has favored allele $A$
- In this case, the natural selection on $A$ is termed Positive Selection


## Different Modes of Selection



## The Frequency Evolution of $A$ under Positive Selection

Positive selection

Random genetic drift


## Summary Statistics

- Summary statistics provide a summarized description of the dataset, e.g., the number of polymorphic sites
- Summary statistics are important because:
- They allow to estimate parameters of the population
- They help us to assess if positive selection occurred
- Differences to phylogenetics
- Given the data (MSA of individuals)
- We don't reconstruct a population tree for the individuals
- We simulate evolution under different scenarios (including more complex models with changing population sizes etc)
- Then we compare if one of the scenarios fits the summary statistics (e.g. \# SNPs) of our empirical dataset

