## **AUTO**

POPULATION SIZE, MIGRATION, DIVERGENCE, ASSIGNMENT, HISTORY

Bayesian inference using the structured coalescent

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

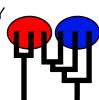
Using Intel AVX (Advanced Vector Extensions)

Compiled for PARALLEL computer architectures

One master and 100 compute nodes are available.

Program started at Sun Aug 13 21:46:26 2017

Program finished at Sun Aug 13 23:07:08 2017 [Runtime:0000:01:20:42]



### **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) 293860571

Start parameters:

Theta values were generated Using a percent value of the prior

M values were generated Using a percent value of the prior

Connection matrix:

m = average (average over a group of Thetas or M,

s = symmetric migration M, S = symmetric 4Nm,

0 = zero, and not estimated,

\* = migration free to vary, Thetas are on diagonal

1

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

Population

1 Romanshorn 0

Order of parameters:

1  $\Theta_1$  <displayed>

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

Analysis strategy:

Bayesian inference

-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 MeanMaximum 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 chains1Recorded steps [a]50000Increment (record every x step [b]200Number 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.1.0

Haplotyping is turned on:

Output file: outfile\_1.0\_0.4

Posterior distribution raw histogram file: bayesfile

Raw data from the MCMC run: bayesallfile\_1.0\_0.4

Print data: No

Print genealogies [only some for some data type]:

None

# Data summary

Data file:

Datatype:

Sequence data

Number of loci:

100

Mutationmodel:	
matation in load.	

iviulation			
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]
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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]
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33	1	Jukes-Cantor	[Basefreq: =0.25]
34	1	Jukes-Cantor	[Basefreq: =0.25]

25	1	Julian Contar	[Decefred: 0.25]
35	1	Jukes-Cantor	[Basefreq: =0.25]
36 37	1	Jukes-Cantor Jukes-Cantor	[Basefreq: =0.25]
38	1 1	Jukes-Cantor	[Basefreq: =0.25] [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]
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45	1	Jukes-Cantor	[Basefreq: =0.25]
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52	1	Jukes-Cantor	[Basefreq: =0.25] [Basefreq: =0.25]
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55 56	1	Jukes-Cantor	[Basefreq: =0.25]
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57 58	1	Jukes-Cantor	[Basefreq: =0.25]
56 59	1	Jukes-Cantor	
60		Jukes-Cantor	[Basefreq: =0.25]
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62	1	Jukes-Cantor	[Basefreq: =0.25]
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			[Basefreq: =0.25]
66 67	1	Jukes-Cantor Jukes-Cantor	[Basefreq: =0.25] [Basefreq: =0.25]
68	1	Jukes-Cantor	[Basefreq: =0.25]
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77 78	1	Jukes-Cantor	[Basefreq: =0.25]
79	1	Jukes-Cantor	[Basefreq: =0.25]
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80	1	Jukes-Cantor	[Basefreq: =0.25]	
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99	1	Jukes-Cantor	[Basefreq: =0.25]	
100	1	Jukes-Cantor	[Basefreq: =0.25]	
Sites per	locus			
Locus		Sites		

Locus	Sites
1	10000
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8	10000
9	10000
10	10000
11	10000
12	10000
13	10000
14	10000
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21	10000	
22	10000	
23	10000	
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	e variation and probab				
Locus S	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		ı	1.000	1.000	Locus	Gene copies
1 Roman					1	10
i Koman	3110111_0				2	10
					3	10
					4	10
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					6	10
					7	10
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41	10
42	10
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85	10

