AUTO

POPULATION SIZE, MIGRATION, DIVERGENCE, ASSIGNMENT, HISTORY

Bayesian inference using the structured coalescent

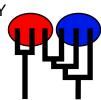
Migrate-n version 5.0.0a [May-20-2017]

Compiled for PARALLEL computer architectures

One master and 20 compute nodes are available.

Program started at Wed Dec 20 22:43:49 2017

Program finished at Wed Dec 20 22:59:19 2017 [Runtime:0000:00:15:30]



Options

Datatype: DNA sequence data

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) 3416088093

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 2 1 Romanshorn_0 * * 2 Arbon_1 * *

Order of parameters:

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

								AUTO 2
3	M 2-	->1		<displa< td=""><td>yed></td><td></td><td></td><td></td></displa<>	yed>			
4		->2		<displa< td=""><td>yed></td><td></td><td></td><td></td></displa<>	yed>			
Mutation	rate among lo	ci:				Mutatio	on rate is con	stant for all loci
Analysis	strategy:						Raye	esian inference
_	ion size estima	ation:						tial Distribution
	w estimation:	dio11.						tial Distribution
Dranasal	diatributiana fe							
Proposal	distributions fo	or paramete		a a a a l				
Theta	ſ	Λ.	•	oosal				
meta M			letropolis sam					
			1etropolis sam 1etropolis sam					
Divergenc Divergenc			•					
Genealogy	•		∕letropolis sam ⁄letropolis-Has	-				
Genealogy	у	IV	netropolis-i ias	ungs				
Prior dist	ribution for par	ameter						
Parameter	r	Prior	Minimum	MeanM	aximum	Delta	Bins U	JpdateFreq
1	Theta -11	Uniform	0.000000	0.100	0.200	0.010	1500	0.05000
2	Theta -11	Uniform	0.000000	0.100	0.200	0.010	1500	0.05000
3	M -11	Uniform	0.000000	500.0	1000.	100.0	1500	0.05000
4	M -11	Uniform	0.000000	500.0	1000.	100.0	1500	0.05000
[-1 -1 me	ans priors wer	e set globa	lly]					
Markov c	hain settings:							Long chain
Number	of chains							1
Record	ded steps [a]							10000
Increm	ent (record ev	ery x step [b]					100
Numbe	er of concurren	t chains (re	eplicates) [c]					2
Visited	(sampled) par	ameter val	ues [a*b*c]					2000000
Numbe	er of discard tre	ees per cha	in (burn-in)					5000
Multiple N	Markov chains:							
Static h	neating schem	е					4 chains with	temperatures
					1000	00.000	3.00	1.50 1.00
							Swappii	ng interval is 1
Print opti	ons:							
Data fil								infile.1.0
Haplot	yping is turned	on:						NO
Output							0	utfile_1.0_1.0s
-	ior distribution	raw histogr	ram file:					bayesfile
	ata from the M	_					bayesa	allfile_1.0_1.0s
							•	

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

Data summary

Data file: infile.1.0

Datatype: Sequence data

Number of loci: 10

Mutationmodel:

Locus S	ublocus	Mutationmodel	Mutationmodel parameters
1	1	Felsenstein 84	[Bf:0.25 0.24 0.25 0.25, t/t ratio=2.000]
2	1	Felsenstein 84	[Bf:0.25 0.25 0.25 0.25, t/t ratio=2.000]
3	1	Felsenstein 84	[Bf:0.25 0.25 0.25 0.25, t/t ratio=2.000]
4	1	Felsenstein 84	[Bf:0.25 0.25 0.25 0.25, t/t ratio=2.000]
5	1	Felsenstein 84	[Bf:0.24 0.26 0.25 0.25, t/t ratio=2.000]
6	1	Felsenstein 84	[Bf:0.25 0.25 0.25 0.25, t/t ratio=2.000]
7	1	Felsenstein 84	[Bf:0.25 0.26 0.25 0.25, t/t ratio=2.000]
8	1	Felsenstein 84	[Bf:0.25 0.25 0.25 0.25, t/t ratio=2.000]
9	1	Felsenstein 84	[Bf:0.25 0.25 0.25 0.25, t/t ratio=2.000]
10	1	Felsenstein 84	[Bf:0.25 0.25 0.25 0.25, t/t ratio=2.000]

Sites per locus

Locus	Sites
1	10000
2	10000
3	10000
4	10000
5	10000
6	10000
7	10000
8	10000
9	10000
10	10000

