Population genomics in 2024



Andrea Manica



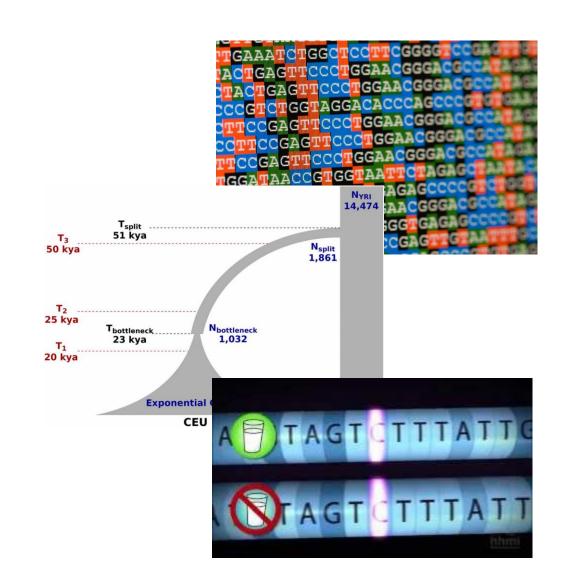
Dept of Zoology

Overview

Type of data

Demography

Selection

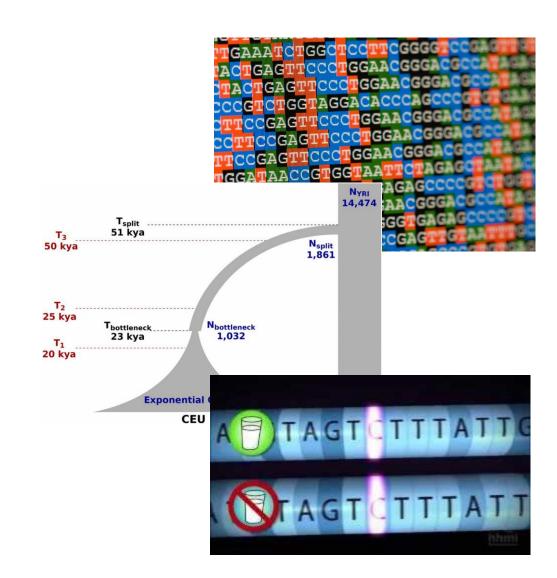


Overview

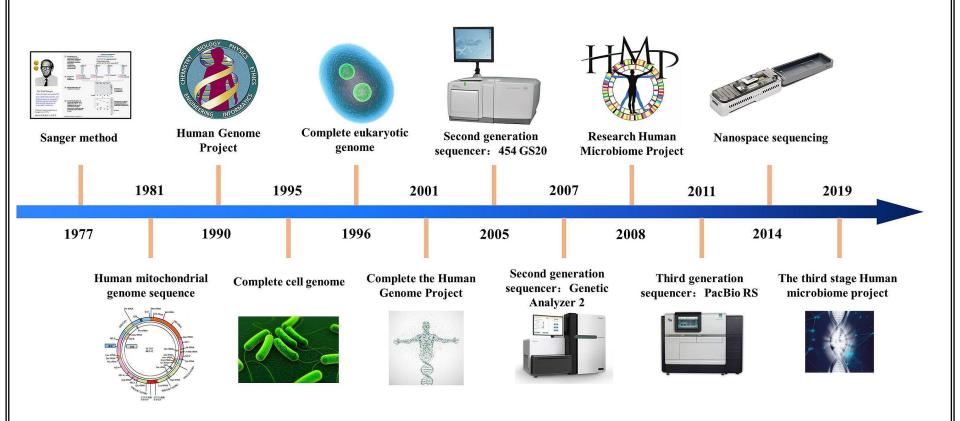
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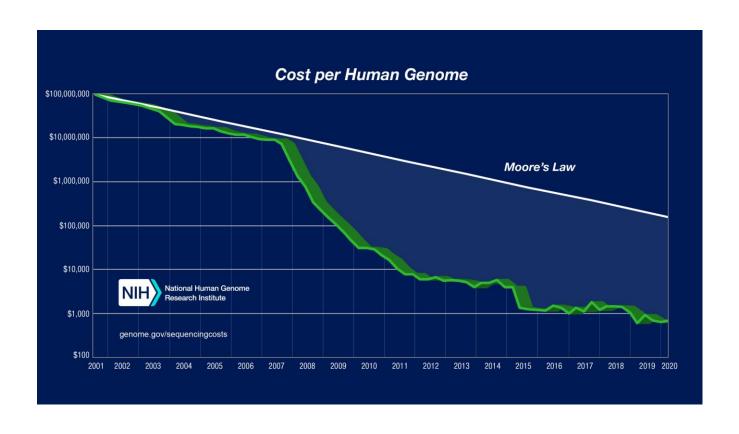
Selection



