sea urchin (single population; split along a gap

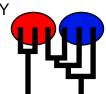
POPULATION SIZE, MIGRATION, DIVERGENCE, ASSIGNMENT, HISTORY

Bayesian inference using the structured coalescent

Migrate-n version 5.0.7 [May-01-2025]

Program started at Sat May 3 12:14:14 2025

Program finished at Sat May 3 12:22:58 2025 [Runtime:0000:00:08:44]



Options

Inheritance multipliers in use for Thetas:

All loci use an inheritance multiplier of 1.0

Random number seed: (with internal timer) 1183202194

Start parameters:

Theta values were generated

Using a percent value of the prior

M values were generated Using a percent value of the prior

Connection matrix:

m = average (average over a group of Thetas or M,

s = symmetric migration M, S = symmetric 4Nm,

0 = zero, and not estimated,

* = migration free to vary, Thetas are on diagonal

d = row population split off column population, D = split and then migration

Population 1 1 1 1 1 1 1 mogap * *

Order of parameters:

1 Θ_1 <displayed>

Mutation rate among loci: Mutation rate is constant for all loci

Analysis strategy:

Bayesian inference

-Population size estimation: Exponential Distribution

Proposal distributions for parameter

Parameter Proposal Theta Slice sampling

M Slice sampling
Genealogy Metropolis-Hastings

Prior distribution for parameter

 Parameter
 Prior
 Minimum
 Mean*
 Maximum
 Delta
 Bins
 UpdateFreq

 1
 Theta
 *
 *
 Uniform
 0.000000
 0.050
 0.100
 0.010000
 1500
 0.50000

[* * means priors were set globally]

Posterior distribution:

Parameter values were collected using MCMC, these values were then used to generate the posterior histograms using KERNEL SMOOTHING (window=41) and subsequent SAVITZKY-GOLAY SMOOTHING (window=41) for combination over loci

Markov chain settings:

Number of chains

Recorded steps [a]
Increment (record every x step [b]
Number of concurrent chains (replicates) [c]
Visited (sampled) parameter values [a*b*c]

Long chain

1

2

1000000

Number of discard trees per replicate (burn-in * b) 1000000

Multiple Markov chains:

Static heating scheme 4 chains with temperatures

100000.00 3.00 1.50 1.00

Swapping interval is 1

Print options:

Data file: infile.gap

parmfile.gap

Haplotyping is turned on:

Output file: outfile-gap
Posterior distribution raw histogram file: bayesfile

Raw data from the MCMC run: bayesallfile-gap
Print data: No

Print genealogies [only some for some data type]:

Data summary

Data file	e:						infile.gap
Datatyp	e:					Ha	plotype data
Numbei	r of loci:						5
	nmodel:						
Locus S	Sublocus	Mutationmo	del M	utationmodel	parameters		
1	1	Jukes-Canto	r [Base	efreq: =0.25]			
2	1	Jukes-Canto	r [Base	efreq: =0.25]			
3	1	Jukes-Canto	r [Base	efreq: =0.25]			
4	1	Jukes-Canto	r [Base	efreq: =0.25]			
5	1	Jukes-Canto	r [Base	efreq: =0.25]			
Sitoo no	or loous						
Sites pe	ei iocus	Sites					
1		252					
2		921					
3		425					
4		459					
5		713					
		7.10					
Site rate	e variation	and probabiliti	es:				
			Rate of change	Probability	Patch size		
1	1	1	1.000	1.000	1.000		
2	1	1	1.000	1.000	1.000		
3	1	1	1.000	1.000	1.000		
4	1	1	1.000	1.000	1.000		
5	1	1	1.000	1.000	1.000		
Populat	ion				Locus	Gene co	pies
						data	(missing)
1 urchir	ns				1	24	, ,
					2	24	
					3	24	
					4	24	
					5	24	
1 nogap)				1	34	
I					•		

2

34

		population, split ai	<u> </u>
	3	34	
	4	34	
	5	34	
Total of all populations	1	58	(0)
	2	58	(0)
	3	58	(0)
	4	58	(0)
	5	58	(0)
			. ,

