

# 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 18:12:33 2017

Program finished at Sat Aug 12 19:13:52 2017 [Runtime:0000:01:01:19]



## 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)

1879882734

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
	Swapping interval is 1			

Print options:

Data file:	infile.0.4
Haplotyping is turned on:	NO
Output file:	outfile_0.4_0.4
Posterior distribution raw histogram file:	bayesfile
Raw data from the MCMC run:	bayesallfile_0.4_0.4
Print data:	No
Print genealogies [only some for some data type]:	None

## Data summary

Data file: infile.0.4  
 Datatype: Sequence data  
 Number of loci: 100

Mutationmodel:

Locus	Sublocus	Mutationmodel	Mutationmodel parameters
-------	----------	---------------	--------------------------

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]
14	1	Jukes-Cantor	[Basefreq: =0.25]
15	1	Jukes-Cantor	[Basefreq: =0.25]
16	1	Jukes-Cantor	[Basefreq: =0.25]
17	1	Jukes-Cantor	[Basefreq: =0.25]
18	1	Jukes-Cantor	[Basefreq: =0.25]
19	1	Jukes-Cantor	[Basefreq: =0.25]
20	1	Jukes-Cantor	[Basefreq: =0.25]
21	1	Jukes-Cantor	[Basefreq: =0.25]
22	1	Jukes-Cantor	[Basefreq: =0.25]
23	1	Jukes-Cantor	[Basefreq: =0.25]
24	1	Jukes-Cantor	[Basefreq: =0.25]
25	1	Jukes-Cantor	[Basefreq: =0.25]
26	1	Jukes-Cantor	[Basefreq: =0.25]
27	1	Jukes-Cantor	[Basefreq: =0.25]
28	1	Jukes-Cantor	[Basefreq: =0.25]
29	1	Jukes-Cantor	[Basefreq: =0.25]
30	1	Jukes-Cantor	[Basefreq: =0.25]
31	1	Jukes-Cantor	[Basefreq: =0.25]
32	1	Jukes-Cantor	[Basefreq: =0.25]
33	1	Jukes-Cantor	[Basefreq: =0.25]
34	1	Jukes-Cantor	[Basefreq: =0.25]

