Rhinichthys osculus dataset

POPULATION SIZE, MIGRATION, DIVERGENCE, ASSIGNMENT, HISTORY

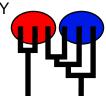
Bayesian inference using the structured coalescent

Migrate-n version 4.4.4(git:v4-series-26-ge85c6ff) [June-1-2019]

Compiled for a SYMMETRIC multiprocessors (Grandcentral)

Program started at Fri May 21 08:50:36 2021

Program finished at Fri May 21 13:09:46 2021 [Runtime:0000:04:19:10]



Options

Inheritance scalers in use for Thetas:

All loci use an inheritance scaler of 1.0

[The locus with a scaler of 1.0 used as reference]

Random number seed: (with internal timer) 1855373913

Start parameters:

Theta values were generated RANDOM start value from the prior

M values were generated RANDOM start value from 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 2 3
1 Great_Salt_lake * d 0
2 Great_Salt_Lake 0 * d
3 Sevier_Desert 0 0 *

Order of parameters:

 $\begin{array}{cccc} \mathbf{1} & & \Theta_1 & & \text{<displayed>} \\ \mathbf{2} & & \Theta_2 & & \text{<displayed>} \\ \mathbf{3} & & \Theta_3 & & \text{<displayed>} \end{array}$

4	Δ 2->1	<displayed></displayed>	
5	σ _{2->1}	<displayed></displayed>	
6	$\Delta_{3\rightarrow 2}$	<displayed></displayed>	
7	σ _{3->2}	<displayed></displayed>	

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

Analysis strategy:

Bayesian inference

-Population size estimation:

Exponential Distribution

-Geneflow estimation:

Exponential Distribution

-Divergence time estimation:

Normal Distribution Shortcut (mean and standard dev.)

Proposal distributions for parameter

Parameter Proposal
Theta Metropolis sampling
M Metropolis sampling
Divergence Metropolis sampling
Divergence Spread Metropolis sampling
Genealogy Metropolis-Hastings

Prior distribution for parameter

Par	ameter		Prior	Minimum	MeanMa	ıximum	Delta	Bins l	JpdateFreq
1	Theta '	* *	Uniform	0.000000	0.050	0.100	0.010	1500	0.07143
2	Theta '	* *	Uniform	0.000000	0.050	0.100	0.010	1500	0.07143
3	Theta '	* *	Uniform	0.000000	0.050	0.100	0.010	1500	0.07143
4	Splittime mean '	* *	Uniform	0.000000	0.250	0.500	0.050	1500	0.07143
5	Splittime std *	* *	Uniform	0.000000	0.250	0.500	0.050	1500	0.07143
6	Splittime mean 3	* *	Uniform	0.000000	0.250	0.500	0.050	1500	0.07143
7	Splittime std '	* *	Uniform	0.000000	0.250	0.500	0.050	1500	0.07143

^{[* *} means priors were set globally]

Markov chain settings:Long chainNumber of chains1Recorded steps [a]10000Increment (record every x step [b]1000Number of concurrent chains (replicates) [c]2Visited (sampled) parameter values [a*b*c]20000000Number of discard trees per chain (burn-in)1000

Multiple Markov chains:

Static heating scheme		4 chains with temperatures
	1000000.00	3.00 1.50 1.00
		Swapping interval is 1
Print options:		
Data file:		infile
Haplotyping is turned on:		NO
Output file:		outfile
Log file:		logfile
Posterior distribution raw histogram file:		bayesfile
Raw data from the MCMC run:		bayesallfile.gz
Print data:		No
Print genealogies [only some for some data type]:		None
Think genealogies formy some for some data type].		None

Data summary

Data file: infile
Datatype: Haplotype data

Number of loci: 2

Mutationmodel:

Locus Sublocus Mutationmodel Mutationmodel parameters

1 1 Kimura [Basefreq: =0.25, kappa=4.0900] 2 1 Kimura [Basefreq: =0.25, kappa=4.0900]

Sites per locus

Site rate variation and probabilities:

Locus Sublocus Region type Rate of change Probability Patch size

1	1	1	1.000	1.000	1.000		
2	1	1	1.000	1.000	1.000		
Popula	tion				Locus	Gene co	ppies
						data	(missing)
1 Grea	t_Salt_lak	е			1	10	
					2	1	
2 Grea	t_Salt_La	ke_Desert			1	22	
					2	11	
3 Sevi	er_Desert				1	21	
					2	6	
Total o	f all popul	ations			1	53	(0)
					2	18	(0)

