

# 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 100 compute nodes are available.

Program started at Sat Aug 12 18:18:33 2017

Program finished at Sat Aug 12 19:42:55 2017 [Runtime:0000:01:24:22]



## 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)

267487929

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

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  
 -Population size estimation: Exponential Distribution

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	Prior	Minimum	Mean	Maximum	Delta	Bins	UpdateFreq
1	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	1
Recorded steps [a]	50000
Increment (record every x step [b])	200
Number of concurrent chains (replicates) [c]	2
Visited (sampled) parameter values [a*b*c]	20000000
Number of discard trees per chain (burn-in)	10000

Multiple Markov chains:

Static heating scheme

1000000.00	4 chains with temperatures	3.00	1.50	1.00
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Swapping interval is 1

Print options:

Data file:	infile.0.7
Haplotyping is turned on:	NO
Output file:	outfile_0.7_0.7
Posterior distribution raw histogram file:	bayesfile
Raw data from the MCMC run:	bayesallfile_0.7_0.7
Print data:	No
Print genealogies [only some for some data type]:	None

## Data summary

Data file: infile.0.7  
 Datatype: Sequence data  
 Number of loci: 100

Mutationmodel:

Locus	Sublocus	Mutationmodel	Mutationmodel parameters
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1	1	Jukes-Cantor	[Basefreq: =0.25]
2	1	Jukes-Cantor	[Basefreq: =0.25]
3	1	Jukes-Cantor	[Basefreq: =0.25]
4	1	Jukes-Cantor	[Basefreq: =0.25]
5	1	Jukes-Cantor	[Basefreq: =0.25]
6	1	Jukes-Cantor	[Basefreq: =0.25]
7	1	Jukes-Cantor	[Basefreq: =0.25]
8	1	Jukes-Cantor	[Basefreq: =0.25]
9	1	Jukes-Cantor	[Basefreq: =0.25]
10	1	Jukes-Cantor	[Basefreq: =0.25]
11	1	Jukes-Cantor	[Basefreq: =0.25]
12	1	Jukes-Cantor	[Basefreq: =0.25]
13	1	Jukes-Cantor	[Basefreq: =0.25]
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40	1	Jukes-Cantor	[Basefreq: =0.25]
41	1	Jukes-Cantor	[Basefreq: =0.25]
42	1	Jukes-Cantor	[Basefreq: =0.25]
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44	1	Jukes-Cantor	[Basefreq: =0.25]
45	1	Jukes-Cantor	[Basefreq: =0.25]
46	1	Jukes-Cantor	[Basefreq: =0.25]
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50	1	Jukes-Cantor	[Basefreq: =0.25]
51	1	Jukes-Cantor	[Basefreq: =0.25]
52	1	Jukes-Cantor	[Basefreq: =0.25]
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71	1	Jukes-Cantor	[Basefreq: =0.25]
72	1	Jukes-Cantor	[Basefreq: =0.25]
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74	1	Jukes-Cantor	[Basefreq: =0.25]
75	1	Jukes-Cantor	[Basefreq: =0.25]
76	1	Jukes-Cantor	[Basefreq: =0.25]
77	1	Jukes-Cantor	[Basefreq: =0.25]
78	1	Jukes-Cantor	[Basefreq: =0.25]
79	1	Jukes-Cantor	[Basefreq: =0.25]

80	1	Jukes-Cantor	[Basefreq: =0.25]
81	1	Jukes-Cantor	[Basefreq: =0.25]
82	1	Jukes-Cantor	[Basefreq: =0.25]
83	1	Jukes-Cantor	[Basefreq: =0.25]
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86	1	Jukes-Cantor	[Basefreq: =0.25]
87	1	Jukes-Cantor	[Basefreq: =0.25]
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89	1	Jukes-Cantor	[Basefreq: =0.25]
90	1	Jukes-Cantor	[Basefreq: =0.25]
91	1	Jukes-Cantor	[Basefreq: =0.25]
92	1	Jukes-Cantor	[Basefreq: =0.25]
93	1	Jukes-Cantor	[Basefreq: =0.25]
94	1	Jukes-Cantor	[Basefreq: =0.25]
95	1	Jukes-Cantor	[Basefreq: =0.25]
96	1	Jukes-Cantor	[Basefreq: =0.25]
97	1	Jukes-Cantor	[Basefreq: =0.25]
98	1	Jukes-Cantor	[Basefreq: =0.25]
99	1	Jukes-Cantor	[Basefreq: =0.25]
100	1	Jukes-Cantor	[Basefreq: =0.25]

