

# 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 39 compute nodes are available.

Program started at Tue Aug 15 06:29:08 2017

Program finished at Tue Aug 15 14:01:01 2017 [Runtime:0000:07:31:53]



## 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)

247724266

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

4 chains with temperatures

1000000.00

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]
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40	1	Jukes-Cantor	[Basefreq: =0.25]
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46	1	Jukes-Cantor	[Basefreq: =0.25]
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50	1	Jukes-Cantor	[Basefreq: =0.25]
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73	1	Jukes-Cantor	[Basefreq: =0.25]
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75	1	Jukes-Cantor	[Basefreq: =0.25]
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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]
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86	1	Jukes-Cantor	[Basefreq: =0.25]
87	1	Jukes-Cantor	[Basefreq: =0.25]
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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
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31	10000
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37	10000
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81	10000
82	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	
			14		10	
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41	10
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	86	10
	87	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
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
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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	$\Theta_1$	0.03380	0.04433	0.04803	0.04993	0.05173	0.04543	0.08573
2	$\Theta_1$	0.03173	0.04373	0.04783	0.04980	0.05153	0.04490	0.08340
3	$\Theta_1$	0.03167	0.04360	0.04777	0.04973	0.05153	0.04477	0.08307
4	$\Theta_1$	0.03187	0.04360	0.04783	0.04980	0.05153	0.04477	0.08132
5	$\Theta_1$	0.03333	0.04447	0.04790	0.04973	0.05160	0.04563	0.08512
6	$\Theta_1$	0.03340	0.04420	0.04790	0.04973	0.05167	0.04543	0.08540
7	$\Theta_1$	0.03240	0.04447	0.04803	0.05007	0.05173	0.04557	0.08537
8	$\Theta_1$	0.03200	0.04380	0.04790	0.04987	0.05160	0.04497	0.08458
9	$\Theta_1$	0.03273	0.04433	0.04783	0.04987	0.05153	0.04543	0.08456
10	$\Theta_1$	0.03347	0.04427	0.04790	0.04980	0.05167	0.04543	0.08390
11	$\Theta_1$	0.03527	0.04473	0.04797	0.04980	0.05153	0.04583	0.08697
12	$\Theta_1$	0.03353	0.04413	0.04790	0.04980	0.05173	0.04537	0.08512
13	$\Theta_1$	0.03487	0.04453	0.04810	0.05007	0.05173	0.04563	0.08698
14	$\Theta_1$	0.03233	0.04380	0.04783	0.04987	0.05153	0.04490	0.08370
15	$\Theta_1$	0.03347	0.04553	0.04783	0.04933	0.05160	0.04570	0.08597
16	$\Theta_1$	0.03380	0.04520	0.04790	0.04960	0.05153	0.04537	0.08565
17	$\Theta_1$	0.03087	0.04333	0.04783	0.04973	0.05160	0.04457	0.08128
18	$\Theta_1$	0.03100	0.04347	0.04770	0.04973	0.05147	0.04463	0.08241