Bayesian Analysis: Posterior distribution table

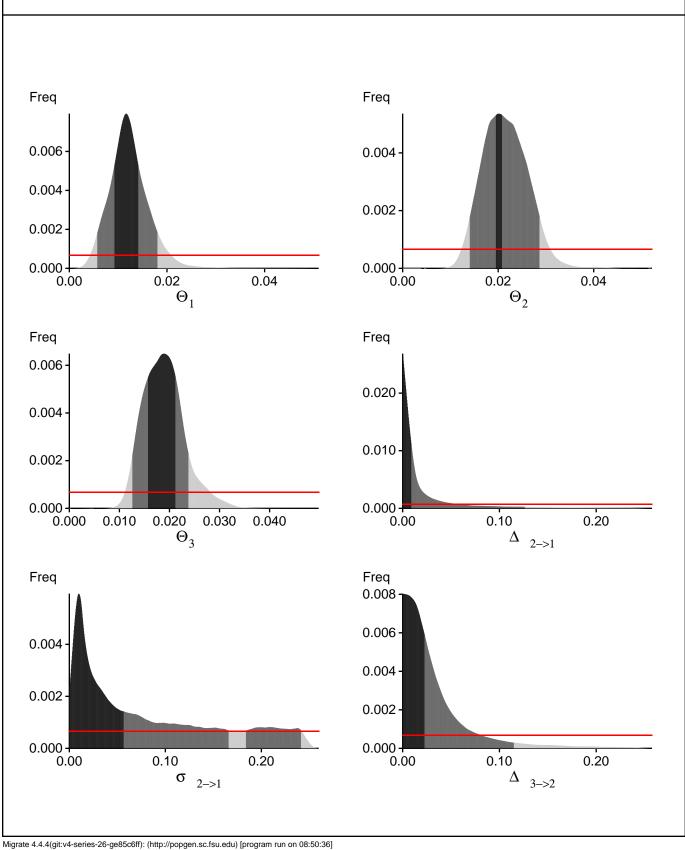
Locus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
1	Θ_1	0.00313	0.00613	0.00863	0.01153	0.02180	0.01097	0.01263
1	Θ_2	0.01020	0.01373	0.01617	0.01807	0.02533	0.01710	0.01770
1	Θ_3	0.01100	0.01740	0.01810	0.01887	0.02740	0.01843	0.01903
1	D _{2->1}	0.00000	0.00000	0.00217	0.02833	0.10567	0.03483	0.09304
1	S _{2->1}	0.00000	0.00000	0.00250	0.05000	0.08000	0.12417	0.17534
1	D _{3->2}	0.00000	0.00000	0.00750	0.04600	0.17200	0.05650	0.10392
1	S _{3->2}	0.00767	0.01200	0.04383	0.06167	0.14833	0.18050	0.21068
2	Θ_1	0.00567	0.00947	0.01230	0.01907	0.02367	0.01803	0.03188
2	Θ_2	0.01473	0.02413	0.02730	0.03020	0.04840	0.03050	0.03315
2	Θ_3^2	0.00813	0.01767	0.01997	0.02300	0.03547	0.02157	0.02364
2	D _{2->1}	0.00000	0.00000	0.00150	0.04800	0.18700	0.11650	0.15671
2	S _{2->1}	0.35833	0.36900	0.37117	0.37133	0.38067	0.27583	0.26568
2	D _{3->2}	0.00000	0.00000	0.00417	0.04900	0.18100	0.05983	0.10685
2	S _{3->2}	0.01200	0.08067	0.08517	0.08633	0.12867	0.18983	0.21641
All	Θ_1	0.00567	0.00920	0.01163	0.01413	0.01807	0.01203	0.01233
All	Θ_2	0.01400	0.01947	0.02003	0.02080	0.02867	0.02137	0.02153
All	Θ_3^2	0.01253	0.01567	0.01897	0.02120	0.02380	0.01870	0.01892
All	D _{2->1}	0.00000	0.00000	0.00017	0.00900	0.12633	0.00917	0.02602
All	S _{2->1}	0.00000	0.00000	0.00983	0.05667	0.16600	0.05683	0.08208
All	D _{3->2}	0.00000	0.00000	0.00017	0.02267	0.11500	0.02283	0.03454
All	S _{3->2}	0.01033	0.01467	0.03617	0.10600	0.21433	0.06783	0.08615