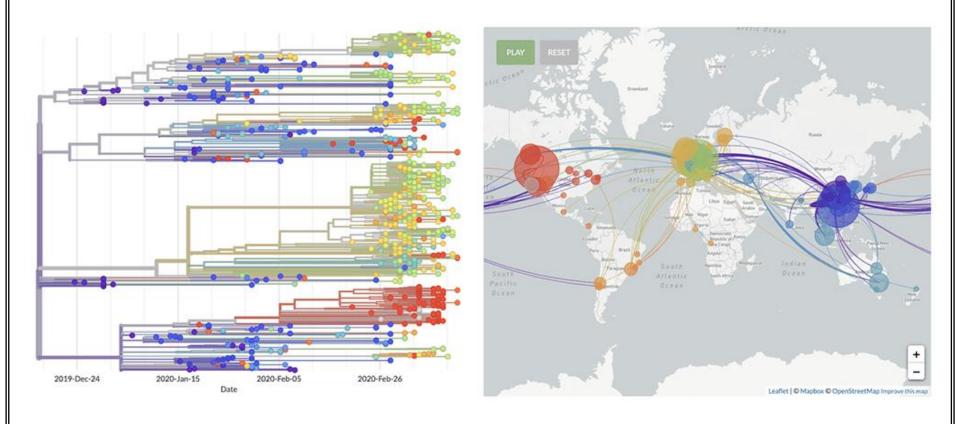
Technological advances over the years



The cost of a genome



Genomics on a big scale: an example

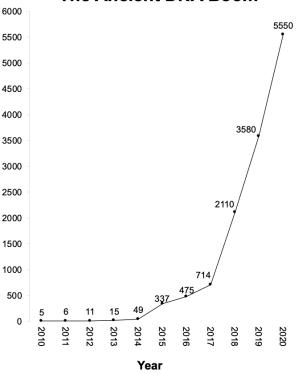


Tracing covid through genomics

Ancient DNA



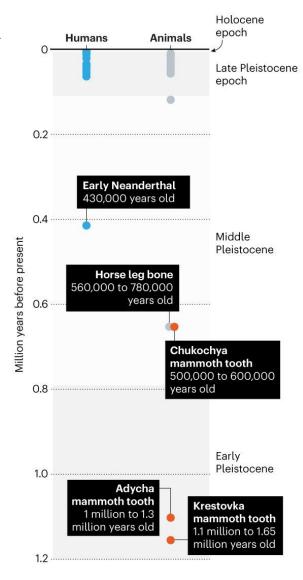




Callaway (2018) Nature

But beware of the ethical issues of aDNA from human remains

Destructive sampling



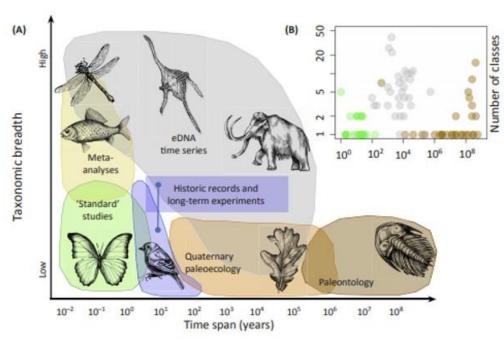
The upper age bound for the mammoth teeth is based on a genetic dating method; the lower bound is based on the age of the sediments in which the teeth were found.

onature

Environmental DNA







But we need extensive databases for eDNA!!!

