### **AUTO**

POPULATION SIZE, MIGRATION, DIVERGENCE, ASSIGNMENT, HISTORY

Bayesian inference using the structured coalescent

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

Using Intel AVX (Advanced Vector Extensions)

Compiled for PARALLEL computer architectures

One master and 40 compute nodes are available.

Program started at Mon Jul 24 14:46:05 2017

Program finished at Mon Jul 24 18:31:12 2017 [Runtime:0000:03:45:07]



### **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) 2422552244

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

1

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

Population

1 Romanshorn 0

Order of parameters:

1  $\Theta_1$  <displayed>

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

Analysis strategy: Bayesian inference

**Exponential Distribution** -Population size estimation:

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

Parameter Delta Prior Minimum Mean Maximum Bins UpdateFreq Theta -11 Uniform 0.000000 0.050 0.100 0.010 1500 0.20000

[-1 -1 means priors were set globally]

Markov chain settings: Long chain

Number of chains 50000 Recorded steps [a] 200 Increment (record every x step [b] Number of concurrent chains (replicates) [c]

20000000 Visited (sampled) parameter values [a\*b\*c] 10000 Number of discard trees per chain (burn-in)

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.0.8 NO

Haplotyping is turned on:

Output file: outfile\_0.8\_0.4

Posterior distribution raw histogram file: bayesfile

bayesallfile\_0.8\_0.4 Print data: No

Print genealogies [only some for some data type]: None

Raw data from the MCMC run:

## Data summary

Data file: infile.0.8
Datatype: Sequence data
Number of loci: 100

Mutationmodel: Locus Sublocus			Mutationmodel	Mutationmodel parameters
Locus Sublocus		abiocas	Watationinoaci	Watatorimoder parameters
	1	1	Jukes-Cantor	[Basefreq: =0.25]
	2	1	Jukes-Cantor	[Basefreq: =0.25]
	3	1	Jukes-Cantor	[Basefreq: =0.25]
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	32	1	Jukes-Cantor	[Basefreq: =0.25]

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Jukes-Cantor

Jukes-Cantor

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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	
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12	1	1	1.000	1.000	1.000	
13	1	1	1.000	1.000	1.000	
14	1	1	1.000	1.000	1.000	
15	1	1	1.000	1.000	1.000	
16	1	1	1.000	1.000	1.000	
17	1	1	1.000	1.000	1.000	
18	1	1	1.000	1.000	1.000	
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28	1	1	1.000	1.000	1.000	
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31	1	1	1.000	1.000	1.000	
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42	1	1	1.000	1.000	1.000	
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65	1	1	1.000	1.000	1.000	
66	1	1	1.000	1.000	1.000	
67	1	1	1.000	1.000	1.000	
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69	1	1	1.000	1.000	1.000	
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# Bayesian Analysis: Posterior distribution table

Locus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
1	$\Theta_1$	0.03433	0.04433	0.04790	0.04967	0.05160	0.04557	0.08720
2	$\Theta_1$	0.03093	0.04347	0.04797	0.04993	0.05160	0.04463	0.08269
3	$\Theta_1$	0.03340	0.04433	0.04797	0.04980	0.05167	0.04550	0.08549
4	$\Theta_1$	0.02960	0.04427	0.04783	0.04980	0.05193	0.04543	0.08553
5	$\Theta_1$	0.03353	0.04427	0.04783	0.04973	0.05153	0.04543	0.08483
6	$\Theta_1$	0.03233	0.04393	0.04783	0.04973	0.05160	0.04517	0.08542
7	$\Theta_1$	0.03067	0.04293	0.04790	0.04967	0.05160	0.04423	0.08217
8	$\Theta_1$	0.03480	0.04500	0.04797	0.04993	0.05167	0.04603	0.08647
9	$\Theta_1$	0.03353	0.04440	0.04797	0.04973	0.05153	0.04563	0.08652
10	$\Theta_1$	0.03373	0.04453	0.04790	0.04987	0.05153	0.04563	0.08761
11	$\Theta_1$	0.03507	0.04493	0.04810	0.05000	0.05173	0.04597	0.08741
12	$\Theta_1$	0.03260	0.04387	0.04777	0.04967	0.05160	0.04510	0.08434
13	$\Theta_1$	0.03267	0.04440	0.04810	0.05000	0.05167	0.04550	0.08596
14	$\Theta_1$	0.03407	0.04427	0.04803	0.04987	0.05160	0.04543	0.08556
15	$\Theta_1$	0.03220	0.04407	0.04777	0.04953	0.05167	0.04537	0.08459
16	$\Theta_1$	0.03333	0.04433	0.04797	0.04980	0.05167	0.04557	0.08484
17	$\Theta_1$	0.03393	0.04447	0.04797	0.04987	0.05153	0.04557	0.08624
18	$\Theta_1$	0.03287	0.04513	0.04797	0.04960	0.05167	0.04537	0.08500

