

# 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 Sun Jul 23 19:32:43 2017

Program finished at Sun Jul 23 22:37:34 2017 [Runtime:0000:03:04:51]



## 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)

3546959936

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.0.9
Haplotyping is turned on:	NO
Output file:	outfile_0.9_0.5
Posterior distribution raw histogram file:	bayesfile
Raw data from the MCMC run:	bayesallfile_0.9_0.5
Print data:	No
Print genealogies [only some for some data type]:	None

## Data summary

Data file: infile.0.9  
 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]
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]
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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]
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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]
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56	1	Jukes-Cantor	[Basefreq: =0.25]
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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]
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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
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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
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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	
			14		10	
			15		10	
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			31		10	
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			38		10	
			39		10	
			40		10	

41	10
42	10
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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
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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	$\Theta_1$	0.03227	0.04407	0.04797	0.04993	0.05167	0.04523	0.08361
2	$\Theta_1$	0.03327	0.04413	0.04797	0.04980	0.05153	0.04537	0.08440
3	$\Theta_1$	0.03140	0.04347	0.04777	0.04960	0.05153	0.04477	0.08230
4	$\Theta_1$	0.03407	0.04447	0.04790	0.04980	0.05153	0.04563	0.08547
5	$\Theta_1$	0.03240	0.04373	0.04777	0.04967	0.05160	0.04497	0.08370
6	$\Theta_1$	0.03113	0.04347	0.04790	0.04980	0.05153	0.04463	0.08233
7	$\Theta_1$	0.03207	0.04387	0.04783	0.04973	0.05153	0.04510	0.08293
8	$\Theta_1$	0.03367	0.04433	0.04790	0.04967	0.05167	0.04563	0.08619
9	$\Theta_1$	0.03147	0.03147	0.04790	0.05160	0.05160	0.04497	0.08293
10	$\Theta_1$	0.03387	0.04540	0.04810	0.04980	0.05167	0.04557	0.08567
11	$\Theta_1$	0.03560	0.04587	0.04790	0.04933	0.05173	0.04603	0.08669
12	$\Theta_1$	0.03340	0.04400	0.04783	0.04973	0.05153	0.04497	0.08444
13	$\Theta_1$	0.03287	0.04420	0.04783	0.04973	0.05160	0.04543	0.08554
14	$\Theta_1$	0.03247	0.04493	0.04797	0.04967	0.05153	0.04510	0.08313
15	$\Theta_1$	0.03107	0.04427	0.04783	0.04947	0.05153	0.04450	0.08205
16	$\Theta_1$	0.02960	0.04253	0.04770	0.04960	0.05147	0.04383	0.07969
17	$\Theta_1$	0.03340	0.04513	0.04770	0.04920	0.05153	0.04530	0.08364
18	$\Theta_1$	0.03080	0.04420	0.04790	0.04960	0.05160	0.04450	0.08232