19	$\Theta_1$	0.03287	0.04387	0.04777	0.04967	0.05147	0.04503	0.08520
20	$\Theta_1$	0.03273	0.04407	0.04783	0.04980	0.05160	0.04523	0.08465
21	$\Theta_1$	0.03193	0.04373	0.04783	0.04980	0.05160	0.04490	0.08263
22	$\Theta_1$	0.03513	0.04493	0.04797	0.04993	0.05160	0.04603	0.08726
23	$\Theta_1$	0.03347	0.04527	0.04797	0.04953	0.05160	0.04550	0.08548
24	$\Theta_1$	0.03233	0.04407	0.04790	0.04980	0.05167	0.04523	0.08342
25	$\Theta_1$	0.03367	0.04440	0.04777	0.04967	0.05160	0.04557	0.08642
26	$\Theta_1$	0.03200	0.04360	0.04797	0.04980	0.05167	0.04483	0.08177
27	$\Theta_1$	0.03347	0.04533	0.04790	0.04947	0.05167	0.04557	0.08509
28	$\Theta_1$	0.03213	0.04407	0.04797	0.04993	0.05167	0.04517	0.08539
29	$\Theta_1$	0.03367	0.04473	0.04797	0.04987	0.05180	0.04590	0.08680
30	$\Theta_1$	0.02947	0.04280	0.04777	0.04973	0.05153	0.04403	0.08036
31	$\Theta_1$	0.03200	0.04373	0.04797	0.04987	0.05167	0.04490	0.08310
32	$\Theta_1$	0.03467	0.04553	0.04790	0.04933	0.05167	0.04577	0.08573
33	$\Theta_1$	0.03453	0.04433	0.04783	0.04960	0.05167	0.04557	0.08554
34	$\Theta_1$	0.03253	0.04387	0.04803	0.04973	0.05173	0.04517	0.08420
35	$\Theta_1$	0.03193	0.04373	0.04790	0.04993	0.05153	0.04483	0.08252
36	$\Theta_1$	0.02940	0.04313	0.04777	0.04967	0.05160	0.04437	0.08303
37	$\Theta_1$	0.03547	0.04473	0.04797	0.04980	0.05173	0.04590	0.08639
38	$\Theta_1$	0.03213	0.04393	0.04797	0.05000	0.05167	0.04503	0.08290
39	$\Theta_1$	0.03193	0.04367	0.04777	0.04980	0.05160	0.04483	0.08258
40	$\Theta_1$	0.03567	0.04507	0.04790	0.04980	0.05173	0.04623	0.08702
41	$\Theta_1$	0.03493	0.04487	0.04810	0.05007	0.05160	0.04597	0.08729



Locus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
42	$\Theta_1$	0.03253	0.04400	0.04797	0.04980	0.05160	0.04517	0.08412
43	$\Theta_1$	0.03147	0.04467	0.04790	0.04953	0.05160	0.04483	0.08329
44	$\Theta_1$	0.03207	0.04373	0.04783	0.04960	0.05153	0.04497	0.08389
45	$\Theta_1$	0.02980	0.04280	0.04770	0.04967	0.05153	0.04403	0.08041
46	$\Theta_1$	0.03220	0.04360	0.04777	0.04967	0.05160	0.04483	0.08352
47	$\Theta_1$	0.03427	0.04460	0.04797	0.04987	0.05160	0.04577	0.08577
48	$\Theta_1$	0.03253	0.04373	0.04777	0.04973	0.05153	0.04490	0.08257
49	$\Theta_1$	0.03053	0.04333	0.04790	0.04980	0.05167	0.04450	0.08131
50	$\Theta_1$	0.03133	0.04327	0.04777	0.04967	0.05153	0.04450	0.08175
51	$\Theta_1$	0.03400	0.04433	0.04797	0.04987	0.05167	0.04550	0.08594
52	$\Theta_1$	0.02867	0.04320	0.04783	0.04947	0.05153	0.04350	0.07790
53	$\Theta_1$	0.03140	0.04380	0.04790	0.04987	0.05160	0.04497	0.08214
54	$\Theta_1$	0.03247	0.04367	0.04777	0.04967	0.05160	0.04490	0.08338
55	$\Theta_1$	0.03360	0.04447	0.04803	0.04993	0.05167	0.04557	0.08551
56	$\Theta_1$	0.03313	0.04427	0.04803	0.04987	0.05160	0.04543	0.08498
57	$\Theta_1$	0.03500	0.04473	0.04777	0.04973	0.05147	0.04590	0.08564
58	$\Theta_1$	0.02853	0.04227	0.04777	0.04967	0.05153	0.04350	0.07669
59	$\Theta_1$	0.03220	0.04367	0.04777	0.04967	0.05153	0.04490	0.08271
60	$\Theta_1$	0.03220	0.04460	0.04797	0.04973	0.05167	0.04497	0.08327
61	$\Theta_1$	0.03113	0.04373	0.04783	0.04993	0.05167	0.04483	0.08235

