Prosopium williamsoni dataset

POPULATION SIZE, MIGRATION, DIVERGENCE, ASSIGNMENT, HISTORY

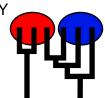
Bayesian inference using the structured coalescent

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

Compiled for a SYMMETRIC multiprocessors (Grandcentral)

Program started at Sat May 22 15:33:55 2021

Program finished at Sun May 23 00:39:41 2021 [Runtime:0000:09:05:46]



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) 4120730003

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 Logan_River * * *
2 Weber_River * * *
3 Provo_River * * *

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 chains with temperatures

							Proso	pium williar	nsoni dataset 2
4		M ₂ .	->1		<displa< td=""><td>yed></td><td></td><td></td><td></td></displa<>	yed>			
5	İ	١/	->1		<displa< td=""><td>yed></td><td></td><td></td><td></td></displa<>	yed>			
6	İ	١/١	->2		<displa< td=""><td>yed></td><td></td><td></td><td></td></displa<>	yed>			
7	ĺ	١./١	->2		<displa< td=""><td>yed></td><td></td><td></td><td></td></displa<>	yed>			
8	I	١/١	->3		<displa< td=""><td>yed></td><td></td><td></td><td></td></displa<>	yed>			
9	I	١./١	->3		<displa< td=""><td>yed></td><td></td><td></td><td></td></displa<>	yed>			
Mutati	on rate amoi	ng lo	oci:				Mutation	rate is con	stant for all loci
Analys	sis strategy:							Bay	esian inference
-Popu	ılation size e	stim	ation:					Exponer	tial Distribution
-Gene	eflow estimat	ion:						Exponer	tial Distribution
Propos	sal distributio	ons f	or paramete	er					
Parame					oosal				
Theta			N	1etropolis sam					
М				letropolis sam					
Diverge	ence			letropolis sam					
_	ence Spread			letropolis sam					
Geneal	-			letropolis-Has					
	listribution fo	r pa		N. Aliani in annuara	N/ 1 /		Dalla	Dia a	la data Faa n
Parame		**	Prior	Minimum	Mean i M		Delta		JpdateFreq
1	Theta	**	Uniform	0.000000	0.010	0.100	0.010	1500	0.05556
2	Theta		Uniform	0.000000	0.010	0.100	0.010	1500	0.05556
3	Theta	**	Uniform	0.000000	0.010	0.100	0.010	1500	0.05556
4	М	**	Uniform	0.000000	100.0	1000.	100.0	1500	0.05556
5	M	**	Uniform	0.000000	100.0	1000.	100.0	1500	0.05556
6	M	**	Uniform	0.000000	100.0	1000.	100.0	1500	0.05556
7	M	**	Uniform	0.000000	100.0	1000.	100.0	1500	0.05556
8	M	**	Uniform	0.000000	100.0	1000.	100.0	1500	0.05556
9	M	**	Uniform	0.000000	100.0	1000.	100.0	1500	0.05556
[* * me	eans priors w	ere	set globally]					
Marko	v chain settii	ngs:							Long chain
Numbe	er of chains								1
Rec	orded steps	[a]							10000
	ement (recor		ery x step [b]					1000
Number of concurrent chains (replicates) [c]							2		
Visited (sampled) parameter values [a*b*c]						20000000			
	nber of disca								1000
Multipl	le Markov ch	ains	:						
wantipi			•				_		

Static heating scheme

	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 No
		None
Print genealogies [only some for some data type]:		None

Data summary

Data file: infile
Datatype: Haplotype data

Number of loci:

Mutationmodel:

Locus Sublocus Mutationmodel Mutationmodel parameters

1 1 HKY [Bf:0.23 0.29 0.18 0.30, kappa=3.010] 2 1 HKY [Bf:0.25 0.31 0.17 0.27, kappa=3.010]

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	ation				Locus	Gene co	ppies
						data	(missing)
1 Loga	n_River				1	25	
					2	25	
2 Web	er_River				1	28	
					2	35	
3 Prov	o_River				1	16	
					2	25	
Total o	of all popul	ations			1	69	(0)
					2	85	(0)