19	$\Theta_1$	0.03420	0.04453	0.04783	0.04980	0.05160	0.04570	0.08546
20	$\Theta_1$	0.03160	0.04407	0.04790	0.04980	0.05160	0.04523	0.08444
21	$\Theta_1$	0.03253	0.04400	0.04803	0.04980	0.05160	0.04517	0.08577
22	$\Theta_1$	0.03367	0.04460	0.04777	0.04967	0.05160	0.04577	0.08690
23	$\Theta_1$	0.03400	0.04453	0.04797	0.04987	0.05167	0.04570	0.08569
24	$\Theta_1$	0.03373	0.04473	0.04803	0.05000	0.05160	0.04577	0.08562
25	$\Theta_1$	0.03313	0.04400	0.04783	0.04967	0.05160	0.04530	0.08470
26	$\Theta_1$	0.03327	0.04447	0.04803	0.04987	0.05167	0.04557	0.08560
27	$\Theta_1$	0.03513	0.04560	0.04797	0.04960	0.05167	0.04577	0.08686
28	$\Theta_1$	0.03447	0.04473	0.04803	0.05000	0.05173	0.04577	0.08799
29	$\Theta_1$	0.03367	0.04440	0.04797	0.04980	0.05167	0.04557	0.08621
30	$\Theta_1$	0.03367	0.04413	0.04790	0.04973	0.05180	0.04530	0.08535
31	$\Theta_1$	0.03373	0.04433	0.04790	0.04987	0.05167	0.04543	0.08651
32	$\Theta_1$	0.03493	0.03973	0.04810	0.05113	0.05180	0.04577	0.08680
33	$\Theta_1$	0.03587	0.04480	0.04797	0.04980	0.05153	0.04597	0.08800
34	$\Theta_1$	0.03280	0.04360	0.04770	0.04953	0.05167	0.04490	0.08524
35	$\Theta_1$	0.03147	0.04347	0.04790	0.04973	0.05160	0.04470	0.08377
36	$\Theta_1$	0.03160	0.04393	0.04770	0.04967	0.05160	0.04510	0.08434
37	$\Theta_1$	0.03413	0.04453	0.04783	0.04960	0.05180	0.04577	0.08654
38	$\Theta_1$	0.03260	0.04460	0.04790	0.04953	0.05160	0.04490	0.08383
39	$\Theta_1$	0.03267	0.04413	0.04797	0.04980	0.05160	0.04530	0.08365
40	$\Theta_1$	0.03293	0.04527	0.04797	0.04960	0.05160	0.04543	0.08552
41	$\Theta_1$	0.03500	0.04480	0.04790	0.04980	0.05153	0.04597	0.08774

Migrate 5.0.0a: (http://popgen.sc.fsu.edu) [program run on 14:46:05]

Locus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
42	$\Theta_1$	0.03300	0.04487	0.04783	0.04947	0.05167	0.04503	0.08341
43	$\Theta_1$	0.03320	0.04527	0.04783	0.04953	0.05153	0.04543	0.08560
44	$\Theta_1$	0.03240	0.04413	0.04783	0.04987	0.05160	0.04530	0.08468
45	$\Theta_1$	0.03313	0.04427	0.04777	0.04973	0.05167	0.04543	0.08417
46	$\Theta_1$	0.03153	0.04360	0.04777	0.04973	0.05160	0.04483	0.08260
47	$\Theta_1$	0.03387	0.04473	0.04810	0.05020	0.05173	0.04577	0.08713
48	$\Theta_1$	0.03207	0.04387	0.04797	0.04987	0.05167	0.04503	0.08412
49	$\Theta_1$	0.03473	0.04487	0.04797	0.04987	0.05173	0.04597	0.08688
50	$\Theta_1$	0.03407	0.04433	0.04797	0.04993	0.05153	0.04543	0.08585
51	$\Theta_1$	0.03153	0.04360	0.04797	0.04980	0.05167	0.04483	0.08419
52	$\Theta_1$	0.03240	0.04393	0.04777	0.04960	0.05160	0.04517	0.08415
53	$\Theta_1$	0.03353	0.04433	0.04803	0.04993	0.05167	0.04550	0.08611
54	$\Theta_1$	0.03240	0.04453	0.04790	0.04980	0.05180	0.04563	0.08627
55	$\Theta_1$	0.03373	0.04427	0.04790	0.04980	0.05173	0.04543	0.08428
56	$\Theta_1$	0.03333	0.04400	0.04777	0.04967	0.05160	0.04523	0.08482
57	$\Theta_1$	0.03207	0.04387	0.04790	0.04987	0.05160	0.04503	0.08340
58	$\Theta_1$	0.03347	0.04520	0.04803	0.04960	0.05167	0.04537	0.08540
59	$\Theta_1$	0.03467	0.04613	0.04803	0.04967	0.05187	0.04630	0.08793
60	$\Theta_1$	0.03500	0.04553	0.04777	0.04913	0.05153	0.04570	0.08639
61	$\Theta_1$	0.03527	0.04513	0.04803	0.04980	0.05173	0.04623	0.08835