Challenges: non-model species

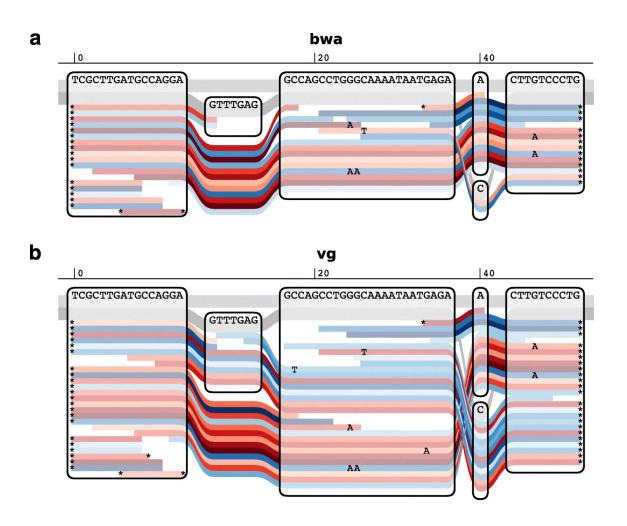


Long and linked reads have dramatically decreased the cost of new reference genomes (~US\$ 3k)

A solution for building the extensive databases needed for eDNA?

Low coverage and genotype likelihoods

Challenges: biases from reference



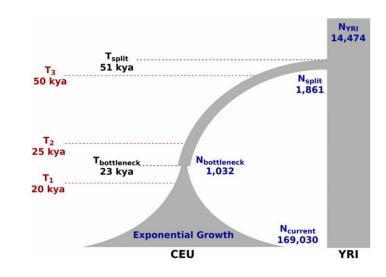
Graph based methods hold a lot of promise

Overview

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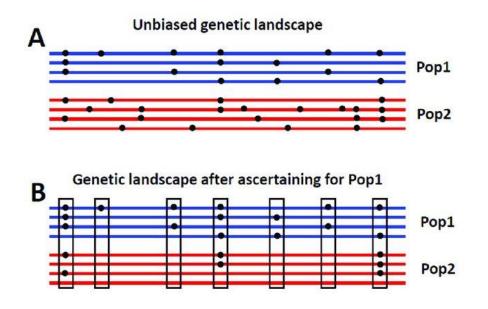
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More of the same?

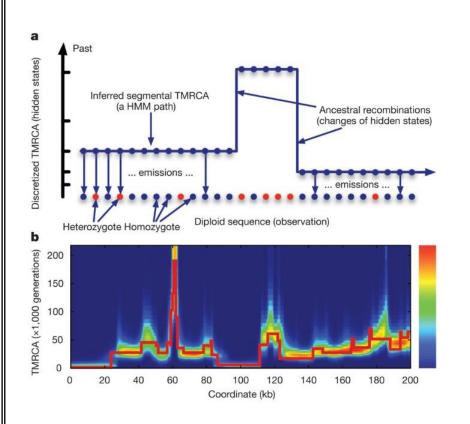
Many approaches based on metrics that describe SNP frequencies (especially the Site Frequency Spectrum)

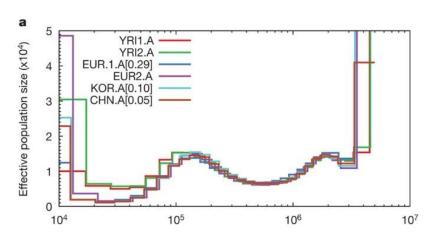


Very challenging to model ascertained data

Whole genomes solve the problem of ascertainment present in SNP chips!!!