35	1	Jukes-Cantor	[Basefreq: =0.25]
36	1	Jukes-Cantor	[Basefreq: =0.25]
37	1	Jukes-Cantor	[Basefreq: =0.25]
38	1	Jukes-Cantor	[Basefreq: =0.25]
39	1	Jukes-Cantor	[Basefreq: =0.25]
40	1	Jukes-Cantor	[Basefreq: =0.25]
41	1	Jukes-Cantor	[Basefreq: =0.25]
42	1	Jukes-Cantor	[Basefreq: =0.25]
43	1	Jukes-Cantor	[Basefreq: =0.25]
44	1	Jukes-Cantor	[Basefreq: =0.25]
45	1	Jukes-Cantor	[Basefreq: =0.25]
46	1	Jukes-Cantor	[Basefreq: =0.25]
47	1	Jukes-Cantor	[Basefreq: =0.25]
48	1	Jukes-Cantor	[Basefreq: =0.25]
49	1	Jukes-Cantor	[Basefreq: =0.25]
50	1	Jukes-Cantor	[Basefreq: =0.25]
51	1	Jukes-Cantor	[Basefreq: =0.25]
52	1	Jukes-Cantor	[Basefreq: =0.25]
53	1	Jukes-Cantor	[Basefreq: =0.25]
54	1	Jukes-Cantor	[Basefreq: =0.25]
55	1	Jukes-Cantor	[Basefreq: =0.25]
56	1	Jukes-Cantor	[Basefreq: =0.25]
57	1	Jukes-Cantor	[Basefreq: =0.25]
58	1	Jukes-Cantor	[Basefreq: =0.25]
59	1	Jukes-Cantor	[Basefreq: =0.25]
60	1	Jukes-Cantor	[Basefreq: =0.25]
61	1	Jukes-Cantor	[Basefreq: =0.25]
62	1	Jukes-Cantor	[Basefreq: =0.25]
63	1	Jukes-Cantor	[Basefreq: =0.25]
64	1	Jukes-Cantor	[Basefreq: =0.25]
65	1	Jukes-Cantor	[Basefreq: =0.25]
66	1	Jukes-Cantor	[Basefreq: =0.25]
67	1	Jukes-Cantor	[Basefreq: =0.25]
68	1	Jukes-Cantor	[Basefreq: =0.25]
69	1	Jukes-Cantor	[Basefreq: =0.25]
70	1	Jukes-Cantor	[Basefreq: =0.25]
71	1	Jukes-Cantor	[Basefreq: =0.25]
72	1	Jukes-Cantor	[Basefreq: =0.25]
73	1	Jukes-Cantor	[Basefreq: =0.25]
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]
84	1	Jukes-Cantor	[Basefreq: =0.25]
85	1	Jukes-Cantor	[Basefreq: =0.25]
86	1	Jukes-Cantor	[Basefreq: =0.25]
87	1	Jukes-Cantor	[Basefreq: =0.25]
88	1	Jukes-Cantor	[Basefreq: =0.25]
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
32	10000
33	10000
34	10000
35	10000
36	10000
37	10000
38	10000
39	10000
40	10000
41	10000
42	10000
43	10000
44	10000
45	10000
46	10000
47	10000
48	10000
49	10000
50	10000
51	10000
52	10000
53	10000
54	10000
55	10000
56	10000
57	10000
58	10000
59	10000
60	10000
61	10000
62	10000
63	10000
64	10000
65	10000

66	10000
67	10000
68	10000
69	10000
70	10000
71	10000
72	10000
73	10000
74	10000
75	10000
76	10000
77	10000
78	10000
79	10000
80	10000
81	10000
82	10000
83	10000
84	10000
85	10000
86	10000
87	10000
88	10000
89	10000
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
-------	----------	--------	------	----------------	-------------	------------

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	
			14		10	
			15		10	
			16		10	
			17		10	
			18		10	
			19		10	
			20		10	
			21		10	
			22		10	
			23		10	
			24		10	
			25		10	
			26		10	
			27		10	
			28		10	
			29		10	
			30		10	
			31		10	
			32		10	
			33		10	
			34		10	
			35		10	
			36		10	
			37		10	
			38		10	
			39		10	
			40		10	

41	10
42	10
43	10
44	10
45	10
46	10
47	10
48	10
49	10
50	10
51	10
52	10
53	10
54	10
55	10
56	10
57	10
58	10
59	10
60	10
61	10
62	10
63	10
64	10
65	10
66	10
67	10
68	10
69	10
70	10
71	10
72	10
73	10
74	10
75	10
76	10
77	10
78	10
79	10
80	10
81	10
82	10
83	10
84	10
85	10

	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
	15	10
	16	10
	17	10
	18	10
	19	10
	20	10
	21	10
	22	10
	23	10
	24	10
	25	10
	26	10
	27	10
	28	10
	29	10
	30	10

31	10
32	10
33	10
34	10
35	10
36	10
37	10
38	10
39	10
40	10
41	10
42	10
43	10
44	10
45	10
46	10
47	10
48	10
49	10
50	10
51	10
52	10
53	10
54	10
55	10
56	10
57	10
58	10
59	10
60	10
61	10
62	10
63	10
64	10
65	10
66	10
67	10
68	10
69	10
70	10
71	10
72	10
73	10
74	10
75	10

76	10
77	10
78	10
79	10
80	10
81	10
82	10
83	10
84	10
85	10
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