62	$\Theta_1$	0.03453	0.04467	0.04797	0.04993	0.05167	0.04577	0.08601
63	$\Theta_1$	0.03347	0.04387	0.04790	0.04980	0.05173	0.04510	0.08630
64	$\Theta_1$	0.03340	0.04413	0.04797	0.04980	0.05160	0.04530	0.08484
65	$\Theta_1$	0.03540	0.04487	0.04797	0.04980	0.05173	0.04610	0.08772
66	$\Theta_1$	0.03320	0.04400	0.04797	0.04987	0.05167	0.04517	0.08441
67	$\Theta_1$	0.03340	0.04447	0.04797	0.04993	0.05173	0.04557	0.08617
68	$\Theta_1$	0.03120	0.04353	0.04783	0.04967	0.05160	0.04483	0.08362
69	$\Theta_1$	0.03400	0.04473	0.04797	0.04987	0.05167	0.04590	0.08605
70	$\Theta_1$	0.03367	0.04393	0.04783	0.04953	0.05160	0.04497	0.08546
71	$\Theta_1$	0.03387	0.04460	0.04797	0.04987	0.05187	0.04577	0.08589
72	$\Theta_1$	0.03180	0.04360	0.04777	0.04960	0.05160	0.04483	0.08326
73	$\Theta_1$	0.03367	0.04520	0.04797	0.04960	0.05173	0.04537	0.08567
74	$\Theta_1$	0.03493	0.04453	0.04790	0.04967	0.05167	0.04583	0.08682
75	$\Theta_1$	0.03360	0.04440	0.04797	0.04980	0.05173	0.04557	0.08737
76	$\Theta_1$	0.03280	0.04373	0.04797	0.04973	0.05160	0.04497	0.08473
77	$\Theta_1$	0.03300	0.04433	0.04790	0.04987	0.05167	0.04543	0.08464
78	$\Theta_1$	0.03167	0.04480	0.04777	0.04953	0.05153	0.04497	0.08368
79	$\Theta_1$	0.03460	0.04607	0.04810	0.04967	0.05160	0.04597	0.08929
80	$\Theta_1$	0.03393	0.04487	0.04817	0.05007	0.05167	0.04597	0.08783
81	$\Theta_1$	0.03320	0.04427	0.04797	0.04993	0.05167	0.04537	0.08564
82	$\Theta_1$	0.03293	0.04413	0.04790	0.04967	0.05160	0.04537	0.08653
83	$\Theta_1$	0.03300	0.04507	0.04777	0.04953	0.05167	0.04523	0.08440
84	$\Theta_1$	0.03667	0.04573	0.04790	0.04947	0.05167	0.04597	0.08816