#### 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
11	10000
12	10000
13	10000
14	10000
15	10000
16	10000
17	10000
18	10000
19	10000
20	10000

21	10000
22	10000
23	10000
24	10000
25	10000
26	10000
27	10000
28	10000
29	10000
30	10000
31	10000
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34	10000
35	10000
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37	10000
38	10000
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41	10000
42	10000
43	10000
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45	10000
46	10000
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81	10000
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90	10000
91	10000
92	10000
93	10000
94	10000
95	10000
96	10000
97	10000
98	10000
99	10000
100	10000

Site rate variation and probabilities:

Locus	Sublocus	Region	type	Rate of change	Probability	Patch size
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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

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
11	1	1	1.000	1.000	1.000
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
19	1	1	1.000	1.000	1.000
20	1	1	1.000	1.000	1.000
21	1	1	1.000	1.000	1.000
22	1	1	1.000	1.000	1.000
23	1	1	1.000	1.000	1.000
24	1	1	1.000	1.000	1.000
25	1	1	1.000	1.000	1.000
26	1	1	1.000	1.000	1.000
27	1	1	1.000	1.000	1.000
28	1	1	1.000	1.000	1.000
29	1	1	1.000	1.000	1.000
30	1	1	1.000	1.000	1.000
31	1	1	1.000	1.000	1.000
32	1	1	1.000	1.000	1.000
33	1	1	1.000	1.000	1.000
34	1	1	1.000	1.000	1.000
35	1	1	1.000	1.000	1.000
36	1	1	1.000	1.000	1.000
37	1	1	1.000	1.000	1.000
38	1	1	1.000	1.000	1.000
39	1	1	1.000	1.000	1.000
40	1	1	1.000	1.000	1.000
41	1	1	1.000	1.000	1.000
42	1	1	1.000	1.000	1.000
43	1	1	1.000	1.000	1.000
44	1	1	1.000	1.000	1.000
45	1	1	1.000	1.000	1.000
46	1	1	1.000	1.000	1.000
47	1	1	1.000	1.000	1.000
48	1	1	1.000	1.000	1.000
49	1	1	1.000	1.000	1.000
50	1	1	1.000	1.000	1.000
51	1	1	1.000	1.000	1.000



52	1	1	1.000	1.000	1.000
53	1	1	1.000	1.000	1.000
54	1	1	1.000	1.000	1.000
55	1	1	1.000	1.000	1.000
56	1	1	1.000	1.000	1.000
57	1	1	1.000	1.000	1.000
58	1	1	1.000	1.000	1.000
59	1	1	1.000	1.000	1.000
60	1	1	1.000	1.000	1.000
61	1	1	1.000	1.000	1.000
62	1	1	1.000	1.000	1.000
63	1	1	1.000	1.000	1.000
64	1	1	1.000	1.000	1.000
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
68	1	1	1.000	1.000	1.000
69	1	1	1.000	1.000	1.000
70	1	1	1.000	1.000	1.000
71	1	1	1.000	1.000	1.000
72	1	1	1.000	1.000	1.000
73	1	1	1.000	1.000	1.000
74	1	1	1.000	1.000	1.000
75	1	1	1.000	1.000	1.000
76	1	1	1.000	1.000	1.000
77	1	1	1.000	1.000	1.000
78	1	1	1.000	1.000	1.000
79	1	1	1.000	1.000	1.000
80	1	1	1.000	1.000	1.000
81	1	1	1.000	1.000	1.000
82	1	1	1.000	1.000	1.000
83	1	1	1.000	1.000	1.000
84	1	1	1.000	1.000	1.000
85	1	1	1.000	1.000	1.000
86	1	1	1.000	1.000	1.000
87	1	1	1.000	1.000	1.000
88	1	1	1.000	1.000	1.000
89	1	1	1.000	1.000	1.000
90	1	1	1.000	1.000	1.000
91	1	1	1.000	1.000	1.000
92	1	1	1.000	1.000	1.000
93	1	1	1.000	1.000	1.000
94	1	1	1.000	1.000	1.000
95	1	1	1.000	1.000	1.000
96	1	1	1.000	1.000	1.000