62	$\Theta_1$	0.03427	0.04467	0.04803	0.05007	0.05167	0.04570	0.08486
63	$\Theta_1$	0.03327	0.04440	0.04790	0.04973	0.05167	0.04557	0.08559
64	$\Theta_1$	0.03347	0.04447	0.04803	0.05000	0.05160	0.04557	0.08547
65	$\Theta_1$	0.03253	0.04407	0.04790	0.04987	0.05153	0.04523	0.08442
66	$\Theta_1$	0.03273	0.04420	0.04797	0.04993	0.05167	0.04530	0.08422
67	$\Theta_1$	0.03333	0.04420	0.04797	0.04987	0.05167	0.04537	0.08494
68	$\Theta_1$	0.03227	0.04413	0.04790	0.04973	0.05167	0.04530	0.08448
69	$\Theta_1$	0.03320	0.04440	0.04783	0.04993	0.05167	0.04550	0.08482
70	$\Theta_1$	0.03313	0.04480	0.04790	0.04980	0.05153	0.04537	0.08435
71	$\Theta_1$	0.03400	0.04473	0.04803	0.04993	0.05173	0.04583	0.08659
72	$\Theta_1$	0.02867	0.04327	0.04770	0.04933	0.05140	0.04343	0.07730
73	$\Theta_1$	0.02893	0.04253	0.04777	0.04967	0.05147	0.04377	0.07770
74	$\Theta_1$	0.03133	0.04433	0.04777	0.04920	0.05153	0.04450	0.08201
75	$\Theta_1$	0.03287	0.04407	0.04783	0.04980	0.05153	0.04523	0.08533
76	$\Theta_1$	0.02940	0.04280	0.04783	0.04973	0.05153	0.04403	0.08027
77	$\Theta_1$	0.03020	0.04300	0.04777	0.04967	0.05153	0.04430	0.07961
78	$\Theta_1$	0.03000	0.04300	0.04777	0.04973	0.05153	0.04423	0.08150
79	$\Theta_1$	0.03060	0.04340	0.04783	0.04993	0.05160	0.04450	0.08202
80	$\Theta_1$	0.03453	0.04420	0.04797	0.04967	0.05173	0.04543	0.08552
81	$\Theta_1$	0.03253	0.04400	0.04790	0.04980	0.05173	0.04523	0.08566
82	$\Theta_1$	0.03013	0.04380	0.04777	0.04960	0.05173	0.04503	0.08386
83	$\Theta_1$	0.03127	0.04460	0.04790	0.04967	0.05153	0.04477	0.08271
84	$\Theta_1$	0.03493	0.04467	0.04797	0.04973	0.05147	0.04583	0.08747

Locus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
85	$\Theta_1$	0.03193	0.04353	0.04797	0.04987	0.05160	0.04470	0.08159
86	$\Theta_1$	0.03247	0.04480	0.04783	0.04947	0.05153	0.04497	0.08348
87	$\Theta_1$	0.03413	0.04500	0.04803	0.05000	0.05173	0.04603	0.08625
88	$\Theta_1$	0.03447	0.04467	0.04797	0.04993	0.05173	0.04577	0.08641
89	$\Theta_1$	0.03367	0.04467	0.04790	0.04973	0.05167	0.04583	0.08532
90	$\Theta_1$	0.03247	0.04413	0.04783	0.04973	0.05160	0.04530	0.08559
91	$\Theta_1$	0.03200	0.04393	0.04777	0.04980	0.05167	0.04510	0.08342
92	$\Theta_1$	0.03300	0.04440	0.04790	0.04993	0.05187	0.04550	0.08578
93	$\Theta_1$	0.03533	0.04513	0.04803	0.05000	0.05160	0.04617	0.08775
94	$\Theta_1$	0.03073	0.04307	0.04783	0.04980	0.05153	0.04423	0.07894
95	$\Theta_1$	0.03187	0.04373	0.04783	0.04967	0.05153	0.04497	0.08244
96	$\Theta_1$	0.03393	0.04500	0.04803	0.05000	0.05180	0.04610	0.08690
97	$\Theta_1$	0.03453	0.04427	0.04797	0.04967	0.05167	0.04557	0.08589
98	$\Theta_1$	0.03353	0.04507	0.04797	0.04953	0.05167	0.04550	0.08552
99	$\Theta_1$	0.03167	0.04407	0.04790	0.04987	0.05160	0.04523	0.08267
100	$\Theta_1$	0.03160	0.04333	0.04777	0.04967	0.05153	0.04457	0.08214
All	$\Theta_1$	0.00967	0.01247	0.01417	0.01613	0.01907	0.01437	0.09974