## *Bayesian Analysis: Posterior distribution table*

Locus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
1	$\Theta_1$	0.02800	0.04247	0.04763	0.04947	0.05147	0.04323	0.07523
2	$\Theta_1$	0.03027	0.04393	0.04777	0.04953	0.05153	0.04423	0.07928
3	$\Theta_1$	0.02820	0.04293	0.04770	0.04933	0.05140	0.04310	0.07515
4	$\Theta_1$	0.02833	0.04167	0.04763	0.04947	0.05140	0.04303	0.07526
5	$\Theta_1$	0.02767	0.04180	0.04770	0.04953	0.05147	0.04310	0.07519
6	$\Theta_1$	0.02793	0.04267	0.04770	0.04940	0.05140	0.04297	0.07508
7	$\Theta_1$	0.02827	0.04207	0.04763	0.04953	0.05147	0.04337	0.07628
8	$\Theta_1$	0.03020	0.04307	0.04783	0.04967	0.05160	0.04430	0.08141
9	$\Theta_1$	0.03067	0.04307	0.04777	0.04960	0.05153	0.04437	0.08090
10	$\Theta_1$	0.02813	0.04020	0.04770	0.04987	0.05133	0.04317	0.07521
11	$\Theta_1$	0.02767	0.04200	0.04777	0.04973	0.05153	0.04317	0.07511
12	$\Theta_1$	0.03160	0.04333	0.04790	0.04973	0.05167	0.04457	0.08098
13	$\Theta_1$	0.02887	0.04353	0.04770	0.04933	0.05147	0.04370	0.07730
14	$\Theta_1$	0.02927	0.04267	0.04777	0.04973	0.05147	0.04383	0.07739
15	$\Theta_1$	0.02827	0.04300	0.04777	0.04940	0.05140	0.04317	0.07518
16	$\Theta_1$	0.02780	0.04287	0.04757	0.04940	0.05140	0.04310	0.07521
17	$\Theta_1$	0.02733	0.04180	0.04770	0.04960	0.05147	0.04303	0.07503
18	$\Theta_1$	0.02827	0.04227	0.04770	0.04960	0.05147	0.04350	0.07652

19	$\Theta_1$	0.03027	0.04300	0.04783	0.04973	0.05160	0.04417	0.07945
20	$\Theta_1$	0.02733	0.04180	0.04770	0.04967	0.05140	0.04310	0.07529
21	$\Theta_1$	0.02773	0.04167	0.04777	0.04953	0.05147	0.04303	0.07508
22	$\Theta_1$	0.02960	0.04287	0.04783	0.04973	0.05153	0.04410	0.07890
23	$\Theta_1$	0.02793	0.04273	0.04763	0.04940	0.05140	0.04303	0.07505
24	$\Theta_1$	0.02787	0.04187	0.04777	0.04973	0.05153	0.04310	0.07483
25	$\Theta_1$	0.02727	0.04173	0.04763	0.04953	0.05153	0.04303	0.07517
26	$\Theta_1$	0.02793	0.04293	0.04763	0.04927	0.05140	0.04310	0.07519
27	$\Theta_1$	0.02833	0.04187	0.04763	0.04953	0.05133	0.04317	0.07512
28	$\Theta_1$	0.02820	0.04033	0.04770	0.04980	0.05140	0.04290	0.07507
29	$\Theta_1$	0.02760	0.04227	0.04783	0.04960	0.05147	0.04297	0.07526
30	$\Theta_1$	0.02853	0.04220	0.04770	0.04960	0.05147	0.04343	0.07691
31	$\Theta_1$	0.02760	0.04173	0.04770	0.04960	0.05140	0.04303	0.07501
32	$\Theta_1$	0.02753	0.04173	0.04770	0.04953	0.05153	0.04303	0.07523
33	$\Theta_1$	0.02787	0.04200	0.04770	0.04953	0.05140	0.04323	0.07547
34	$\Theta_1$	0.02760	0.04180	0.04770	0.04960	0.05147	0.04303	0.07513
35	$\Theta_1$	0.02827	0.04173	0.04770	0.04947	0.05140	0.04303	0.07496
36	$\Theta_1$	0.02913	0.04227	0.04790	0.04967	0.05153	0.04350	0.07777
37	$\Theta_1$	0.02873	0.04240	0.04777	0.04967	0.05147	0.04357	0.07798
38	$\Theta_1$	0.02793	0.04173	0.04763	0.04947	0.05147	0.04303	0.07505
39	$\Theta_1$	0.02787	0.04187	0.04777	0.04973	0.05147	0.04303	0.07515
40	$\Theta_1$	0.02787	0.04287	0.04770	0.04927	0.05140	0.04303	0.07517
41	$\Theta_1$	0.02793	0.04173	0.04777	0.04960	0.05140	0.04297	0.07512