Bayesian Analysis: Posterior distribution table

Locus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
1	Θ_1	0.00800	0.01260	0.01557	0.01933	0.02960	0.01717	0.01797
2	Θ_1	0.03013	0.03913	0.04403	0.05120	0.06547	0.04623	0.04703
3	Θ_1	0.04067	0.05200	0.06090	0.06827	0.08693	0.06183	0.06274
4	Θ_1	0.01673	0.02333	0.02817	0.03273	0.04473	0.02937	0.03009
5	Θ_1	0.05313	0.06427	0.07103	0.08227	0.09847	0.07410	0.07441
All	Θ_1	0.00893	0.01587	0.02263	0.02667	0.03660	0.03737	0.04226

Citation suggestions:

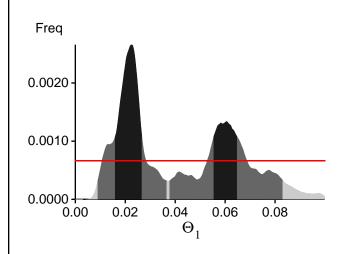
Beerli P., 2006. Comparison of Bayesian and maximum-likelihood inference of population genetic parameters. Bioinformatics 22:341-345

Beerli, P., H. Ashki, S. Mashayekhi, and M. Palczewski, 2022. Population divergence time estimation using individual lineage label switching. G3 Genesâ Genomesâ Genetics, 12(4), 02 2022.

Beerli P., 2009. How to use MIGRATE or why are Markov chain Monte Carlo programs difficult to use? In Population Genetics for Animal Conservation, G. Bertorelle, M. W. Bruford, H. C. Hauffe, A. Rizzoli, and C. Vernesi, eds., vol. 17 of Conservation Biology, Cambridge University Press, Cambridge UK, pp. 42-79.

Beerli, P., S. Mashayekhi, M. Sadeghi, M. Khodaei, and K. Shaw, 2019. Population genetic inference with migrate Current Protocols in Bioinformatics, 68(1):e87.

Bayesian Analysis: Posterior distribution over all loci



Log-Probability of the data given the model (marginal likelihood)

Use this value for Bayes factor calculations:

BF = Exp[In(Prob(D | thisModel) - In(Prob(D | otherModel) or as LBF = 2 (In(Prob(D | thisModel) - In(Prob(D | otherModel)) shows the support for thisModel]

Locus	TI(1a)	BTI(1b)	HS(3)	
1	-554.37	-521.67	-468.32	
2	-2187.57	-2046.29	-1992.25	
3	-1374.57	-1239.62	-1186.08	
4	-1109.73	-1059.75	-1013.36	
5	-3226.10	-2678.17	-2583.11	
All	-8460.76	-7553.90	-7251.53	

- (1a) TI: Thermodynamic integration: log(Prob(D|Model)): Good approximation with many temperatures
- (1b) BTI: Bezier-approximated Thermodynamic integration: when using few temperatures USE THIS!
- (3) HS: Harmonic mean approximation: Overestimates the marginal likelihood, poor variance [Scaling factor = -8.406196]

Citation suggestions:

Beerli P. and M. Palczewski, 2010. Unified framework to evaluate panmixia and migration direction among multiple sampling locations, Genetics, 185: 313-326.

Palczewski M. and P. Beerli, 2014. Population model comparison using multi-locus datasets.

In M.-H. Chen, L. Kuo, and P. O. Lewis, editors, Bayesian Phylogenetics: Methods, Algorithms, and Applications, pages 187-200. CRC Press, 2014.

Acceptance ratios for all parameters and the genealogies

Parameter	Accepted changes	Ratio	
Θ_1	2498040/2498040	1.00000	
Genealogies	484381/2501960	0.19360	

MCMC-Autocorrelation and Effective MCMC Sample Size

Parameter	Autocorrelation	Effective Sampe Size
Θ_1	0.48021	17849.29
Genealogies	0.48021	17849.29

Average temperatures during the run

Chain	Temperatures			
1	1.00000			
2	0.66667			
3	0.33333			
4	0.00001			

Potential Problems

This section reports potential problems with your run, but such reporting is often not very accurate. With many parameters in a multilocus analysis, it is common that some parameters for some loci will not be informative. These parameters then trigger suggestions to increase the prior range that are not sensible. Do not blindly follow

No warning was recorded during the run