## Catostomus ardens ND2

POPULATION SIZE, MIGRATION, DIVERGENCE, ASSIGNMENT, HISTORY

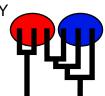
Bayesian inference using the structured coalescent

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

Compiled for a SYMMETRIC multiprocessors (Grandcentral)

Program started at Thu May 20 03:38:25 2021

Program finished at Thu May 20 07:56:27 2021 [Runtime:0000:04:18:02]



### **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) 813498322

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 1 Great\_Salt\_Lake \* D 2 Sevier\_Desert \* \*

#### Order of parameters:

1	$\Theta_1$	<displayed></displayed>
2	$\Theta_2^{-}$	<displayed></displayed>
3	$M_{2->1}^{2}$	<displayed></displayed>
4	$M_{1\rightarrow 2}$	<displayed></displayed>

						Odiosionias	arachi NDZ Z
5 Δ	2->1		<displa< td=""><td>yed&gt;</td><td></td><td></td><td></td></displa<>	yed>			
6 σ			<displa< td=""><td>ved&gt;</td><td></td><td></td><td></td></displa<>	ved>			
	2->1		Laiopia	your			
Mutation rate among	loci:					Mutation ra	ate is constant
Analysis strategy:						-	esian inference
-Population size esti						•	tial Distribution
-Geneflow estimation			No	rmal Diatrib	ution Charton		tial Distribution
-Divergence time es	umauon.		INO	חוומו טואנווט	ulion Shortet	it (illeali allu	standard dev.)
Proposal distributions	s for paramete	er					
Parameter	'		osal				
Theta	M	letropolis sam	pling				
M	M	letropolis sam	pling				
Divergence	M	letropolis sam	pling				
Divergence Spread	M	letropolis sam	pling				
Genealogy	N	1etropolis-Has	tings				
Prior distribution for p		Minimo	N/*//-		Dalta	Dina I	la data Casa
Parameter  1 Theta *	Prior * Uniform	Minimum 0.000000	MeanMa 0.050	0.100	Delta 0.010	1500	JpdateFreq 0.08333
1 Theta * 2 Theta *		0.000000	0.050	0.100	0.010	1500	0.08333
3 M *		0.000000	500.0	1000.	100.0	1500	0.08333
4 M *		0.000000	500.0	1000.	100.0	1500	0.08333
5 Splittime mean *	* Uniform	0.000000	0.250	0.500	0.050	1500	0.08333
6 Splittime std *	* Uniform	0.000000	0.250	0.500	0.050	1500	0.08333
[* * means priors wer	e set globally	]					
Markov chain setting	S:						Long chain
Number of chains							1
Recorded steps [a]		L1					10000
Increment (record		-					1000
Number of concurr Visited (sampled) p	,	. ,					2 20000000
Number of discard							1000
ramber of discard	tices per che	iii (baiii iii)					1000
Multiple Markov chair	ns:						
Static heating sche	eme					4 chains with	temperatures
				1000	00.000	3.00	1.50 1.00
						Swappir	ng 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
	1

### Data summary

Data file: infile

Datatype: Haplotype data

Number of loci:

Mutationmodel:

Locus Sublocus Mutationmodel Mutationmodel parameters

1 1 Kimura [Basefreq: =0.25, kappa=3.7400]

Sites per locus

Locus Sites

1 400

Site rate variation and probabilities:

Locus Sublocus Region type Rate of change Probability Patch size

1	1	1	1.000	1.000	1.000		
Populat	tion				Locus	Gene co	pies
						data	(missing)
1 Great	t_Salt_La	ıke			1	81	
2 Sevie	r_Desert				1	19	
Total of	f all popu	lations			1	100	(0)