	86	10	
	87	10	
	88	10	
	89	10	
	90	10	
	91	10	
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	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	
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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	$\Theta_1$	0.03453	0.04447	0.04770	0.04960	0.05167	0.04577	0.08823
2	$\Theta_1$	0.03567	0.04480	0.04803	0.04973	0.05167	0.04597	0.08869
3	$\Theta_1$	0.03567	0.04600	0.04823	0.04980	0.05173	0.04630	0.08918
4	$\Theta_1$	0.03647	0.04520	0.04790	0.04987	0.05167	0.04630	0.08933
5	$\Theta_1$	0.03553	0.04540	0.04810	0.05007	0.05160	0.04643	0.08840
6	$\Theta_1$	0.03507	0.04533	0.04817	0.05000	0.05180	0.04637	0.08811
7	$\Theta_1$	0.03913	0.04507	0.04803	0.04987	0.05187	0.04630	0.08919
8	$\Theta_1$	0.03413	0.04493	0.04770	0.04973	0.05173	0.04610	0.08927
9	$\Theta_1$	0.03420	0.04320	0.04810	0.05040	0.05180	0.04577	0.08798
10	$\Theta_1$	0.03407	0.04527	0.04823	0.05013	0.05167	0.04630	0.08803
11	$\Theta_1$	0.03607	0.04520	0.04797	0.04987	0.05153	0.04630	0.08852
12	$\Theta_1$	0.03393	0.04453	0.04803	0.04993	0.05167	0.04570	0.08880
13	$\Theta_1$	0.03513	0.04527	0.04823	0.05020	0.05173	0.04630	0.08825
14	$\Theta_1$	0.03553	0.04513	0.04803	0.04980	0.05173	0.04630	0.08851
15	$\Theta_1$	0.03333	0.04453	0.04777	0.04973	0.05167	0.04570	0.08942
16	$\Theta_1$	0.03487	0.04487	0.04777	0.04967	0.05153	0.04603	0.08851
17	$\Theta_1$	0.03687	0.04607	0.04797	0.04933	0.05153	0.04623	0.08884
18	$\Theta_1$	0.03373	0.04467	0.04790	0.04987	0.05180	0.04577	0.08844

19	$\Theta_1$	0.03373	0.04460	0.04810	0.05007	0.05160	0.04563	0.08756
20	$\Theta_1$	0.03480	0.04527	0.04803	0.05000	0.05173	0.04630	0.08787
21	$\Theta_1$	0.03600	0.04480	0.04783	0.04953	0.05160	0.04603	0.08805
22	$\Theta_1$	0.03467	0.04527	0.04817	0.05000	0.05167	0.04630	0.08769
23	$\Theta_1$	0.03653	0.04533	0.04817	0.05000	0.05173	0.04643	0.08919
24	$\Theta_1$	0.03440	0.04560	0.04790	0.04927	0.05167	0.04597	0.08836
25	$\Theta_1$	0.03687	0.04547	0.04823	0.05013	0.05187	0.04650	0.08909
26	$\Theta_1$	0.03747	0.04527	0.04790	0.04960	0.05153	0.04650	0.08925
27	$\Theta_1$	0.03340	0.04473	0.04817	0.05013	0.05167	0.04577	0.08739
28	$\Theta_1$	0.03633	0.04433	0.04737	0.04913	0.05160	0.04577	0.08922
29	$\Theta_1$	0.03513	0.04500	0.04757	0.04933	0.05140	0.04623	0.08889
30	$\Theta_1$	0.03247	0.04413	0.04797	0.04980	0.05167	0.04537	0.08706
31	$\Theta_1$	0.03527	0.04587	0.04810	0.05013	0.05167	0.04677	0.08937
32	$\Theta_1$	0.03567	0.04493	0.04803	0.04993	0.05153	0.04603	0.08799
33	$\Theta_1$	0.03733	0.04520	0.04797	0.04967	0.05167	0.04643	0.08931
34	$\Theta_1$	0.03540	0.04613	0.04843	0.05047	0.05193	0.04703	0.08989
35	$\Theta_1$	0.03540	0.04460	0.04777	0.04973	0.05160	0.04577	0.08773
36	$\Theta_1$	0.03500	0.04473	0.04797	0.05007	0.05173	0.04583	0.08875
37	$\Theta_1$	0.03367	0.04493	0.04810	0.05013	0.05180	0.04597	0.08869
38	$\Theta_1$	0.03553	0.04513	0.04810	0.05013	0.05167	0.04617	0.08910
39	$\Theta_1$	0.03360	0.04527	0.04803	0.04960	0.05173	0.04557	0.08711
40	$\Theta_1$	0.03593	0.04600	0.04790	0.04927	0.05160	0.04623	0.08828
41	$\Theta_1$	0.03453	0.04473	0.04797	0.04987	0.05160	0.04583	0.08704