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
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
6	1	1	1.000	1.000	1.000

	4	4	1.000	4.000	4.000	
7	1	1	1.000	1.000	1.000	
8	1	1	1.000	1.000	1.000	
9	1	1	1.000	1.000	1.000	
10	1	1	1.000	1.000	1.000	0
Popula					Locus	Gene copies
1 Rom	anshorn_	0			1	10
					2	10
					3	10
					4	10
					5	10
					6	10
					7	10
					8	10
					9	10
					10	10
2 Arbo	n_1				1	10
					2	10
					3	10
					4	10
					5	10
					6	10
					7	10
					8	10
					9	10
					10	10
Total o	f all popu	lations			1	20
					2	20
					3	20
					4	20
					5	20
					6	20
					7	20
					8	20
					9	20
					10	20

Bayesian Analysis: Posterior distribution table

Locus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
1	Θ_1	0.00000	0.00520	0.00913	0.01440	0.03760	0.01233	0.01505
1	Θ_2	0.00293	0.00840	0.01207	0.01587	0.02600	0.01327	0.01394
1	M _{2->1}	301.3	686.7	959.7	986.0	999.3	697.7	675.5
1	M _{1->2}	0.0	0.0	0.3	207.3	849.3	227.0	314.9
2	Θ_1	0.00320	0.00960	0.01407	0.02000	0.05147	0.01753	0.02173
2	Θ_2	0.00013	0.00400	0.00673	0.00920	0.01507	0.00727	0.00765
2	M _{2->1}	0.0	0.0	109.7	404.7	729.3	211.7	268.5
2	M _{1->2}	0.0	0.0	0.3	185.3	748.7	185.7	258.2
3	Θ_1	0.00000	0.00387	0.00647	0.00880	0.01400	0.00687	0.00707
3	Θ_2	0.00240	0.00453	0.00967	0.01667	0.02213	0.01140	0.01298
3	M _{2->1}	0.0	7.3	152.3	243.3	708.0	237.0	276.8
3	M _{1->2}	0.0	5.3	49.0	108.0	857.3	247.0	335.2
4	Θ_1	0.00080	0.00587	0.00927	0.01333	0.02627	0.01113	0.01263
4	Θ_2	0.00253	0.00933	0.01327	0.01773	0.03680	0.01527	0.01810
4	M _{2->1}	0.0	0.0	0.3	332.7	834.7	343.7	390.2
4	M _{1->2}	0.0	0.0	51.0	127.3	398.0	156.3	284.3
5	Θ_1	0.00120	0.00560	0.00847	0.01147	0.01960	0.00940	0.00996
5	Θ_2	0.00213	0.00667	0.00980	0.01307	0.02227	0.01087	0.01157
5	M _{2->1}	0.0	0.0	0.3	62.7	444.7	63.0	111.7
5	M _{1->2}	0.0	0.0	0.3	54.0	239.3	54.3	80.4
6	Θ_1	0.00000	0.00333	0.00580	0.00813	0.01267	0.00633	0.00633
6	Θ_2	0.00307	0.00973	0.01713	0.02813	0.09933	0.02780	0.04356
6	M _{2->1}	0.0	0.0	60.3	302.0	726.0	259.7	307.3
6	M _{1->2}	160.7	428.7	611.0	756.7	998.7	590.3	580.9
7	Θ_1	0.00000	0.00520	0.00993	0.01733	0.06280	0.01513	0.02265
7	Θ_2	0.00027	0.00667	0.01207	0.02133	0.09467	0.01900	0.03499
7	M _{2->1}	176.7	486.7	509.0	546.7	992.0	551.7	549.9
7	M _{1->2}	107.3	255.3	393.7	657.3	949.3	509.7	514.6
8	Θ_1	0.00547	0.01013	0.01447	0.01960	0.03000	0.01673	0.01832

Migrate 5.0.0a: (http://popgen.sc.fsu.edu) [program run on 22:43:49]