19	$\Theta_1$	0.03400	0.04453	0.04803	0.04993	0.05173	0.04570	0.08715
20	$\Theta_1$	0.03320	0.04460	0.04823	0.05013	0.05173	0.04563	0.08471
21	$\Theta_1$	0.03313	0.04433	0.04810	0.04993	0.05173	0.04550	0.08609
22	$\Theta_1$	0.03313	0.04387	0.04790	0.04980	0.05147	0.04510	0.08256
23	$\Theta_1$	0.03447	0.04453	0.04803	0.05000	0.05167	0.04563	0.08494
24	$\Theta_1$	0.03387	0.04553	0.04797	0.04953	0.05160	0.04570	0.08569
25	$\Theta_1$	0.03260	0.04407	0.04803	0.05000	0.05167	0.04517	0.08428
26	$\Theta_1$	0.03367	0.04440	0.04797	0.04987	0.05153	0.04550	0.08618
27	$\Theta_1$	0.03493	0.04493	0.04803	0.05000	0.05180	0.04603	0.08646
28	$\Theta_1$	0.03333	0.04420	0.04783	0.04973	0.05160	0.04537	0.08370
29	$\Theta_1$	0.03213	0.04487	0.04797	0.04967	0.05173	0.04503	0.08397
30	$\Theta_1$	0.03060	0.04327	0.04783	0.04973	0.05167	0.04450	0.07997
31	$\Theta_1$	0.03207	0.04367	0.04777	0.04967	0.05153	0.04483	0.08321
32	$\Theta_1$	0.03233	0.04367	0.04783	0.04973	0.05167	0.04490	0.08328
33	$\Theta_1$	0.03293	0.04440	0.04783	0.04993	0.05167	0.04550	0.08493
34	$\Theta_1$	0.03153	0.04347	0.04783	0.04973	0.05167	0.04470	0.08305
35	$\Theta_1$	0.03360	0.04467	0.04797	0.05000	0.05167	0.04577	0.08712
36	$\Theta_1$	0.03180	0.04327	0.04777	0.04953	0.05160	0.04457	0.08268
37	$\Theta_1$	0.03053	0.04340	0.04783	0.04980	0.05167	0.04457	0.08144
38	$\Theta_1$	0.03360	0.04447	0.04810	0.04980	0.05167	0.04563	0.08584
39	$\Theta_1$	0.03233	0.04427	0.04797	0.04987	0.05173	0.04537	0.08453
40	$\Theta_1$	0.03093	0.04407	0.04783	0.04953	0.05147	0.04443	0.08039
41	$\Theta_1$	0.03233	0.04380	0.04777	0.04953	0.05167	0.04510	0.08193



Locus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
42	$\Theta_1$	0.03267	0.04427	0.04783	0.04980	0.05173	0.04543	0.08424
43	$\Theta_1$	0.03073	0.04373	0.04783	0.04980	0.05167	0.04490	0.08233
44	$\Theta_1$	0.03180	0.04473	0.04797	0.04953	0.05153	0.04490	0.08239
45	$\Theta_1$	0.03253	0.04393	0.04797	0.04987	0.05173	0.04510	0.08221
46	$\Theta_1$	0.03393	0.04460	0.04777	0.04980	0.05160	0.04570	0.08791
47	$\Theta_1$	0.03127	0.04353	0.04777	0.04967	0.05153	0.04477	0.08189
48	$\Theta_1$	0.03307	0.04420	0.04810	0.05000	0.05167	0.04530	0.08557
49	$\Theta_1$	0.03153	0.04347	0.04777	0.04953	0.05167	0.04477	0.08507
50	$\Theta_1$	0.03007	0.04293	0.04777	0.04973	0.05153	0.04417	0.07956
51	$\Theta_1$	0.03460	0.04440	0.04790	0.04973	0.05167	0.04557	0.08598
52	$\Theta_1$	0.03147	0.04413	0.04790	0.04980	0.05173	0.04530	0.08383
53	$\Theta_1$	0.03360	0.04453	0.04790	0.04987	0.05160	0.04563	0.08577
54	$\Theta_1$	0.03147	0.04447	0.04783	0.04940	0.05160	0.04463	0.08263
55	$\Theta_1$	0.03387	0.04413	0.04783	0.04967	0.05180	0.04537	0.08600
56	$\Theta_1$	0.03333	0.04393	0.04790	0.04973	0.05167	0.04517	0.08497
57	$\Theta_1$	0.03260	0.04420	0.04803	0.04980	0.05180	0.04543	0.08560
58	$\Theta_1$	0.03433	0.04473	0.04797	0.04987	0.05160	0.04590	0.08657
59	$\Theta_1$	0.03140	0.04453	0.04790	0.04967	0.05160	0.04470	0.08173
60	$\Theta_1$	0.03353	0.04433	0.04770	0.04960	0.05167	0.04557	0.08486
61	$\Theta_1$	0.03107	0.04320	0.04783	0.04960	0.05153	0.04450	0.08121