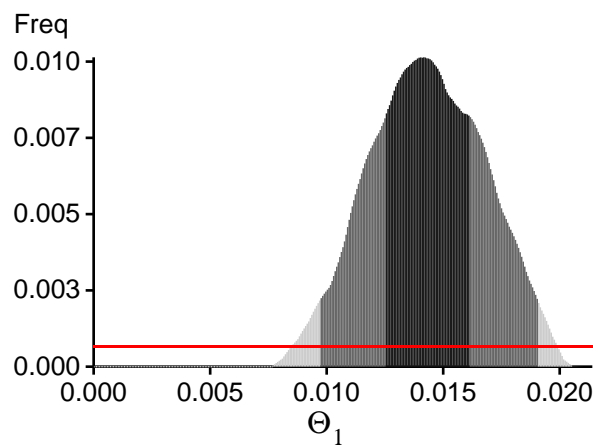
Citation suggestions:

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	-15209.06	-14816.77	-14857.52	-14913.31
2	-15127.64	-14765.22	-14811.94	-14868.55
3	-16442.66	-15803.03	-15806.02	-15865.12
4	-14891.12	-14632.31	-14685.16	-14754.12
5	-15132.45	-14741.12	-14780.95	-14838.90
6	-18451.22	-17205.83	-17119.78	-17170.43
7	-15759.27	-15235.82	-15261.73	-15314.82
8	-15006.52	-14703.35	-14759.87	-14816.14
9	-16660.52	-15566.05	-15480.21	-15539.26
10	-15504.66	-14943.98	-14953.77	-15013.20
11	-16254.98	-15582.59	-15585.02	-15637.38
12	-20068.06	-17565.32	-17234.21	-17290.17
13	-18536.43	-17366.14	-17291.91	-17344.26
14	-15438.07	-14913.06	-14925.01	-14987.32
15	-15374.54	-14889.22	-14916.09	-14971.43
16	-15073.27	-14701.34	-14746.86	-14801.68
17	-14559.03	-14303.36	-14360.80	-14424.30
18	-15191.12	-14716.63	-14731.67	-14797.46
19	-15353.79	-15014.73	-15066.47	-15124.45
20	-14983.40	-14676.45	-14728.65	-14789.59
21	-15081.69	-14675.25	-14709.01	-14771.17
22	-16556.87	-15668.53	-15629.20	-15682.43
23	-15718.49	-15103.83	-15104.02	-15160.76
24	-16731.96	-15764.33	-15703.04	-15764.68
25	-21274.26	-18220.85	-17787.34	-17845.59
26	-14556.88	-14298.22	-14355.20	-14418.76
27	-15299.73	-14865.66	-14900.02	-14956.84
28	-16833.14	-16125.01	-16128.65	-16180.13
29	-17371.24	-16030.27	-15904.03	-15961.54