Bayesian Analysis: Posterior distribution table

Locus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
1	Θ_1	0.00013	0.00020	0.00037	0.00047	0.00073	0.00050	0.00043
1	Θ_2	0.00013	0.00027	0.00043	0.00053	0.00080	0.00050	0.00047
1	Θ_3^2	0.00000	0.00007	0.00023	0.00027	0.00053	0.00030	0.00027
1	M _{2->1}	712.667	912.667	916.333	916.667	999.333	587.667	563.304
1	M _{3->1}	287.333	289.333	290.333	290.667	290.667	519.667	511.429
1	M _{1->2}	514.000	516.000	517.667	518.000	525.333	527.000	519.406
1	M _{3->2}	58.667	71.333	73.000	73.333	74.667	500.333	499.870
1	M _{1->3}	466.667	496.000	497.000	497.333	511.333	557.000	538.305
1	M _{2->3}	870.667	885.333	887.000	888.000	964.667	592.333	565.886
2	Θ_1	0.00027	0.00040	0.00057	0.00073	0.00113	0.00070	0.00068
2	Θ_2	0.00040	0.00060	0.00083	0.00100	0.00153	0.00097	0.00095
2	Θ_3^2	0.00027	0.00040	0.00063	0.00073	0.00120	0.00070	0.00070
2	M _{2->1}	764.000	840.000	841.667	842.000	854.667	561.667	545.810
2	M _{3->1}	497.333	873.333	875.000	875.333	999.333	633.000	600.932
2	M _{1->2}	706.667	754.000	757.000	759.333	901.333	581.000	561.927
2	M _{3->2}	864.000	870.000	871.000	873.333	880.667	525.000	517.441
2	M _{1->3}	632.667	964.667	968.333	971.333	999.333	669.667	630.953
2	M _{2->3}	434.000	958.667	988.333	999.333	999.333	759.000	715.309
All	Θ_1	0.00000	0.00000	0.00057	0.00107	0.00227	0.00103	0.00104
All	Θ_2	0.00000	0.00013	0.00070	0.00127	0.00233	0.00110	0.00110
All	Θ_3	0.00000	0.00000	0.00050	0.00093	0.00220	0.00097	0.00099
All	M _{2->1}	50.000	333.333	475.000	484.667	498.667	313.667	297.203
All	M _{3->1}	59.333	390.667	475.667	486.667	498.667	313.667	296.458
All	M _{1->2}	46.667	301.333	475.000	487.333	497.333	311.000	292.104
All	M _{3->2}	56.000	264.000	415.667	481.333	486.667	266.333	259.380
All	M _{1->3}	71.333	328.667	475.000	488.000	501.333	339.000	315.769
All	M _{2->3}	145.333	380.667	475.667	489.333	504.667	388.333	361.413

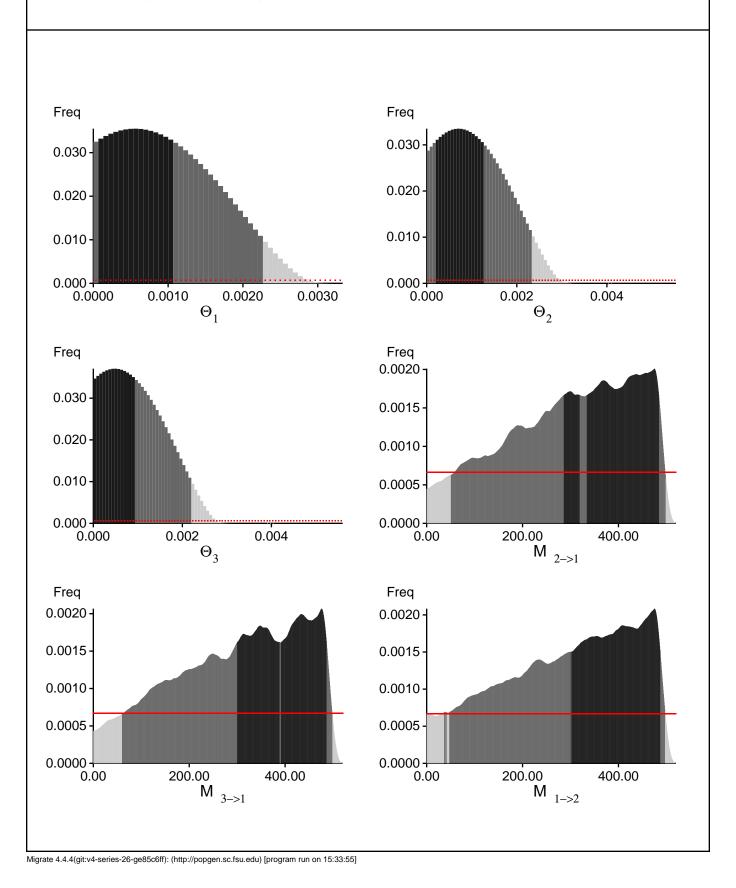
Citation suggestions:

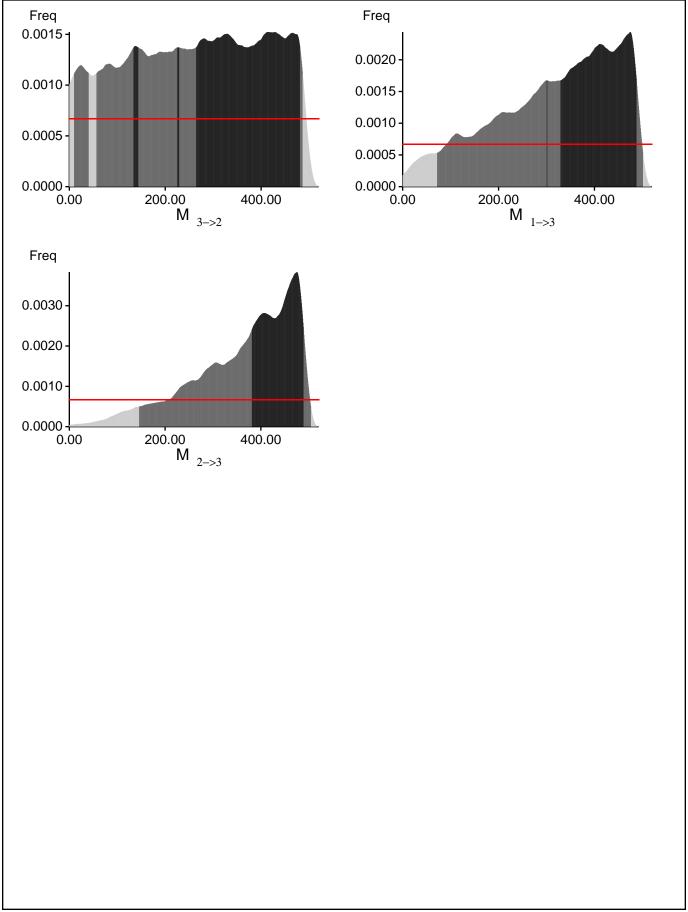
Beerli P., 2006. Comparison of Bayesian and maximum-likelihood inference of population genetic parameters. Bioinformatics 22:341-345

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 15:33:55]

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	-1687.81	-1658.70	-1672.85
2	-1622.53	-1549.80	-1527.00
All	-3326.56	-3224.72	-3216.07

(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 = -16.220510]

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	942201/2221429	0.42414
Θ_2	1117549/2221764	0.50300
Θ_3	915119/2221207	0.41199
$M_{2\rightarrow 1}$	1926656/2223810	0.86638
$M_{3->1}$	1922633/2223226	0.86479
$M_{1\rightarrow 2}$	1910856/2221456	0.86018
$M_{3->2}$	1912873/2222529	0.86067
$M_{1\rightarrow 3}$	1891384/2221519	0.85139
$M_{2\rightarrow3}$	1779411/2221833	0.80088
Genealogies	8588861/20001227	0.42942

MCMC-Autocorrelation and Effective MCMC Sample Size

Parameter	Autocorrelation	Effective Sampe Size
Θ_1	0.55441	12114.68
$\Theta_2^{'}$	0.50076	14514.04
Θ_3^2	0.56839	12055.91
$M_{2->1}$	0.05996	35498.86
$M_{3->1}^{2}$	0.05899	35660.55
$M_{1->2}$	0.07997	34195.63
$M_{3->2}$	0.09806	33156.24
$M_{1->3}^{3>2}$	0.07789	34313.81
$M_{2->3}$	0.05366	35960.52
Genealogies	0.32157	21250.11

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.

Param 4 (Locus 1): Upper prior boundary seems too low! Param 8 (Locus 2): Upper prior boundary seems too low! Param 9 (Locus 2): Upper prior boundary seems too low!