62	$\Theta_1$	0.03300	0.04427	0.04797	0.04993	0.05167	0.04543	0.08472
63	$\Theta_1$	0.03267	0.04413	0.04783	0.04973	0.05160	0.04530	0.08482
64	$\Theta_1$	0.03440	0.04453	0.04790	0.04980	0.05167	0.04570	0.08538
65	$\Theta_1$	0.03313	0.04420	0.04790	0.04993	0.05160	0.04530	0.08438
66	$\Theta_1$	0.03173	0.04387	0.04777	0.04980	0.05167	0.04503	0.08305
67	$\Theta_1$	0.03140	0.04460	0.04777	0.04927	0.05153	0.04477	0.08252
68	$\Theta_1$	0.03273	0.04433	0.04777	0.04973	0.05160	0.04550	0.08407
69	$\Theta_1$	0.03200	0.04473	0.04777	0.04953	0.05167	0.04503	0.08475
70	$\Theta_1$	0.03373	0.04460	0.04797	0.04987	0.05160	0.04577	0.08620
71	$\Theta_1$	0.02900	0.04320	0.04777	0.04967	0.05173	0.04437	0.07967
72	$\Theta_1$	0.03260	0.04380	0.04790	0.04987	0.05147	0.04490	0.08323
73	$\Theta_1$	0.03307	0.04433	0.04790	0.04987	0.05160	0.04550	0.08513
74	$\Theta_1$	0.03280	0.04387	0.04783	0.04953	0.05167	0.04517	0.08358
75	$\Theta_1$	0.03240	0.04427	0.04790	0.04980	0.05167	0.04537	0.08366
76	$\Theta_1$	0.03187	0.04387	0.04777	0.04973	0.05173	0.04503	0.08262
77	$\Theta_1$	0.03367	0.04427	0.04797	0.04987	0.05153	0.04543	0.08412
78	$\Theta_1$	0.03387	0.04473	0.04810	0.04993	0.05173	0.04590	0.08594
79	$\Theta_1$	0.03273	0.04373	0.04797	0.04973	0.05160	0.04497	0.08437
80	$\Theta_1$	0.03020	0.04307	0.04783	0.04973	0.05160	0.04430	0.08007
81	$\Theta_1$	0.03193	0.04380	0.04797	0.04987	0.05173	0.04497	0.08328
82	$\Theta_1$	0.03347	0.04420	0.04797	0.04973	0.05167	0.04543	0.08478
83	$\Theta_1$	0.03527	0.04487	0.04797	0.04980	0.05160	0.04597	0.08735
84	$\Theta_1$	0.03160	0.04333	0.04783	0.04967	0.05147	0.04457	0.08118

Locus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
85	$\Theta_1$	0.03473	0.04473	0.04803	0.04987	0.05153	0.04583	0.08709
86	$\Theta_1$	0.03213	0.04087	0.04777	0.05053	0.05167	0.04503	0.08404
87	$\Theta_1$	0.03267	0.04407	0.04790	0.04973	0.05167	0.04523	0.08545
88	$\Theta_1$	0.03320	0.04413	0.04777	0.04960	0.05160	0.04537	0.08538
89	$\Theta_1$	0.03207	0.04393	0.04783	0.04980	0.05167	0.04510	0.08354
90	$\Theta_1$	0.03567	0.04487	0.04803	0.04987	0.05160	0.04597	0.08694
91	$\Theta_1$	0.03407	0.04533	0.04783	0.04940	0.05160	0.04550	0.08504
92	$\Theta_1$	0.03227	0.04407	0.04797	0.04993	0.05167	0.04517	0.08365
93	$\Theta_1$	0.03467	0.04427	0.04790	0.04973	0.05160	0.04543	0.08610
94	$\Theta_1$	0.03027	0.04293	0.04783	0.04973	0.05153	0.04417	0.08077
95	$\Theta_1$	0.03240	0.04387	0.04783	0.04973	0.05160	0.04503	0.08237
96	$\Theta_1$	0.03267	0.04387	0.04797	0.04973	0.05167	0.04510	0.08357
97	$\Theta_1$	0.03087	0.04327	0.04763	0.04953	0.05153	0.04457	0.08045
98	$\Theta_1$	0.03233	0.04307	0.04783	0.04980	0.05153	0.04477	0.08262
99	$\Theta_1$	0.03213	0.04373	0.04777	0.04967	0.05160	0.04497	0.08249
100	$\Theta_1$	0.03247	0.04460	0.04790	0.04967	0.05153	0.04477	0.08534
All	$\Theta_1$	0.01020	0.01547	0.01570	0.01613	0.01967	0.01563	0.09965