97	1	1	1.000	1.000	1.000	
98	1	1	1.000	1.000	1.000	
99	1	1	1.000	1.000	1.000	
100	1	1	1.000	1.000	1.000	
Population			Locus		Gene copies	
1 Romanshorn_0			1		10	
			2		10	
			3		10	
			4		10	
			5		10	
			6		10	
			7		10	
			8		10	
			9		10	
			10		10	
			11		10	
			12		10	
			13		10	
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41	10
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	86	10
	87	10
	88	10
	89	10
	90	10
	91	10
	92	10
	93	10
	94	10
	95	10
	96	10
	97	10
	98	10
	99	10
	100	10
Total of all populations	1	10
	2	10
	3	10
	4	10
	5	10
	6	10
	7	10
	8	10
	9	10
	10	10
	11	10
	12	10
	13	10
	14	10
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92	10
93	10
94	10
95	10
96	10
97	10
98	10
99	10
100	10

## *Bayesian Analysis: Posterior distribution table*

Locus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
1	$\Theta_1$	0.00687	0.01173	0.01890	0.03187	0.04900	0.02390	0.02922
2	$\Theta_1$	0.00400	0.00987	0.01217	0.01527	0.03373	0.01517	0.01702
3	$\Theta_1$	0.00473	0.01233	0.01557	0.01833	0.04273	0.01870	0.02140
4	$\Theta_1$	0.00320	0.00953	0.01103	0.01267	0.03033	0.01410	0.01591
5	$\Theta_1$	0.00533	0.00847	0.01217	0.01740	0.02627	0.01510	0.01697
6	$\Theta_1$	0.00720	0.01080	0.01757	0.02820	0.04153	0.02117	0.02405
7	$\Theta_1$	0.00480	0.00833	0.01590	0.02827	0.04413	0.01937	0.02211
8	$\Theta_1$	0.00167	0.00500	0.00750	0.01100	0.02200	0.00943	0.01059
9	$\Theta_1$	0.01527	0.02187	0.03250	0.04500	0.04993	0.03270	0.04170
10	$\Theta_1$	0.00420	0.00580	0.00957	0.01500	0.01927	0.01190	0.01336
11	$\Theta_1$	0.00220	0.00667	0.01110	0.01713	0.04180	0.01563	0.01961
12	$\Theta_1$	0.00660	0.01000	0.01583	0.02493	0.03653	0.01937	0.02194
13	$\Theta_1$	0.00773	0.01113	0.01750	0.02560	0.03553	0.02150	0.02547
14	$\Theta_1$	0.00580	0.00907	0.01397	0.02160	0.03273	0.01830	0.02189
15	$\Theta_1$	0.00640	0.01540	0.01783	0.02040	0.04627	0.02150	0.02449
16	$\Theta_1$	0.00480	0.00953	0.01290	0.01660	0.03000	0.01550	0.01739
17	$\Theta_1$	0.00840	0.01553	0.01903	0.02333	0.04220	0.02303	0.02654
18	$\Theta_1$	0.00040	0.00287	0.00470	0.00700	0.01420	0.00590	0.00657