Locus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
42	$\Theta_1$	0.03013	0.04173	0.04777	0.05000	0.05153	0.04423	0.07851
43	$\Theta_1$	0.02793	0.04280	0.04777	0.04940	0.05140	0.04297	0.07507
44	$\Theta_1$	0.02760	0.04173	0.04757	0.04947	0.05140	0.04303	0.07513
45	$\Theta_1$	0.02760	0.04200	0.04777	0.04973	0.05153	0.04323	0.07515
46	$\Theta_1$	0.02740	0.04147	0.04777	0.04953	0.05147	0.04277	0.07509
47	$\Theta_1$	0.02753	0.04173	0.04763	0.04953	0.05147	0.04303	0.07526
48	$\Theta_1$	0.02760	0.04180	0.04770	0.04953	0.05153	0.04310	0.07506
49	$\Theta_1$	0.02733	0.04167	0.04770	0.04960	0.05153	0.04297	0.07513
50	$\Theta_1$	0.02813	0.04280	0.04777	0.04947	0.05147	0.04317	0.07535
51	$\Theta_1$	0.02927	0.04253	0.04777	0.04967	0.05147	0.04377	0.07784
52	$\Theta_1$	0.02813	0.04187	0.04777	0.04967	0.05147	0.04310	0.07524
53	$\Theta_1$	0.02700	0.04160	0.04763	0.04940	0.05140	0.04297	0.07506
54	$\Theta_1$	0.02873	0.04240	0.04770	0.04967	0.05147	0.04363	0.07733
55	$\Theta_1$	0.02813	0.04300	0.04777	0.04933	0.05147	0.04317	0.07514
56	$\Theta_1$	0.03007	0.04293	0.04777	0.04967	0.05153	0.04417	0.07946
57	$\Theta_1$	0.02760	0.04180	0.04770	0.04960	0.05147	0.04303	0.07505
58	$\Theta_1$	0.03147	0.04413	0.04777	0.04960	0.05147	0.04457	0.08102
59	$\Theta_1$	0.03067	0.04300	0.04783	0.04973	0.05167	0.04423	0.08058
60	$\Theta_1$	0.02773	0.04200	0.04777	0.04973	0.05147	0.04317	0.07540
61	$\Theta_1$	0.02893	0.04233	0.04777	0.04953	0.05153	0.04363	0.07731