The power of whole genomes

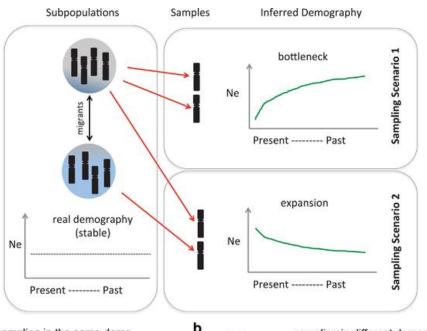


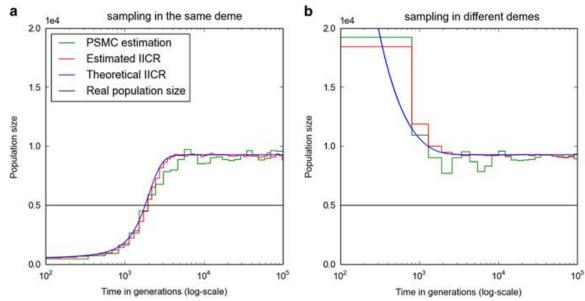


But it assumes a single panmictic population (same assumption of Bayesian Skyline Plots)

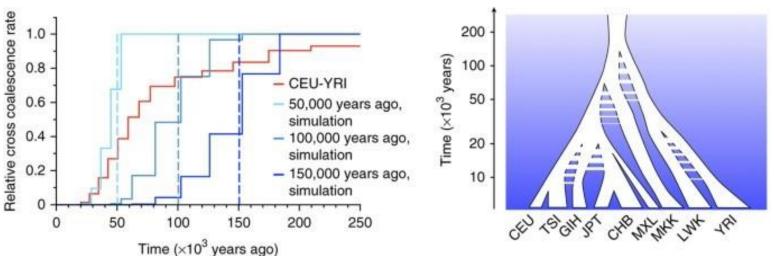
Be careful about the assumptions behind methods, they can affect your interpretation

Beware of assumptions





The power of whole genomes

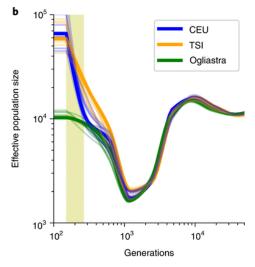


MSMC extends to multiple genomes (but only a few,

and need to be phased...)

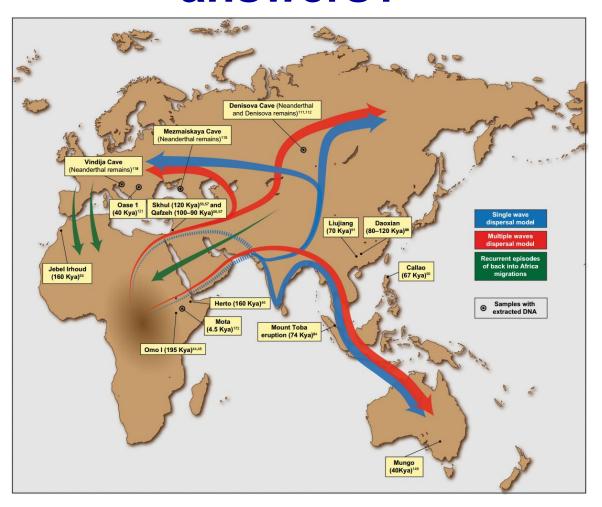
Split times assume a clean split without migrations

Methods coming out to deal with 1000s of genomes (SMC++, Relate)



Schiffel & Durbin (2014) Nat Gen, Chiang et al. (2018) Nat Gen

But does more data mean better answers?



The story of 3 papers...