Locus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
8	Θ_2	0.00000	0.00240	0.00487	0.00707	0.01320	0.00553	0.00575
8	M _{2->1}	0.0	39.3	159.0	226.7	676.0	201.0	254.8
8	M _{1->2}	115.3	217.3	366.3	594.0	964.7	493.0	514.4
9	Θ_1	0.00640	0.01613	0.01753	0.01907	0.04107	0.02060	0.02245
9	Θ_2	0.00160	0.00640	0.00980	0.01400	0.02627	0.01167	0.01285
9	M _{2->1}	0.0	0.0	0.3	91.3	652.7	91.7	172.1
9	M _{1->2}	0.0	35.3	125.7	308.7	675.3	213.0	264.1
10	Θ_1	0.00800	0.01547	0.02100	0.02760	0.04760	0.02393	0.02570
10	Θ_2	0.00027	0.00427	0.00713	0.00987	0.01680	0.00793	0.00835
10	M _{2->1}	10.0	28.7	50.3	75.3	109.3	73.7	91.3
10	M _{1->2}	2.0	52.0	66.3	89.3	364.7	142.3	185.5
All	Θ_1	0.00520	0.00827	0.01047	0.01240	0.01560	0.01060	0.01049
All	Θ_2	0.00413	0.00707	0.00913	0.01107	0.01413	0.00927	0.00919
All	M _{2->1}	0.0	0.0	9.7	35.3	105.3	35.7	40.5
All	M _{1->2}	0.0	0.0	19.0	47.3	59.3	53.0	59.8

Citation suggestions:

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

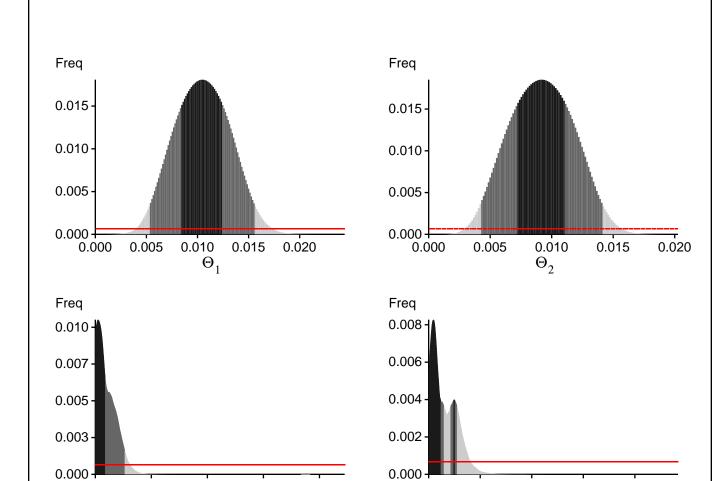
Beerli P., 2007. Estimation of the population scaled mutation rate from microsatellite data, Genetics, 177:1967-1968.

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.

200.00 400.00 600.00 800.00

M _{1->2}

Bayesian Analysis: Posterior distribution over all loci



200.00

0.00

400.00 600.00

M _{2->1}

800.00

0.00

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]

Locus	TI(1a)	BTI(1b)	SS(2)	HS(3)
1	-21820.20	-19431.70	-18519.82	-19173.67
2	-20197.15	-18479.06	-nan	-18357.20
3	-19169.05	-17520.06	-nan	-17397.14
4	-23080.45	-20084.13	-nan	-19705.41
5	-22375.10	-19554.71	-nan	-19230.35
6	-20647.53	-18349.15	-nan	-18099.94
7	-25069.22	-21155.57	-nan	-20602.10
8	-24282.50	-20941.68	-nan	-20504.70
9	-24605.03	-21541.55	-nan	-21155.78
10	-28607.09	-23343.85	-nan	-22539.36
All	-229823.79	-200371.94	-nan	-196736.13

- (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!
- (2) SS: Steppingstone Sampling (Xie et al 2011)
- (3) HS: Harmonic mean approximation: Overestimates the marginal likelihood, poor variance [Scaling factor = 29.525617]

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.

Xie W., P. O. Lewis, Y. Fan, L. Kuo, and M.-H. Chen. 2011. Improving marginal likelihood estimation for Bayesian phylogenetic model selection. Systematic Biology, 60(2):150â 160, 2011.

Acceptance ratios for all parameters and the genealogies

Parameter	Accepted changes	Ratio
Θ_1	729870/999071	0.73055
Θ_2	710241/1000569	0.70984
$M_{2\rightarrow 1}$	827778/1000172	0.82764
$M_{1\rightarrow 2}$	852198/999361	0.85274
Genealogies	520196/16000827	0.03251

Average temperatures during the run

Chain Temperatures

- 1 0.00000
- 2 0.00000
- 3 0.00000
- 4 0.00000

Adaptive heating often fails, if the average temperatures are very close together try to rerun using static heating! If you want to compare models using marginal likelihoods then you MUST use static heating

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 reducing number of parameters may help in such situations.

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, Param 3 (Locus 1): Upper prior boundary seems too low!