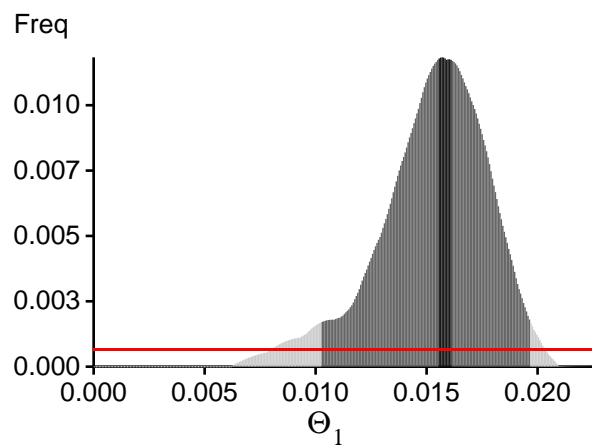
Citation suggestions:

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	-15407.27	-14890.56	-14904.13	-14967.23
2	-15069.68	-14600.95	-14623.63	-14683.36
3	-15121.07	-14644.98	-14658.73	-14723.31
4	-16073.49	-15562.88	-15591.46	-15644.19
5	-14699.72	-14420.13	-14478.48	-14537.75
6	-16011.29	-15160.62	-15112.11	-15176.46
7	-15620.13	-14920.43	-14900.33	-14962.74
8	-23907.45	-19848.40	-19243.83	-19299.63
9	-14783.68	-14498.78	-14553.18	-14617.18
10	-15581.00	-14951.02	-14949.91	-15007.18
11	-16538.70	-15989.60	-16014.36	-16068.80
12	-16065.21	-15587.46	-15618.75	-15678.95
13	-15732.18	-15190.96	-15210.13	-15265.62
14	-15103.49	-14687.46	-14716.84	-14779.96
15	-17238.66	-16283.16	-16235.28	-16292.43
16	-14676.79	-14400.31	-14454.46	-14517.84
17	-14656.88	-14401.65	-14460.66	-14520.74
18	-14981.05	-14572.93	-14600.71	-14666.06
19	-16253.57	-15656.71	-15676.63	-15725.43
20	-16081.36	-15256.12	-15220.05	-15277.03
21	-15665.22	-15079.44	-15089.76	-15144.14
22	-14712.93	-14440.25	-14496.08	-14558.44
23	-15496.41	-14900.25	-14901.62	-14961.08
24	-15059.11	-14629.63	-14662.82	-14718.60
25	-14709.10	-14433.45	-14491.10	-14550.65
26	-21510.77	-18581.20	-18180.46	-18234.12
27	-15669.33	-15011.82	-15009.69	-15062.58
28	-15102.60	-14706.17	-14740.90	-14802.15
29	-15777.47	-15285.54	-15316.24	-15369.15