_ocus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
42	$\Theta_1$	0.03587	0.04673	0.04817	0.04927	0.05180	0.04697	0.08932
43	$\Theta_1$	0.03553	0.04500	0.04803	0.04993	0.05180	0.04610	0.08777
44	$\Theta_1$	0.03540	0.04547	0.04803	0.05013	0.05160	0.04643	0.08926
45	$\Theta_1$	0.03547	0.04580	0.04777	0.04920	0.05147	0.04603	0.08844
46	$\Theta_1$	0.03180	0.04453	0.04817	0.04993	0.05173	0.04563	0.08740
47	$\Theta_1$	0.03540	0.04520	0.04790	0.04980	0.05180	0.04643	0.08903
48	$\Theta_1$	0.03600	0.04487	0.04777	0.04953	0.05167	0.04617	0.08909
49	$\Theta_1$	0.03373	0.04460	0.04817	0.05013	0.05167	0.04570	0.08678
50	$\Theta_1$	0.03613	0.04487	0.04803	0.04993	0.05167	0.04603	0.08833
51	$\Theta_1$	0.03660	0.04600	0.04797	0.04947	0.05153	0.04617	0.08847
52	$\Theta_1$	0.03700	0.04513	0.04783	0.04960	0.05173	0.04637	0.08968
53	$\Theta_1$	0.03373	0.04433	0.04817	0.04987	0.05180	0.04557	0.08867
54	$\Theta_1$	0.03500	0.04487	0.04797	0.04980	0.05167	0.04603	0.08780
55	$\Theta_1$	0.03587	0.04487	0.04810	0.04980	0.05187	0.04617	0.08898
56	$\Theta_1$	0.03673	0.04587	0.04837	0.05027	0.05160	0.04683	0.08908
57	$\Theta_1$	0.03473	0.04460	0.04790	0.04967	0.05167	0.04583	0.08756
58	$\Theta_1$	0.03500	0.04507	0.04790	0.04993	0.05160	0.04610	0.08704
59	$\Theta_1$	0.03253	0.04447	0.04790	0.04980	0.05167	0.04563	0.08750
60	$\Theta_1$	0.03660	0.04507	0.04810	0.04980	0.05153	0.04617	0.08923
61	$\Theta_1$	0.03527	0.04553	0.04830	0.05013	0.05180	0.04663	0.08925

62	$\Theta_1$	0.03440	0.04447	0.04803	0.04993	0.05160	0.04563	0.08879
63	$\Theta_1$	0.03533	0.04467	0.04790	0.04973	0.05153	0.04583	0.08735
64	$\Theta_1$	0.03580	0.04513	0.04797	0.04980	0.05180	0.04630	0.08926
65	$\Theta_1$	0.03467	0.04447	0.04763	0.04940	0.05153	0.04577	0.08876
66	$\Theta_1$	0.03740	0.04533	0.04817	0.05000	0.05167	0.04637	0.08889
67	$\Theta_1$	0.03633	0.04460	0.04783	0.04973	0.05160	0.04583	0.08866
68	$\Theta_1$	0.03600	0.04473	0.04783	0.04967	0.05147	0.04590	0.08875
69	$\Theta_1$	0.03520	0.04493	0.04810	0.05000	0.05173	0.04610	0.08863
70	$\Theta_1$	0.03747	0.04560	0.04783	0.04960	0.05167	0.04677	0.08959
71	$\Theta_1$	0.03413	0.04540	0.04803	0.05000	0.05180	0.04637	0.08903
72	$\Theta_1$	0.03560	0.04560	0.04817	0.05007	0.05193	0.04663	0.08906
73	$\Theta_1$	0.03547	0.04480	0.04797	0.04980	0.05160	0.04597	0.08749
74	$\Theta_1$	0.03473	0.04487	0.04797	0.04980	0.05173	0.04597	0.08802
75	$\Theta_1$	0.03453	0.04513	0.04797	0.04993	0.05173	0.04623	0.08967
76	$\Theta_1$	0.03387	0.04473	0.04783	0.04973	0.05167	0.04590	0.08729
77	$\Theta_1$	0.03487	0.04493	0.04817	0.05000	0.05167	0.04603	0.08813
78	$\Theta_1$	0.03533	0.04507	0.04770	0.04960	0.05153	0.04623	0.08864
79	$\Theta_1$	0.03447	0.04287	0.04790	0.05053	0.05167	0.04597	0.08859
80	$\Theta_1$	0.03353	0.04507	0.04823	0.05020	0.05180	0.04603	0.08850
81	$\Theta_1$	0.03400	0.04493	0.04790	0.04967	0.05187	0.04623	0.08925
82	$\Theta_1$	0.03547	0.04467	0.04810	0.04987	0.05173	0.04583	0.08817
83	$\Theta_1$	0.03553	0.04500	0.04823	0.04993	0.05167	0.04610	0.08896
84	$\Theta_1$	0.03473	0.04467	0.04803	0.04987	0.05147	0.04583	0.08952