19	$\Theta_1$	0.00173	0.00393	0.00530	0.00700	0.01147	0.00670	0.00750
20	$\Theta_1$	0.00407	0.00433	0.00937	0.01853	0.01933	0.01170	0.01318
21	$\Theta_1$	0.00487	0.00967	0.01303	0.01773	0.03187	0.01630	0.01832
22	$\Theta_1$	0.00567	0.01133	0.01503	0.02067	0.03713	0.01903	0.02159
23	$\Theta_1$	0.00287	0.00560	0.00723	0.00913	0.01520	0.00903	0.01009
24	$\Theta_1$	0.00513	0.01060	0.01403	0.01880	0.03533	0.01797	0.02047
25	$\Theta_1$	0.00540	0.01013	0.01330	0.01747	0.03073	0.01677	0.01904
26	$\Theta_1$	0.00873	0.01347	0.01977	0.02713	0.04047	0.02343	0.02774
27	$\Theta_1$	0.00447	0.00780	0.01237	0.01913	0.03240	0.01637	0.01944
28	$\Theta_1$	0.00453	0.01007	0.01117	0.01213	0.02373	0.01377	0.01555
29	$\Theta_1$	0.00447	0.00780	0.00990	0.01227	0.01933	0.01223	0.01378
30	$\Theta_1$	0.00300	0.00760	0.01057	0.01427	0.02893	0.01337	0.01515
31	$\Theta_1$	0.00267	0.00680	0.01030	0.01480	0.02900	0.01277	0.01431
32	$\Theta_1$	0.00073	0.00347	0.00550	0.00807	0.01607	0.00683	0.00762
33	$\Theta_1$	0.01067	0.01540	0.02190	0.03080	0.04393	0.02623	0.03341
34	$\Theta_1$	0.01373	0.01373	0.02803	0.04933	0.04933	0.03070	0.04082
35	$\Theta_1$	0.00173	0.00320	0.00470	0.00660	0.00940	0.00597	0.00668
36	$\Theta_1$	0.00520	0.01320	0.01423	0.01527	0.03547	0.01730	0.01957
37	$\Theta_1$	0.00300	0.00700	0.01017	0.01460	0.02833	0.01263	0.01415
38	$\Theta_1$	0.00487	0.00887	0.01357	0.02133	0.03453	0.01723	0.01956
39	$\Theta_1$	0.00460	0.01040	0.01430	0.01880	0.03747	0.01770	0.02018
40	$\Theta_1$	0.01627	0.02380	0.03217	0.04333	0.04980	0.03297	0.04291
41	$\Theta_1$	0.00987	0.01507	0.02117	0.02867	0.04300	0.02577	0.03347



Locus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
42	$\Theta_1$	0.00553	0.00880	0.01137	0.01467	0.02207	0.01423	0.01601
43	$\Theta_1$	0.00727	0.01013	0.01823	0.03287	0.04527	0.02303	0.02786
44	$\Theta_1$	0.00627	0.01080	0.01590	0.02347	0.04053	0.02057	0.02487
45	$\Theta_1$	0.00800	0.00940	0.01850	0.03407	0.03847	0.02270	0.02697
46	$\Theta_1$	0.00507	0.00693	0.01437	0.02707	0.03513	0.01710	0.01926
47	$\Theta_1$	0.00760	0.01347	0.01477	0.01633	0.02827	0.01810	0.02045
48	$\Theta_1$	0.00333	0.00440	0.00917	0.01727	0.02080	0.01163	0.01314
49	$\Theta_1$	0.00153	0.00480	0.00717	0.01053	0.02080	0.00897	0.01000
50	$\Theta_1$	0.00267	0.00893	0.01083	0.01307	0.03447	0.01330	0.01492
51	$\Theta_1$	0.00407	0.00760	0.00937	0.01133	0.01927	0.01157	0.01294
52	$\Theta_1$	0.01473	0.02613	0.03463	0.04227	0.05000	0.03290	0.04712
53	$\Theta_1$	0.00173	0.00500	0.00743	0.01080	0.02127	0.00923	0.01033
54	$\Theta_1$	0.00293	0.00353	0.01070	0.02700	0.03073	0.01410	0.01680
55	$\Theta_1$	0.00627	0.01407	0.01743	0.02160	0.04440	0.02123	0.02448
56	$\Theta_1$	0.00580	0.01240	0.01643	0.02153	0.04187	0.02017	0.02315
57	$\Theta_1$	0.00480	0.00767	0.01503	0.02820	0.04287	0.01943	0.02333
58	$\Theta_1$	0.00700	0.01207	0.01630	0.02200	0.03640	0.01997	0.02260
59	$\Theta_1$	0.00153	0.00507	0.00797	0.01207	0.02513	0.01037	0.01177
60	$\Theta_1$	0.00180	0.00387	0.00603	0.00893	0.01427	0.00757	0.00846
61	$\Theta_1$	0.00953	0.02120	0.02823	0.03727	0.05073	0.03043	0.03944