Locus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
85	$\Theta_1$	0.03287	0.04480	0.04797	0.05000	0.05173	0.04590	0.08574
86	$\Theta_1$	0.03353	0.04407	0.04763	0.04953	0.05153	0.04537	0.08632
87	$\Theta_1$	0.03353	0.04433	0.04777	0.04973	0.05153	0.04550	0.08505
88	$\Theta_1$	0.03213	0.04400	0.04783	0.04987	0.05153	0.04510	0.08470
89	$\Theta_1$	0.03320	0.04440	0.04797	0.04980	0.05173	0.04557	0.08633
90	$\Theta_1$	0.03280	0.04433	0.04803	0.05007	0.05180	0.04537	0.08590
91	$\Theta_1$	0.03387	0.04447	0.04797	0.04980	0.05160	0.04563	0.08694
92	$\Theta_1$	0.03380	0.04460	0.04803	0.05000	0.05173	0.04570	0.08563
93	$\Theta_1$	0.03293	0.04400	0.04790	0.04980	0.05167	0.04523	0.08347
94	$\Theta_1$	0.03453	0.04447	0.04797	0.04967	0.05167	0.04570	0.08654
95	$\Theta_1$	0.03340	0.04433	0.04790	0.04980	0.05173	0.04550	0.08647
96	$\Theta_1$	0.03227	0.04353	0.04777	0.04953	0.05153	0.04483	0.08364
97	$\Theta_1$	0.03327	0.04433	0.04783	0.04980	0.05160	0.04550	0.08617
98	$\Theta_1$	0.03293	0.04453	0.04803	0.05007	0.05173	0.04557	0.08500
99	$\Theta_1$	0.03293	0.04527	0.04790	0.04947	0.05160	0.04543	0.08525
100	$\Theta_1$	0.03260	0.04407	0.04797	0.04980	0.05167	0.04523	0.08501
All	$\Theta_1$	0.01293	0.01587	0.01730	0.01887	0.02100	0.01723	0.09977

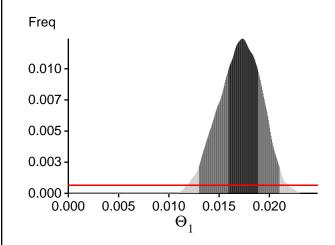
#### Citation suggestions:

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.					

## Bayesian Analysis: Posterior distribution over all loci



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

Use this value for Bayes factor calculations:  $BF = Exp[\ ln(Prob(D \mid thisModel) - ln(\ Prob(\ D \mid otherModel)) \\ or \ as \ LBF = 2 \ (ln(Prob(D \mid thisModel) - ln(\ Prob(\ D \mid otherModel))) \\ shows the \ support for \ thisModel]$ 

Locus	TI(1a)	BTI(1b)	SS(2)	HS(3)
1	-15065.86	-14556.54	-14562.53	-14629.90
2	-14100.33	-13945.26	-13984.21	-14077.61
3	-15016.52	-14756.18	-14779.01	-14872.02
4	-14346.83	-14128.66	-14182.74	-14254.08
5	-14521.55	-14266.89	-14311.39	-14388.00
6	-14191.35	-14002.12	-14051.84	-14127.95
7	-14061.31	-13912.50	-13956.06	-14048.53
8	-14452.82	-14173.55	-14213.21	-14287.29
9	-14686.68	-14344.44	-14375.41	-14449.26
10	-43248.03	-29033.45	-26580.28	-26638.52
11	-16336.77	-15408.02	-15350.92	-15413.68
12	-14350.39	-14140.19	-14190.51	-14268.09
13	-14409.64	-14138.21	-14176.60	-14252.67
14	-14627.14	-14351.00	-14389.91	-14465.67
15	-14260.48	-14027.42	-14065.44	-14148.09
16	-15874.24	-15096.34	-15051.07	-15125.97
17	-14606.49	-14321.28	-14362.48	-14437.53
18	-14163.52	-13966.43	-14016.88	-14093.33
19	-14919.63	-14486.84	-14505.36	-14576.70
20	-14591.41	-14251.56	-14279.36	-14357.16
21	-14265.45	-14074.52	-14128.59	-14203.28
22	-14669.70	-14303.80	-14330.71	-14402.36
23	-23532.94	-21357.95	-21057.81	-21150.93
24	-14805.68	-14387.76	-14404.50	-14478.86
25	-14257.79	-14010.06	-14047.83	-14127.78
26	-14395.89	-14158.94	-14196.97	-14276.76
27	-17475.38	-15856.11	-15652.87	-15732.33
28	-16072.68	-15348.52	-15324.64	-15388.72
29	-15380.95	-14628.72	-14580.66	-14656.47

Migrate 5.0.0a: (http://popgen.sc.fsu.edu) [program run on 14:46:05]