62	$\Theta_1$	0.02780	0.04293	0.04770	0.04933	0.05147	0.04310	0.07512
63	$\Theta_1$	0.02747	0.04173	0.04777	0.04960	0.05147	0.04297	0.07519
64	$\Theta_1$	0.02880	0.04347	0.04777	0.04940	0.05153	0.04363	0.07710
65	$\Theta_1$	0.02800	0.04207	0.04777	0.04960	0.05147	0.04310	0.07525
66	$\Theta_1$	0.03107	0.04373	0.04790	0.04993	0.05160	0.04483	0.08248
67	$\Theta_1$	0.02953	0.04353	0.04783	0.04940	0.05153	0.04370	0.07775
68	$\Theta_1$	0.02913	0.04253	0.04777	0.04973	0.05153	0.04377	0.07736
69	$\Theta_1$	0.02780	0.04180	0.04777	0.04960	0.05147	0.04303	0.07534
70	$\Theta_1$	0.02820	0.04247	0.04763	0.04933	0.05133	0.04303	0.07531
71	$\Theta_1$	0.02753	0.04167	0.04777	0.04960	0.05147	0.04297	0.07497
72	$\Theta_1$	0.02827	0.04200	0.04777	0.04967	0.05147	0.04323	0.07630
73	$\Theta_1$	0.02947	0.04293	0.04777	0.04980	0.05153	0.04410	0.07881
74	$\Theta_1$	0.02733	0.04260	0.04763	0.04927	0.05140	0.04283	0.07483
75	$\Theta_1$	0.02993	0.04367	0.04777	0.04953	0.05147	0.04403	0.07817
76	$\Theta_1$	0.02760	0.04173	0.04777	0.04967	0.05140	0.04297	0.07500
77	$\Theta_1$	0.02947	0.04273	0.04770	0.04953	0.05153	0.04403	0.07827
78	$\Theta_1$	0.02807	0.04207	0.04777	0.04973	0.05147	0.04330	0.07663
79	$\Theta_1$	0.02933	0.04293	0.04777	0.04953	0.05147	0.04377	0.07836
80	$\Theta_1$	0.02753	0.04207	0.04783	0.04980	0.05153	0.04323	0.07534
81	$\Theta_1$	0.03060	0.04293	0.04777	0.04973	0.05147	0.04410	0.07981
82	$\Theta_1$	0.02787	0.04300	0.04770	0.04940	0.05140	0.04317	0.07521
83	$\Theta_1$	0.02960	0.04353	0.04777	0.04940	0.05153	0.04377	0.07708
84	$\Theta_1$	0.02780	0.04187	0.04770	0.04960	0.05147	0.04317	0.07505

Locus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
85	$\Theta_1$	0.03100	0.04347	0.04783	0.04980	0.05160	0.04463	0.08097
86	$\Theta_1$	0.02800	0.04180	0.04770	0.04953	0.05147	0.04310	0.07530
87	$\Theta_1$	0.02740	0.04273	0.04777	0.04960	0.05140	0.04303	0.07517
88	$\Theta_1$	0.03013	0.04293	0.04777	0.04967	0.05147	0.04417	0.08016
89	$\Theta_1$	0.02753	0.04173	0.04770	0.04947	0.05147	0.04310	0.07533
90	$\Theta_1$	0.02813	0.04293	0.04770	0.04940	0.05140	0.04310	0.07503
91	$\Theta_1$	0.02820	0.04180	0.04777	0.04960	0.05140	0.04310	0.07528
92	$\Theta_1$	0.03060	0.04313	0.04777	0.04973	0.05147	0.04423	0.07964
93	$\Theta_1$	0.02753	0.04253	0.04763	0.04947	0.05147	0.04297	0.07522
94	$\Theta_1$	0.02993	0.04313	0.04770	0.04967	0.05153	0.04437	0.08146
95	$\Theta_1$	0.03047	0.04320	0.04783	0.04967	0.05160	0.04443	0.07958
96	$\Theta_1$	0.02800	0.04287	0.04777	0.04947	0.05147	0.04303	0.07527
97	$\Theta_1$	0.02900	0.04340	0.04770	0.04940	0.05147	0.04357	0.07659
98	$\Theta_1$	0.03127	0.04340	0.04777	0.04967	0.05153	0.04463	0.08104
99	$\Theta_1$	0.03053	0.04327	0.04777	0.04967	0.05160	0.04450	0.08089
100	$\Theta_1$	0.02773	0.04187	0.04770	0.04960	0.05147	0.04310	0.07508
All	$\Theta_1$	0.00533	0.00860	0.01017	0.01140	0.01353	0.00990	0.09838