62	$\Theta_1$	0.00720	0.01267	0.01943	0.02940	0.04727	0.02390	0.02900
63	$\Theta_1$	0.00767	0.01787	0.02377	0.03053	0.05040	0.02723	0.03332
64	$\Theta_1$	0.00747	0.01520	0.01903	0.02400	0.04793	0.02297	0.02612
65	$\Theta_1$	0.00380	0.00740	0.01010	0.01360	0.02353	0.01263	0.01421
66	$\Theta_1$	0.00607	0.00807	0.01517	0.02807	0.03553	0.01883	0.02126
67	$\Theta_1$	0.00900	0.01493	0.01937	0.02507	0.04027	0.02317	0.02657
68	$\Theta_1$	0.01047	0.01413	0.02130	0.03100	0.04120	0.02550	0.03166
69	$\Theta_1$	0.00187	0.00700	0.00817	0.00953	0.02373	0.01017	0.01131
70	$\Theta_1$	0.00260	0.00480	0.00723	0.01053	0.01607	0.00897	0.01001
71	$\Theta_1$	0.01060	0.02040	0.02710	0.03500	0.05060	0.02977	0.03632
72	$\Theta_1$	0.01553	0.02573	0.03010	0.04020	0.04980	0.03270	0.04375
73	$\Theta_1$	0.00240	0.00433	0.00563	0.00713	0.01087	0.00703	0.00788
74	$\Theta_1$	0.00907	0.01233	0.02017	0.03233	0.04353	0.02417	0.02860
75	$\Theta_1$	0.02293	0.03867	0.04750	0.04880	0.05100	0.03957	0.05940
76	$\Theta_1$	0.00553	0.00973	0.01623	0.02693	0.04500	0.01990	0.02253
77	$\Theta_1$	0.00293	0.00680	0.00997	0.01407	0.02680	0.01217	0.01360
78	$\Theta_1$	0.01087	0.01460	0.02190	0.03027	0.04107	0.02497	0.02911
79	$\Theta_1$	0.00200	0.00500	0.00823	0.01313	0.02573	0.01110	0.01302
80	$\Theta_1$	0.00180	0.00500	0.00603	0.00733	0.01440	0.00763	0.00854
81	$\Theta_1$	0.00347	0.00653	0.00990	0.01427	0.02380	0.01297	0.01513
82	$\Theta_1$	0.00407	0.00913	0.00970	0.01033	0.02100	0.01230	0.01381
83	$\Theta_1$	0.00387	0.00693	0.01030	0.01460	0.02313	0.01283	0.01437
84	$\Theta_1$	0.01360	0.02020	0.02650	0.03353	0.04840	0.02943	0.03673

Locus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
85	$\Theta_1$	0.01373	0.01753	0.02497	0.03460	0.04360	0.02803	0.03327
86	$\Theta_1$	0.00133	0.00687	0.00750	0.00820	0.02407	0.00943	0.01055
87	$\Theta_1$	0.00700	0.01347	0.01890	0.02687	0.04633	0.02330	0.02763
88	$\Theta_1$	0.00353	0.00927	0.01343	0.01920	0.04160	0.01710	0.01956
89	$\Theta_1$	0.00433	0.00713	0.01030	0.01480	0.02200	0.01277	0.01425
90	$\Theta_1$	0.00253	0.00647	0.00970	0.01407	0.02787	0.01217	0.01364
91	$\Theta_1$	0.00447	0.01273	0.01377	0.01487	0.03633	0.01690	0.01894
92	$\Theta_1$	0.01460	0.02547	0.03123	0.03713	0.04993	0.03230	0.04306
93	$\Theta_1$	0.00667	0.01153	0.01643	0.02267	0.03800	0.02003	0.02290
94	$\Theta_1$	0.00273	0.00660	0.01197	0.02047	0.03780	0.01530	0.01743
95	$\Theta_1$	0.00587	0.01080	0.01497	0.02087	0.03680	0.01837	0.02061
96	$\Theta_1$	0.00487	0.00980	0.01117	0.01260	0.02313	0.01390	0.01565
97	$\Theta_1$	0.01460	0.02367	0.02977	0.04113	0.04967	0.03197	0.04248
98	$\Theta_1$	0.01460	0.02493	0.03130	0.03827	0.04973	0.03197	0.04179
99	$\Theta_1$	0.00380	0.00733	0.01083	0.01593	0.02700	0.01370	0.01550
100	$\Theta_1$	0.00173	0.00707	0.00870	0.01067	0.03113	0.01177	0.01388
All	$\Theta_1$	0.01027	0.01180	0.01283	0.01380	0.01527	0.01290	0.01287