30	-14410.68	-14200.18	-14250.93	-14327.22
31	-15055.72	-14652.24	-14681.73	-14747.14
32	-15704.85	-14849.46	-14782.49	-14858.30
33	-20619.05	-18795.79	-18585.83	-18666.65
34	-14594.33	-14267.87	-14300.34	-14373.66
35	-14285.03	-14081.38	-14127.60	-14207.09
36	-14362.99	-14110.77	-14150.52	-14231.34
37	-15562.10	-15150.25	-15167.56	-15241.09
38	-22301.38	-20771.07	-20560.19	-20695.60
39	-14206.67	-13989.50	-14033.50	-14115.08
40	-18425.98	-16929.24	-16758.78	-16837.05
41	-15036.26	-14587.89	-13998.17	-14668.25
42	-14130.77	-13966.08	-13967.89	-14100.20
43	-21026.25	-19286.97	-14029.38	-19163.32
44	-14702.12	-14305.91	-14074.20	-14400.52
45	-14535.61	-14246.97	-14055.67	-14364.22
46	-14033.54	-13879.49	-13931.07	-14016.62
47	-14646.55	-14311.88	-14060.72	-14416.79
48	-14202.07	-14034.29	-14084.18	-14167.26
49	-15881.87	-15370.93	-15388.34	-15455.83
50	-17194.69	-15754.90	-14582.78	-15656.66
51	-14134.95	-13955.75	-14009.19	-14085.87
52	-14304.77	-14061.23	-14104.59	-14181.90
53	-14455.16	-14211.14	-14212.07	-14328.78
54	-18521.18	-17093.12	-14216.92	-17021.34
55	-14426.67	-14161.85	-14197.70	-14279.61
56	-14196.09	-13982.32	-14024.10	-14104.90
57	-14058.82	-13906.32	-13957.23	-14042.10
58	-14332.19	-14116.10	-14163.62	-14240.85
59	-15029.47	-14580.45	-14177.02	-14665.01
60	-16262.03	-15584.04	-14312.50	-15631.62
61	-17993.25	-16439.33	-14374.14	-16332.83
62	-14295.28	-14076.56	-14118.87	-14199.32
63	-14762.69	-14489.23	-14420.07	-14602.71
64	-14132.75	-13947.06	-13997.43	-14076.36
65	-15089.54	-14565.67	-14327.90	-14636.78
66	-14180.31	-14006.88	-14059.63	-14139.93
67	-15908.03	-15157.62	-14564.38	-15178.64
68	-15236.25	-14763.30	-14385.39	-14846.82
69	-17355.89	-15629.22	-14256.48	-15481.26
70	-14680.42	-14294.84	-14312.94	-14390.01
71	-14602.22	-14303.72	-14339.16	-14416.68
72	-14074.68	-13911.65	-13963.53	-14044.31
73	-14309.93	-14094.73	-14147.97	-14219.93
74	-15163.22	-14669.63	-14678.89	-14749.04

75	-28456.84	-21816.30	-15347.42	-20796.20
76	-14616.95	-14290.61	-14318.28	-14397.35
77	-14217.95	-13986.81	-14027.69	-14109.78
78	-14166.60	-13977.79	-14027.05	-14112.34
79	-90642.13	-53385.79	-13949.08	-46832.15
80	-14789.52	-14509.00	-14040.84	-14621.72
81	-16932.25	-15753.93	-14086.49	-15714.18
82	-15183.87	-14711.52	-14016.21	-14795.00
83	-14521.13	-14203.82	-13959.92	-14311.90
84	-15331.09	-14815.98	-14041.73	-14891.31
85	-14356.99	-14111.49	-14152.63	-14236.42
86	-15205.66	-14755.01	-14356.24	-14836.22
87	-14253.84	-14006.55	-14048.11	-14125.10
88	-14361.87	-14140.02	-14185.38	-14267.39
89	-14511.79	-14232.98	-14115.18	-14347.87
90	-14649.69	-14361.75	-14393.40	-14472.59
91	-14898.54	-14421.88	-14254.46	-14502.63
92	-16105.58	-14995.99	-14002.89	-14958.37
93	-14184.06	-13965.53	-14010.38	-14090.09
94	-16187.92	-15336.40	-14333.32	-15355.14
95	-15061.45	-14681.76	-14300.52	-14777.15
96	-14136.22	-13969.10	-14022.38	-14106.96
97	-14606.54	-14329.71	-14081.67	-14439.80
98	-14474.08	-14258.43	-13970.06	-14378.26
99	-14599.02	-14272.12	-14305.55	-14379.12
100	-14497.32	-14257.84	-14299.78	-14377.69
All	-1637903.68	-1532369.29	-1460493.10	-1530011.88

- (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 = 156.916058]

#### 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
$\Theta_1$	372744131/400014960	0.93183
Genealogies	188365507/1599985040	0.11773

## MCMC-Autocorrelation and Effective MCMC Sample Size

Parameter	Autocorrelation	Effective Sampe Size
$\Theta_1$	0.45814	9660538.23
Genealogies	0.44091	10679924.16

## 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
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