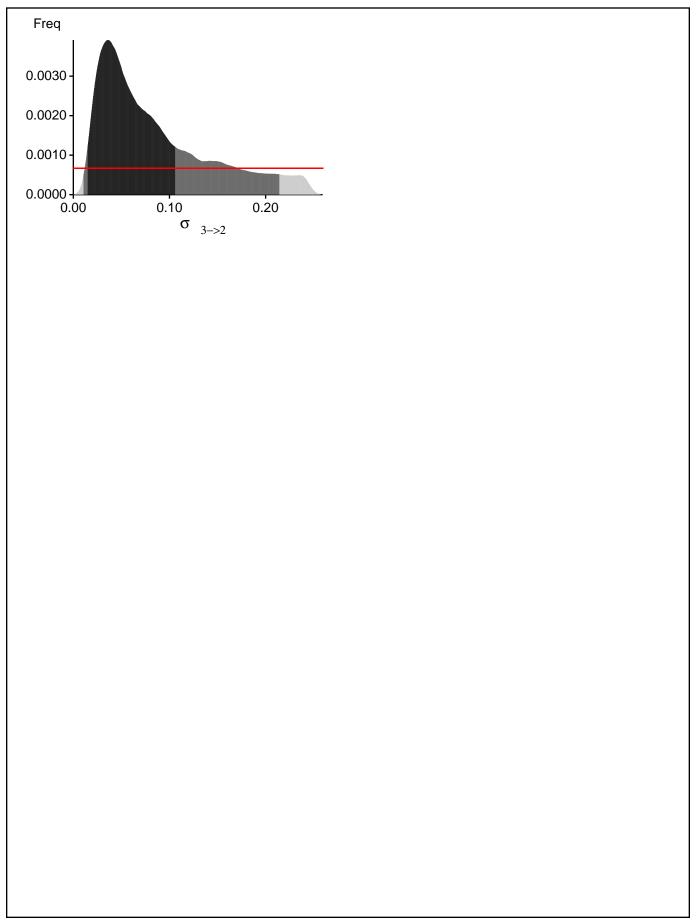
Citation suggestions:

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

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.

Bayesian Analysis: Posterior distribution over all loci





Migrate 4.4.4(git:v4-series-26-ge85c6ff): (http://popgen.sc.fsu.edu) [program run on 08:50:36]

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	Raw thermodynamic score(1a)	Bezier approximation score(1b)	Harmonic mean(2)
1	-4430.38	-3503.96	-3331.24
2	-889.71	-762.95	-741.28
All	-5345.64	-4292.46	-4098.06

(1a, 1b and 2) are approximations to the marginal likelihood, make sure that the program run long enough! (1a, 1b) and (2) should give similar results, in principle.

But (2) is overestimating the likelihood, it is presented for historical reasons and should not be used (1a, 1b) needs heating with chains that span a temperature range of 1.0 to at least 100,000.

(1b) is using a Bezier-curve to get better approximations for runs with low number of heated chains [Scaling factor = -25.547627]

Citation suggestions:

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

Acceptance ratios for all parameters and the genealogies

Parameter	Accepted changes	Ratio
Θ_1	1973546/2856659	0.69086
Θ_2	1538159/2859273	0.53795
Θ_3	1516255/2858252	0.53048
$\Delta_{2\rightarrow 1}$	1733076/2859174	0.60615
$\sigma_{2\rightarrow 1}$	1922179/2856594	0.67289
$\Delta_{3\rightarrow 2}$	1580123/2855625	0.55334
$\sigma_{3\rightarrow 2}$	1964855/2857143	0.68770
Genealogies	2176785/19997280	0.10885

MCMC-Autocorrelation and Effective MCMC Sample Size

Parameter	Autocorrelation	Effective Sampe Size
Θ_1	0.37144	22194.65
Θ_2	0.34764	23202.69
Θ_3	0.34478	23486.96
$\Delta^{\circ}_{2\rightarrow 1}$	0.09521	33523.76
$\sigma_{2\rightarrow 1}$	0.12440	32077.78
$\Delta_{3\rightarrow 2}$	0.01152	39092.25
$\sigma_{3\rightarrow 2}$	0.02834	37798.62
Genealogies	0.02834	37798.62

Potential Problems

This section reports potential problems with your run, but such reporting is often not very accurate. Whith many parameters in a multilocus analysi s, it is very common that some parameters for some loci will not be very informative, triggering suggestions (for example to increase the prior ran ge) that are not sensible. This suggestion tool will improve with time, therefore do not blindly follow its suggestions. If some parameters are fla

gged, inspect the tables carefully and judge wether an action is required. For example, if you run a Bayesian inference with sequence data, for mac roscopic species there is rarely the need to increase the prior for Theta beyond 0.1; but if you use microsatellites it is rather common that your prior distribution for Theta should have a range from 0.0 to 100 or more. With many populations (>3) it is also very common that some migration rou tes are estimated poorly because the data contains little or no information for that route. Increasing the range will not help in such situations, reducing number of parameters may help in such situations.
No warning was recorded during the run