## Bayesian Analysis: Posterior distribution table

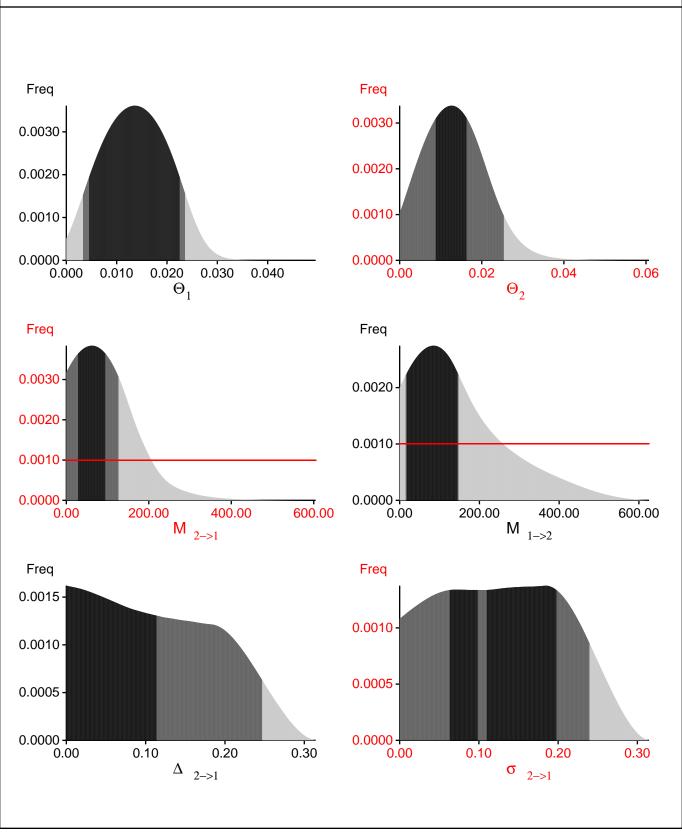
Locus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
1	$\Theta_1$	0.00327	0.00447	0.01363	0.02253	0.02360	0.01377	0.01362
1	$\Theta_2$	0.00000	0.00873	0.01263	0.01633	0.02540	0.01323	0.01305
1	M <sub>2-&gt;1</sub>	0.000	28.000	62.333	94.667	126.667	91.667	81.958
1	M <sub>1-&gt;2</sub>	14.667	17.333	85.000	145.333	148.000	131.667	154.177
1	D <sub>2-&gt;1</sub>	0.00000	0.00000	0.00017	0.11400	0.24700	0.11417	0.23303
1	S <sub>2-&gt;1</sub>	0.00000	0.10967	0.18450	0.19800	0.23967	0.12983	0.25475

#### 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 for locus 1



Catostomus ardens ND2 7

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

Use this value for Bayes factor calculations:

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

Method	In(Prob(D Model))	Notes
Thermodynamic integration	-1528.111410	(1a)
	-998.893218	(1b)
Harmonic mean	-884.969496	(2)

(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

#### 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
$\Theta_1$	709623/1667626	0.42553
$\Theta_2$	486649/1666205	0.29207
$M_{2\rightarrow 1}$	1145483/1665952	0.68758
$M_{1\rightarrow 2}$	736510/1664963	0.44236
$\Delta$ $2\rightarrow 1$	1532041/1665058	0.92011
$\sigma_{2\rightarrow 1}$	1560628/1668736	0.93522
Genealogies	3095279/10001460	0.30948

# MCMC-Autocorrelation and Effective MCMC Sample Size

Parameter	Autocorrelation	Effective Sampe Size
$\Theta_1$	0.48947	6877.85
$\Theta_2$	0.38526	8880.14
$M^{2}$	0.74738	3367.83
$M_{1->2}$	0.61396	4785.97
$\Delta$ $2\rightarrow 1$	0.06001	17799.83
$\sigma_{2\rightarrow 1}$	0.01057	19590.13
Genealogies	0.01057	19590.13

#### 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.
not help in such situations, reducing number of parameters may help in such situations.
No warning was recorded during the run