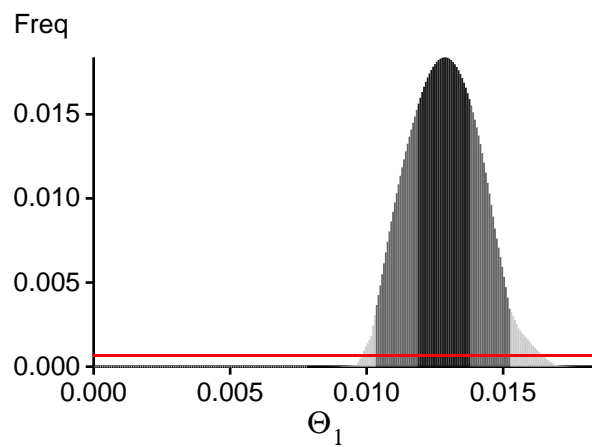
Citation suggestions:

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:

$BF = \text{Exp}[\ln(\text{Prob}(D \mid \text{thisModel}) - \ln(\text{Prob}(D \mid \text{otherModel}))]$

or as  $LBF = 2 (\ln(\text{Prob}(D \mid \text{thisModel}) - \ln(\text{Prob}(D \mid \text{otherModel})))$

shows the support for thisModel]

Locus	TI(1a)	BTI(1b)	SS(2)	HS(3)
1	-17267.49	-16297.81	-16241.69	-16305.53
2	-14303.79	-14043.29	-14088.76	-14160.38
3	-14145.73	-13948.59	-14006.45	-14077.67
4	-14254.21	-14025.01	-14072.56	-14149.20
5	-14122.10	-13937.83	-13993.96	-14065.75
6	-14485.09	-14194.10	-14239.37	-14309.10
7	-14384.09	-14158.60	-14215.80	-14283.87
8	-14003.27	-13824.75	-13872.39	-13952.73
9	-15377.17	-14845.24	-14860.68	-14920.14
10	-14361.94	-14109.33	-14152.29	-14227.49
11	-17169.89	-16805.10	-16857.08	-16934.49
12	-14692.96	-14372.96	-14415.27	-14482.43
13	-14489.53	-14240.60	-14292.67	-14361.49
14	-18142.72	-16778.31	-16647.65	-16715.11
15	-14391.25	-14123.06	-14174.55	-14239.46
16	-14187.95	-13971.06	-14023.93	-14095.23
17	-14297.62	-14078.89	-14138.92	-14204.67
18	-13937.44	-13763.07	-13802.13	-13891.91
19	-13949.94	-13775.34	-13817.69	-13904.25
20	-14032.92	-13851.57	-13903.97	-13980.52
21	-14498.48	-14164.05	-14196.59	-14268.84
22	-14398.59	-14163.72	-14219.01	-14286.96
23	-14013.12	-13825.51	-13873.09	-13954.14
24	-14964.81	-14698.11	-14751.53	-14824.71
25	-14079.90	-13897.77	-13956.40	-14027.37
26	-14923.29	-14540.83	-14574.22	-14639.85
27	-16976.99	-15907.92	-15822.59	-15893.37
28	-14677.61	-14300.23	-14323.81	-14397.34
29	-14035.29	-13857.11	-13910.39	-13988.34