Locus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
85	$\Theta_1$	0.03653	0.04533	0.04817	0.05007	0.05167	0.04637	0.08924
86	$\Theta_1$	0.03340	0.04447	0.04803	0.04993	0.05180	0.04557	0.08681
87	$\Theta_1$	0.03313	0.04493	0.04790	0.04980	0.05187	0.04610	0.08730
88	$\Theta_1$	0.03507	0.04513	0.04803	0.04973	0.05180	0.04630	0.08945
89	$\Theta_1$	0.03587	0.04513	0.04817	0.04993	0.05180	0.04630	0.08859
90	$\Theta_1$	0.03473	0.04460	0.04803	0.04987	0.05160	0.04577	0.08811
91	$\Theta_1$	0.03753	0.04573	0.04810	0.04980	0.05167	0.04690	0.08929
92	$\Theta_1$	0.03620	0.04527	0.04810	0.04973	0.05187	0.04650	0.08877
93	$\Theta_1$	0.03333	0.04507	0.04777	0.04940	0.05153	0.04523	0.08669
94	$\Theta_1$	0.03593	0.04533	0.04823	0.05020	0.05173	0.04637	0.08918
95	$\Theta_1$	0.03527	0.04473	0.04797	0.04940	0.05167	0.04610	0.08938
96	$\Theta_1$	0.03520	0.04533	0.04810	0.05000	0.05160	0.04643	0.08913
97	$\Theta_1$	0.03720	0.04507	0.04797	0.04973	0.05147	0.04617	0.08904
98	$\Theta_1$	0.03660	0.04527	0.04797	0.04987	0.05160	0.04637	0.08943
99	$\Theta_1$	0.03780	0.04487	0.04823	0.04987	0.05173	0.04610	0.08921
100	$\Theta_1$	0.03493	0.04487	0.04783	0.04973	0.05160	0.04597	0.08861
All	$\Theta_1$	0.01593	0.02067	0.02317	0.02507	0.02847	0.02270	0.09982

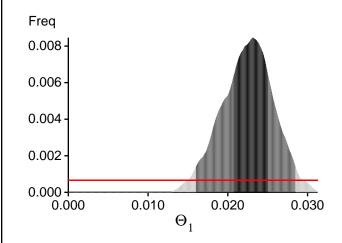
Citation suggestions:

Beerli P., 2006. Comparison of Bayesian and maximum-likelihood inference of population genetic parameters. Bioinformatics 22:341-345

Beerli P., 2007. Estimation of the population scaled mutation rate from microsatellite data, Genetics, 177:1967-1968.

Beerli P., 2009. How to use MIGRATE or why are Markov chain Monte Carlo programs difficult to use?
In Population Genetics for Animal Conservation, G. Bertorelle, M. W. Bruford, H. C. Hauffe, A. Rizzoli,
and C. Vernesi, eds., vol. 17 of Conservation Biology, Cambridge University Press, Cambridge UK, pp. 42-79.