But does more data mean better answers?

ARTICLE

One wave only!

doi:10.1038/nature18299

A genomic history of Aboriginal Australia

ARTICLE

One wave (but it could have been two...)

doi:10.1038/nature18964

The Simons Genome Diversity Project: 300 genomes from 142 diverse populations

A list of authors and affiliations appears at the end of the paper.

LETTER

Two waves!

doi:10.1038/nature19792

Genomic analyses inform on migration events during the peopling of Eurasia

A list of authors and affiliations appears at the end of the paper

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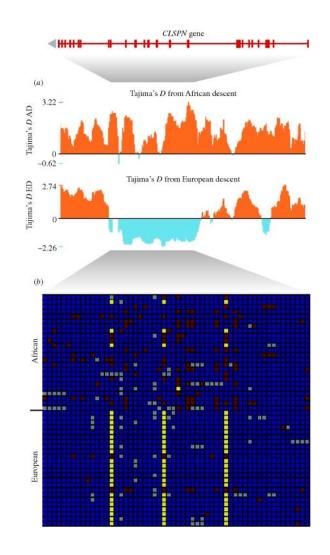
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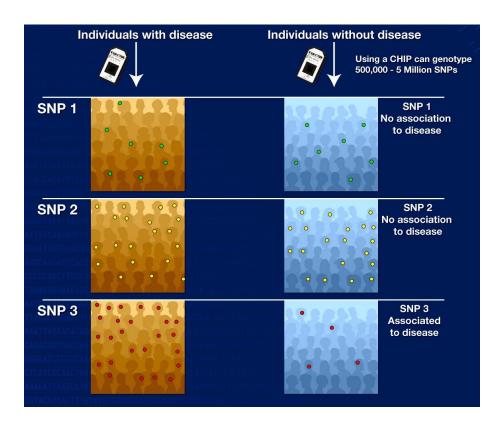
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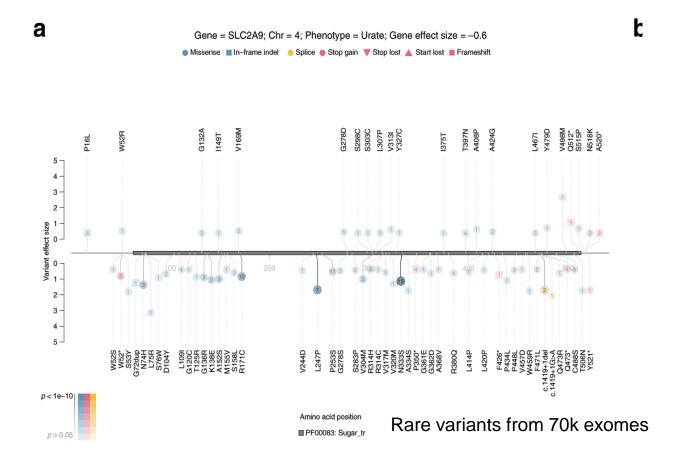
More of the same?