30	-14029.97	-13853.17	-13902.91	-13982.28
31	-14759.62	-14308.73	-14317.42	-14392.62
32	-13967.20	-13785.07	-13828.18	-13913.36
33	-18791.63	-17591.39	-17503.08	-17568.78
34	-17363.92	-16943.67	-17003.06	-17060.60
35	-13933.38	-13761.52	-13803.94	-13892.47
36	-14291.11	-14049.52	-14099.05	-14169.90
37	-14153.96	-13928.64	-13976.95	-14051.22
38	-14790.26	-14352.68	-14369.67	-14439.28
39	-14592.35	-14291.69	-14333.59	-14404.71
40	-14785.51	-14489.89	-14545.82	-14607.95
41	-14920.98	-14712.31	-14774.90	-14843.40
42	-14075.67	-13885.01	-13941.14	-14013.90
43	-14914.75	-14609.99	-14658.24	-14725.46
44	-16732.84	-15958.10	-15932.09	-15998.27
45	-14202.91	-14020.06	-14079.25	-14151.02
46	-14581.48	-14240.49	-14275.08	-14345.31
47	-14217.55	-13995.32	-14051.04	-14118.50
48	-14244.15	-14000.23	-14043.36	-14120.08
49	-14041.68	-13844.46	-13891.12	-13971.02
50	-14172.08	-13945.16	-13993.62	-14067.25
51	-14104.70	-13900.09	-13949.06	-14025.89
52	-33889.39	-24058.15	-22393.20	-22448.87
53	-14060.93	-13883.92	-13934.61	-14015.02
54	-16373.87	-15665.72	-15639.73	-15715.41
55	-14213.92	-14021.90	-14081.52	-14152.43
56	-14147.00	-13955.16	-14013.90	-14083.49
57	-15341.77	-14911.91	-14937.60	-15006.74
58	-14252.17	-14045.62	-14105.91	-14174.40
59	-14173.28	-13984.50	-14035.21	-14115.38
60	-13999.26	-13813.97	-13858.73	-13942.69
61	-15101.31	-14839.36	-14902.61	-14965.42
62	-14300.37	-14111.75	-14169.35	-14242.47
63	-15694.90	-15054.83	-15046.98	-15109.57
64	-14587.27	-14249.60	-14288.15	-14353.66
65	-14361.74	-14113.89	-14159.50	-14235.08
66	-14533.26	-14318.54	-14377.28	-14446.36
67	-15003.71	-14494.12	-14502.74	-14568.94
68	-14435.68	-14225.53	-14287.44	-14353.40
69	-14187.49	-13944.70	-13986.27	-14064.05
70	-14140.01	-13928.48	-13973.75	-14053.81
71	-14675.39	-14401.51	-14459.51	-14519.45
72	-15698.93	-15205.95	-15232.89	-15292.58
73	-13986.43	-13813.36	-13858.89	-13943.59
74	-17886.41	-16164.30	-15960.51	-16024.73

75	-19826.45	-17479.38	-17182.84	-17234.87
76	-14218.43	-14018.06	-14077.59	-14145.08
77	-14295.96	-14033.00	-14075.21	-14151.20
78	-14694.14	-14351.78	-14390.60	-14456.01
79	-17003.53	-16397.02	-16399.40	-16475.53
80	-13991.45	-13812.09	-13857.99	-13941.22
81	-14828.68	-14463.29	-14489.45	-14564.84
82	-14042.21	-13859.21	-13912.60	-13988.29
83	-14198.30	-13977.21	-14026.30	-14100.89
84	-14837.10	-14512.82	-14561.98	-14624.21
85	-14917.18	-14658.97	-14722.48	-14783.55
86	-14062.92	-13864.72	-13911.33	-13991.44
87	-14588.56	-14340.90	-14396.99	-14463.47
88	-14625.63	-14285.91	-14320.72	-14390.47
89	-14260.92	-14028.16	-14077.52	-14151.89
90	-14075.58	-13880.30	-13931.24	-14008.20
91	-14683.98	-14295.24	-14320.43	-14390.39
92	-19622.94	-18765.98	-18769.81	-18823.66
93	-14144.45	-13957.15	-14019.95	-14088.15
94	-14088.45	-13903.34	-13955.12	-14032.40
95	-14386.78	-14112.86	-14160.58	-14229.37
96	-14079.33	-13884.26	-13937.63	-14010.76
97	-15921.76	-15409.24	-15434.95	-15493.07
98	-16450.43	-15554.27	-15507.16	-15567.53
99	-14062.44	-13875.55	-13931.56	-14004.93
100	-18630.54	-17033.22	-16853.56	-16926.95
All	-1507709.18	-1460821.33	-1462042.94	-1469165.17
(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 = 106.342606]				
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
$\Theta_1$	333869588/400016547	0.83464
Genealogies	269053974/1599983453	0.16816

### *MCMC-Autocorrelation and Effective MCMC Sample Size*

Parameter	Autocorrelation	Effective Sampe Size
$\Theta_1$	0.43159	4186092.91
Genealogies	0.11889	7999360.22

## *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. With many parameters in a multilocus analysis, it is very common that some parameters for some loci will not be very informative, triggering suggestions (for example to increase the prior range) that are not sensible. This suggestion tool will improve with time, therefore do not blindly follow its suggestions. If some parameters are flagged, inspect the tables carefully and judge whether an action is required. For example, if you run a Bayesian inference with sequence data, for macroscopic 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 routes 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