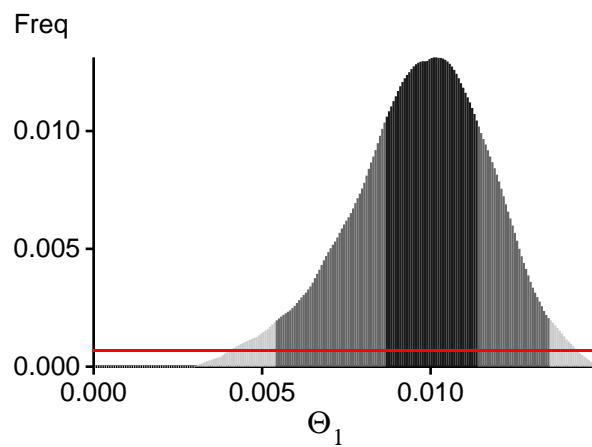
Citation suggestions:

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	-13865.33	-13725.08	-13768.35	-13861.99
2	-14051.39	-13852.39	-13891.48	-13979.47
3	-13866.01	-13725.75	-13768.90	-13862.73
4	-13867.33	-13726.98	-13769.95	-13864.31
5	-13865.82	-13725.68	-13768.99	-13862.74
6	-13866.88	-13726.71	-13769.67	-13863.66
7	-13879.69	-13739.27	-13782.91	-13876.33
8	-14150.68	-13983.31	-14019.51	-14114.74
9	-13962.04	-13810.97	-13857.87	-13945.96
10	-13866.42	-13726.20	-13769.37	-13863.18
11	-13867.87	-13727.63	-13770.65	-13864.62
12	-27633.35	-19829.85	-18479.09	-18562.23
13	-13901.71	-13750.45	-13794.34	-13885.73
14	-13902.62	-13751.98	-13795.83	-13887.28
15	-13864.75	-13724.47	-13767.52	-13862.19
16	-13865.71	-13725.46	-13768.04	-13862.56
17	-13865.61	-13725.41	-13768.78	-13862.40
18	-13883.52	-13741.31	-13785.56	-13881.08
19	-14141.61	-13910.80	-13945.25	-14032.50
20	-13867.53	-13727.18	-13770.12	-13864.20
21	-13867.88	-13727.71	-13770.99	-13864.66
22	-13912.29	-13764.37	-13809.56	-13899.68
23	-13867.44	-13727.18	-13770.09	-13864.12
24	-13866.47	-13726.23	-13768.07	-13863.26
25	-13867.82	-13727.51	-13770.43	-13864.44
26	-13864.41	-13724.17	-13767.47	-13861.10
27	-13867.80	-13727.62	-13770.66	-13865.41
28	-13864.37	-13724.17	-13767.27	-13861.20
29	-13867.93	-13727.69	-13771.10	-13865.19