## Bayesian Analysis: Posterior distribution over all loci



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

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

Locus	TI(1a)	BTI(1b)	SS(2)	HS(3)
1	-15911.08	-15226.20	-15209.64	-15270.83
2	-16431.13	-15633.14	-15608.87	-15662.35
3	-17691.92	-16558.13	-16466.99	-16532.07
4	-17365.79	-16125.21	-16014.15	-16072.12
5	-16223.01	-15615.09	-15611.65	-15672.54
6	-15413.05	-14925.87	-14938.77	-15003.67
7	-18865.72	-16822.61	-16568.88	-16619.00
8	-17524.89	-16641.40	-16592.75	-16663.08
9	-15173.70	-14754.75	-14775.54	-14845.92
10	-15514.75	-14999.69	-15014.44	-15077.54
11	-18330.74	-16939.26	-16815.05	-16870.93
12	-17481.98	-16652.17	-16620.32	-16678.55
13	-17190.00	-16085.59	-16007.43	-16063.11
14	-16468.40	-15596.21	-15544.94	-15609.92
15	-17695.89	-16548.50	-16456.39	-16514.94
16	-16392.63	-15678.81	-15669.76	-15723.10
17	-16478.70	-15587.91	-15543.21	-15598.52
18	-18613.13	-16868.79	-16669.98	-16727.66
19	-15410.90	-14877.00	-14871.19	-14943.34
20	-16535.60	-15801.74	-15789.23	-15843.13
21	-15712.01	-15203.30	-15220.03	-15278.71
22	-14950.54	-14617.22	-14653.80	-14720.50
23	-15912.10	-15416.13	-15438.27	-15494.63
24	-16862.05	-15678.31	-15576.92	-15629.81
25	-17651.81	-16281.54	-16152.59	-16206.30
26	-15789.36	-15353.91	-15389.30	-15444.26
27	-15470.91	-14985.68	-15006.56	-15067.45
28	-16996.38	-16229.55	-16198.11	-16266.67
29	-16549.77	-15744.13	-15704.42	-15768.94

Migrate 5.0.0a: (http://popgen.sc.fsu.edu) [program run on 21:46:26]

30	-15443.11	-14891.55	-14894.39	-14961.63
31	-18415.93	-16842.17	-16680.20	-16736.80
32	-16787.24	-15856.33	-15805.19	-15861.37
33	-16387.16	-15685.54	-15669.29	-15728.37
34	-18283.17	-17270.13	-17231.88	-17271.61
35	-15589.91	-15086.27	-15107.80	-15165.36
36	-16169.23	-15482.64	-15476.82	-15529.97
37	-16744.04	-15939.01	-15910.81	-15967.50
38	-17173.45	-16266.40	-16233.07	-16279.55
39	-16311.86	-15584.85	-15560.18	-15624.36
40	-16203.81	-15466.80	-15448.79	-15500.67
41	-15871.61	-15332.84	-15345.20	-15406.51
42	-17213.40	-16234.85	-16179.38	-16229.62
43	-16065.29	-15303.99	-15272.03	-15337.51
44	-16762.78	-15983.94	-15971.44	-16019.25
45	-17043.37	-15959.98	-15879.40	-15939.01
46	-16459.49	-15597.28	-15559.21	-15616.23
47	-18222.87	-16726.46	-16580.14	-16634.00
48	-16326.18	-15624.73	-15613.14	-15670.33
49	-15122.13	-14781.31	-14802.02	-14885.18
50	-16975.59	-16040.72	-15987.32	-16045.48
51	-17730.09	-16690.31	-16620.64	-16680.67
52	-18904.02	-17238.66	-17055.49	-17121.20
53	-17710.55	-16308.90	-16162.69	-16224.90
54	-15573.39	-15087.40	-15090.86	-15160.79
55	-16300.37	-15685.66	-15694.36	-15747.01
56	-16964.26	-15941.55	-15877.84	-15932.66
57	-15206.58	-14786.06	-14817.26	-14876.76
58	-15030.21	-14721.79	-14762.00	-14834.63
59	-15470.07	-15006.50	-15014.10	-15080.51
60	-18389.78	-16925.98	-16789.22	-16836.66
61	-17588.77	-16666.00	-16631.84	-16679.79
62	-16477.90	-15883.68	-15898.51	-15950.93
63	-15034.66	-14655.80	-14683.15	-14751.02
64	-19965.80	-17282.09	-16900.51	-16961.36
65	-18415.28	-17019.06	-16904.61	-16948.14
66	-17664.68	-16374.08	-16257.55	-16311.76
67	-17354.72	-16192.29	-16098.92	-16156.93
68	-17532.19	-16403.03	-16325.18	-16377.68
69	-17307.81	-16281.32	-16217.90	-16267.33
70	-18309.02	-16993.18	-16890.83	-16936.12
71	-17755.27	-16175.77	-15996.46	-16057.02
72	-17812.88	-16543.09	-16432.79	-16487.41
73	-15660.09	-15118.88	-15114.70	-15185.88
74	-17102.81	-16221.09	-16196.92	-16241.08