30	-15358.62	-14710.43	-14692.28	-14760.40
31	-17519.45	-16386.63	-16287.91	-16355.33
32	-15207.89	-14770.90	-14799.46	-14861.34
33	-16166.18	-15437.79	-15422.58	-15478.88
34	-14751.97	-14438.55	-14488.59	-14550.04
35	-17908.56	-16413.26	-16261.72	-16319.82
36	-15395.04	-14928.30	-14958.30	-15015.82
37	-14629.22	-14390.15	-14451.64	-14513.92
38	-16297.31	-15697.73	-15713.80	-15767.24
39	-17101.96	-16303.84	-16290.64	-16343.48
40	-15027.04	-14628.53	-14660.64	-14726.09
41	-14726.62	-14347.42	-14381.42	-14444.05
42	-16388.92	-15567.21	-14802.21	-15593.57
43	-14952.54	-14584.89	-14622.68	-14687.44
44	-15015.90	-14653.43	-14684.52	-14753.22
45	-14580.49	-14320.98	-14373.69	-14442.40
46	-20148.51	-18071.92	-15135.11	-17884.04
47	-14454.42	-14197.08	-14252.78	-14316.92
48	-15946.40	-15307.52	-14709.47	-15369.08
49	-18407.05	-17194.54	-14915.18	-17161.36
50	-14650.26	-14380.85	-14431.00	-14498.40
51	-17128.88	-15829.29	-14503.55	-15766.86
52	-14768.81	-14449.59	-14499.93	-14558.60
53	-15611.49	-15031.45	-14472.04	-15096.75
54	-14677.25	-14366.64	-14416.09	-14476.64
55	-15932.32	-15350.70	-14497.78	-15420.26
56	-15271.16	-14882.21	-14628.90	-14979.79
57	-16006.92	-15375.53	-14464.39	-15438.05
58	-16610.28	-15522.88	-14496.16	-15498.04
59	-15805.95	-14911.21	-14608.88	-14915.62
60	-15978.99	-15185.30	-14480.85	-15213.23
61	-14501.11	-14240.06	-14293.59	-14357.86
62	-16678.08	-15770.90	-15726.35	-15783.14
63	-16197.51	-15692.34	-15211.07	-15778.57
64	-15689.79	-15069.89	-15010.90	-15126.27
65	-15838.71	-15304.78	-15228.54	-15383.73
66	-15967.40	-15399.97	-14557.66	-15474.82
67	-15772.93	-15024.88	-14949.33	-15059.67
68	-22557.36	-19206.95	-14958.54	-18790.12
69	-15489.08	-15070.29	-15112.14	-15168.13
70	-15701.39	-15134.41	-15151.34	-15203.64
71	-14552.24	-14331.28	-14390.55	-14459.38
72	-17707.69	-16861.16	-16289.02	-16893.19
73	-16100.02	-15506.57	-15523.97	-15576.62
74	-14959.37	-14557.83	-14593.95	-14653.54

75	-16354.46	-15499.22	-15448.36	-15521.60
76	-15029.67	-14603.80	-14632.36	-14694.53
77	-15211.19	-14667.85	-14673.35	-14734.67
78	-15838.08	-15035.22	-15002.69	-15060.83
79	-15258.07	-14848.86	-14884.93	-14943.78
80	-14361.83	-14130.09	-14186.83	-14252.60
81	-14983.77	-14658.93	-14383.35	-14770.43
82	-17523.14	-16151.84	-14266.44	-16075.86
83	-15748.48	-15293.05	-14869.81	-15393.78
84	-14538.28	-14275.36	-14327.31	-14393.15
85	-16718.84	-15785.43	-14301.01	-15791.19
86	-14948.69	-14629.27	-14440.69	-14739.88
87	-19563.64	-17795.10	-14428.12	-17665.49
88	-16967.18	-15971.58	-14701.57	-15967.44
89	-14975.55	-14551.61	-14580.63	-14642.38
90	-16370.41	-15566.20	-15542.61	-15595.12
91	-15271.43	-14929.32	-14630.19	-15039.78
92	-14826.57	-14484.57	-14502.99	-14590.77
93	-16499.57	-15390.10	-15304.11	-15360.01
94	-14547.75	-14293.90	-14348.48	-14414.84
95	-15007.14	-14512.90	-14525.69	-14589.45
96	-15164.93	-14650.54	-14663.95	-14726.05
97	-14993.62	-14608.82	-14642.12	-14706.78
98	-15410.87	-14875.83	-14886.89	-14950.97
99	-14819.82	-14402.26	-14399.91	-14493.51
100	-15607.97	-15239.83	-15004.21	-15346.43
All	-1589844.04	-1522110.80	-1493950.40	-1527047.91
(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 = 141.627018]				
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
$\Theta_1$	376331569/400010157	0.94081
Genealogies	108453735/1599989843	0.06778

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

Parameter	Autocorrelation	Effective Sample Size
$\Theta_1$	0.49023	8880111.36
Genealogies	0.37747	12095087.33

## *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