30	-13884.75	-13741.29	-13784.94	-13876.83
31	-13867.75	-13727.49	-13770.92	-13864.54
32	-13867.59	-13727.32	-13770.16	-13864.29
33	-13865.82	-13725.59	-13768.64	-13862.65
34	-13867.11	-13726.90	-13769.79	-13863.98
35	-13866.46	-13726.23	-13769.79	-13863.27
36	-13899.38	-13754.02	-13799.29	-13892.34
37	-13899.94	-13754.52	-13798.26	-13889.57
38	-13866.56	-13726.29	-13769.54	-13863.29
39	-13867.25	-13726.97	-13769.32	-13863.88
40	-13867.40	-13727.10	-13770.67	-13864.36
41	-13867.45	-13727.11	-13770.21	-13864.10
42	-13937.24	-13791.76	-13839.05	-13928.56
43	-13866.37	-13726.21	-13769.15	-13863.17
44	-13866.99	-13726.62	-13769.95	-13863.89
45	-13866.78	-13726.56	-13769.94	-13863.57
46	-13867.51	-13727.07	-13769.83	-13863.99
47	-13867.35	-13727.00	-13770.15	-13864.11
48	-13866.87	-13726.57	-13769.79	-13863.71
49	-13867.35	-13727.06	-13770.28	-13864.02
50	-13867.70	-13727.33	-13770.70	-13864.62
51	-13896.35	-13754.31	-13798.32	-13893.39
52	-13863.83	-13723.64	-13766.75	-13860.59
53	-13867.67	-13727.46	-13771.11	-13864.43
54	-13891.82	-13751.14	-13797.91	-13888.21
55	-13867.98	-13727.66	-13767.05	-13864.64
56	-13926.90	-13772.64	-13818.34	-13907.18
57	-13866.63	-13726.35	-13769.92	-13863.27
58	-21206.13	-19113.09	-18854.86	-18937.78
59	-16188.33	-14957.53	-14813.59	-14898.86
60	-13864.35	-13724.17	-13762.69	-13861.17
61	-13886.92	-13746.33	-13790.34	-13883.83
62	-13867.59	-13727.31	-13770.45	-13864.38
63	-13867.76	-13727.41	-13770.50	-13864.33
64	-13892.63	-13751.91	-13796.00	-13889.08
65	-13866.84	-13726.65	-13770.23	-13863.60
66	-14035.19	-13887.97	-13930.86	-14022.88
67	-13910.01	-13758.51	-13802.65	-13893.73
68	-13901.40	-13750.99	-13795.16	-13886.25
69	-13862.41	-13722.21	-13764.52	-13860.60
70	-13867.24	-13726.98	-13770.44	-13864.00
71	-13865.88	-13725.61	-13769.10	-13862.91
72	-13880.81	-13740.37	-13785.40	-13878.15
73	-13910.89	-13765.30	-13810.70	-13901.94
74	-13865.84	-13725.54	-13768.41	-13862.52

75	-13925.32	-13768.91	-13812.90	-13904.32
76	-13867.72	-13727.33	-13770.96	-13864.65
77	-13963.95	-13791.35	-13833.45	-13922.97
78	-13882.14	-13738.65	-13782.44	-13876.27
79	-13906.18	-13765.34	-13812.93	-13902.84
80	-13865.01	-13724.74	-13768.34	-13861.63
81	-13928.87	-13779.03	-13825.81	-13913.94
82	-13866.46	-13726.11	-13769.02	-13863.12
83	-13897.38	-13749.79	-13793.72	-13885.87
84	-13864.21	-13723.87	-13766.88	-13860.79
85	-14306.61	-14046.32	-14079.91	-14163.92
86	-13867.60	-13727.24	-13770.64	-13864.16
87	-13867.98	-13727.62	-13770.39	-13864.85
88	-13937.69	-13792.76	-13838.04	-13929.21
89	-13867.76	-13727.45	-13770.46	-13864.44
90	-13867.48	-13727.26	-13770.55	-13864.27
91	-13866.17	-13725.88	-13769.13	-13862.77
92	-13932.06	-13779.68	-13826.80	-13914.53
93	-13866.58	-13726.35	-13769.60	-13863.76
94	-14686.37	-14405.20	-14432.54	-14520.42
95	-14131.94	-13925.79	-13965.17	-14052.53
96	-13865.72	-13725.45	-13767.11	-13862.49
97	-13885.41	-13740.89	-13784.18	-13877.26
98	-14105.65	-13912.53	-13956.38	-14041.70
99	-21173.20	-17215.55	-16573.97	-16658.25
100	-13867.79	-13727.49	-13770.30	-13865.06
All	-1421144.58	-1391836.19	-1393571.10	-1402821.46
(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 = 84.002055]				
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
$\Theta_1$	386228945/399978997	0.96562
Genealogies	1119762155/1600021003	0.69984

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

Parameter	Autocorrelation	Effective Sampe Size
$\Theta_1$	0.60286	2480809.10
Genealogies	0.04313	9265333.21

## *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