Selection scans and GWAS already possible with SNP chips



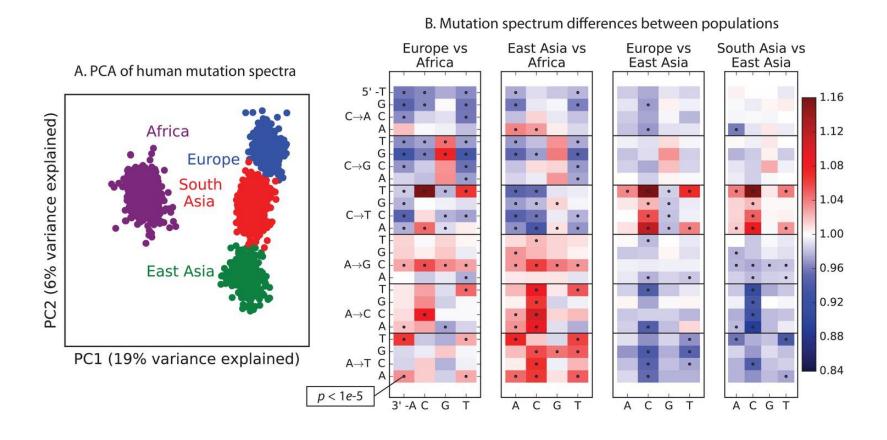


Rare alleles

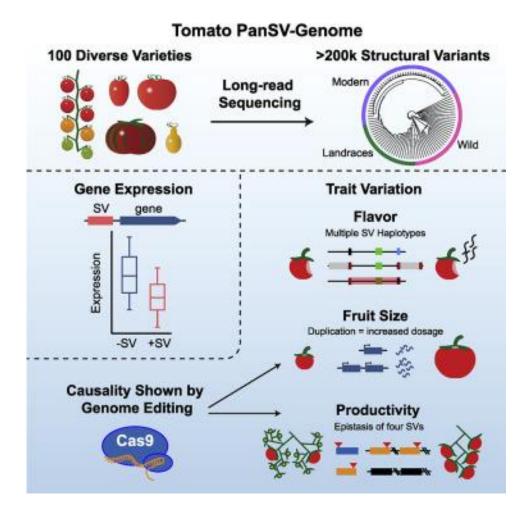


But still a big need for functional validation of candidates! False positives are a big issue in selection studies.

Mutation spectrum



Structural variation



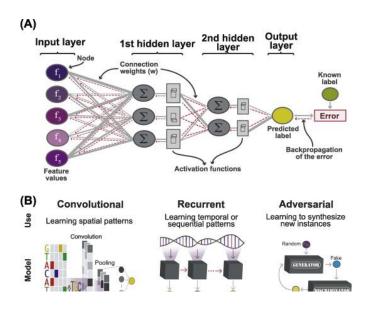
Long read sequencing helps a lot with structural variation!

Metagenome Sequencing machine Taxonomic profiling Gene & function profiling Taxonomy tree Gene function databas

Phenotypes might be greatly affected by other organisms

Zhang et al (2016) IEEE proceedings

Big data and Machine Learning



A black box? Not necessarily

Trends in Genetics



Review

Opening the Black Box: Interpretable Machine Learning for Geneticists

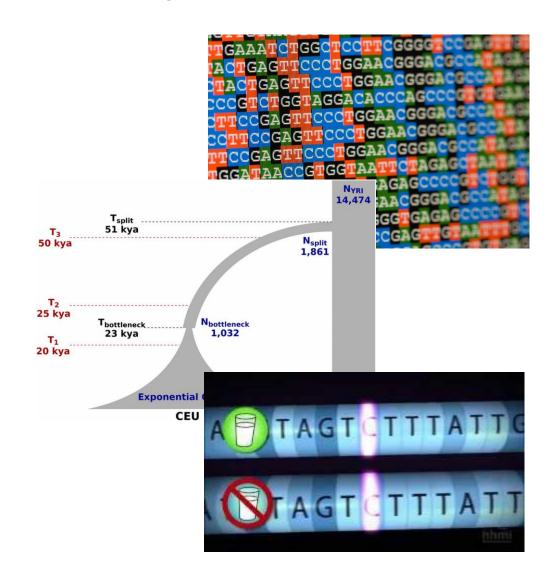
Christina B. Azodi, 1,2,* Jiliang Tang, and Shin-Han Shiu , 1,4,*

Summary

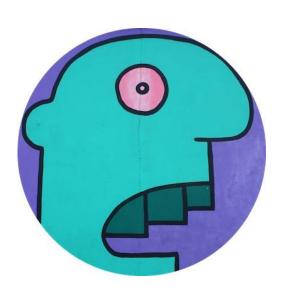
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EMBO Population Genomics 2024





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