30	-14652.48	-14399.95	-14457.42	-14520.91
31	-14898.82	-14511.21	-14544.11	-14607.19
32	-15434.60	-14905.69	-14925.39	-14980.36
33	-17125.39	-16242.70	-16209.03	-16266.83
34	-16024.40	-15078.89	-15014.12	-15074.36
35	-15634.86	-15109.23	-15125.36	-15186.76
36	-56664.70	-47704.04	-46390.69	-46478.39
37	-15185.34	-14833.91	-14888.76	-14940.52
38	-14866.11	-14522.14	-14564.86	-14627.86
39	-14669.74	-14406.75	-14464.61	-14525.81
40	-15942.37	-15225.66	-14925.46	-15267.90
41	-16895.79	-16103.39	-15027.16	-16140.88
42	-15036.20	-14640.29	-14550.83	-14738.91
43	-15983.65	-15349.36	-14747.40	-15407.62
44	-15488.30	-14920.52	-14574.23	-14987.95
45	-14810.53	-14426.56	-14354.04	-14525.17
46	-14892.88	-14506.14	-14543.19	-14603.43
47	-15806.16	-15308.27	-14364.34	-15393.92
48	-15093.01	-14677.42	-14463.82	-14771.17
49	-14996.19	-14667.06	-14461.02	-14781.11
50	-14694.99	-14396.00	-14445.68	-14508.66
51	-15718.58	-15049.97	-14718.93	-15101.47
52	-20562.63	-19596.30	-14754.62	-19604.30
53	-14644.03	-14388.79	-14445.89	-14510.44
54	-16374.45	-15349.58	-15273.21	-15335.01
55	-15442.33	-14876.06	-14884.75	-14942.52
56	-17082.05	-15706.75	-14911.33	-15629.96
57	-15115.36	-14712.81	-14740.81	-14807.63
58	-14256.59	-14055.47	-14113.77	-14185.55
59	-14628.12	-14343.62	-14393.84	-14457.19
60	-15081.62	-14661.47	-14693.29	-14752.84
61	-14836.04	-14524.15	-14570.38	-14634.85
62	-16023.14	-15418.05	-15132.98	-15483.32
63	-16550.61	-15678.26	-14817.12	-15698.28
64	-18477.93	-16963.60	-15910.50	-16872.64
65	-15690.68	-15019.47	-15008.01	-15068.44
66	-15047.11	-14627.11	-14661.62	-14719.32
67	-15343.00	-14847.42	-14870.20	-14926.73
68	-15412.44	-14967.14	-14997.87	-15055.41
69	-16084.60	-15200.74	-15152.82	-15211.86
70	-15917.23	-15452.05	-14901.19	-15544.06
71	-16012.71	-15296.82	-14860.18	-15340.95
72	-14333.19	-14116.89	-14170.79	-14242.13
73	-14249.59	-14046.59	-14105.32	-14175.54
74	-16366.12	-15395.44	-15255.22	-15390.93

75	-16640.50	-15640.04	-15574.16	-15630.50
76	-14841.53	-14530.17	-14575.89	-14641.51
77	-14325.29	-14089.76	-14146.43	-14215.15
78	-16237.13	-15398.95	-14465.87	-15420.99
79	-15364.38	-14982.63	-14678.42	-15085.00
80	-15012.65	-14693.65	-14744.65	-14805.67
81	-22535.06	-20412.10	-14554.35	-20243.31
82	-15978.59	-15403.76	-14129.35	-15476.07
83	-14837.57	-14574.99	-14629.13	-14695.51
84	-16279.75	-15619.39	-14462.97	-15675.97
85	-15197.51	-14748.85	-14712.92	-14836.91
86	-14994.27	-14610.87	-14450.86	-14707.50
87	-16429.33	-15358.83	-14397.06	-15333.48
88	-15716.64	-15151.14	-14895.11	-15222.19
89	-15589.22	-15073.84	-14189.26	-15151.40
90	-15989.89	-15371.46	-15222.65	-15432.45
91	-16129.79	-15667.43	-15280.51	-15759.23
92	-15180.01	-14848.30	-14905.32	-14963.27
93	-16209.60	-15535.38	-14584.30	-15587.41
94	-14248.07	-14049.61	-14111.16	-14177.65
95	-14804.97	-14505.84	-14117.81	-14617.46
96	-15816.53	-15296.59	-15050.56	-15377.14
97	-16120.27	-15276.79	-14665.18	-15296.62
98	-15733.25	-15161.89	-15168.86	-15231.16
99	-15207.75	-14699.91	-14715.22	-14778.32
100	-15470.95	-14832.38	-14694.61	-14888.05
All	-1623237.34	-1552516.65	-1523690.98	-1557138.91
(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 = 141.916369]				
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
$\Theta_1$	376128689/399965545	0.94040
Genealogies	108068139/1600034455	0.06754

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

Parameter	Autocorrelation	Effective Sample Size
$\Theta_1$	0.49040	9191370.17
Genealogies	0.39106	12093739.91

## *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