75	-18319.95	-16659.17	-16486.71	-16534.67
76	-17246.26	-16225.12	-16154.66	-16218.43
77	-16185.90	-15524.03	-15522.44	-15577.18
78	-16735.76	-15794.55	-15739.11	-15799.25
79	-16085.36	-15433.57	-15426.28	-15481.50
80	-15901.17	-15435.78	-15462.08	-15520.10
81	-18736.32	-17045.95	-16855.69	-16914.27
82	-15863.35	-15278.55	-15282.06	-15341.07
83	-20068.28	-17594.56	-17256.00	-17325.47
84	-17116.95	-16333.24	-16314.71	-16372.91
85	-16380.38	-15716.02	-15721.12	-15767.27
86	-16623.09	-15769.86	-15724.96	-15791.98
87	-17493.97	-16226.02	-16102.29	-16167.69
88	-19614.23	-17360.37	-17066.36	-17133.79
89	-16017.88	-15308.54	-15289.66	-15350.91
90	-14995.96	-14699.30	-14749.52	-14810.82
91	-16628.58	-16000.76	-16014.08	-16065.57
92	-15955.53	-15336.42	-15318.23	-15386.23
93	-15347.26	-14966.56	-14996.86	-15062.69
94	-17276.26	-16240.58	-16173.80	-16231.84
95	-16842.37	-15905.61	-15857.57	-15906.12
96	-19790.88	-17650.90	-17385.98	-17445.77
97	-16382.50	-15778.85	-15784.43	-15841.71
98	-17966.34	-16741.86	-16648.68	-16701.95
99	-18238.70	-17134.51	-17079.81	-17122.59
100	-15961.89	-15439.49	-15454.94	-15517.81
All	-1687029.74	-1593021.75	-1587453.08	-1593291.58

- (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 = 186.171595]

#### 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
$\Theta_1$ Genealogies	364037330/399995852 63957532/1600004148	0.91010 0.03997

## MCMC-Autocorrelation and Effective MCMC Sample Size

Parameter	Autocorrelation	Effective Sampe Size
$\Theta_1$ Genealogies	0.37467 0.67016	4554773.15 2014047.19

## Average temperatures during the run

#### Chain Temperatures

- 1 0.00000
- 2 0.00000
- 3 0.00000
- 4 0.00000

Adaptive heating often fails, if the average temperatures are very close together try to rerun using static heating! If you want to compare models using marginal likelihoods then you MUST use static heating

#### Potential Problems

This section reports potential problems with your run, but such reporting is often not very accurate. Whith many parameters in a multilocus analysi s, it is very common that some parameters for some loci will not be very informative, triggering suggestions (for example to increase the prior ran ge) that are not sensible. This suggestion tool will improve with time, therefore do not blindly follow its suggestions. If some parameters are fla

inference with sequence data, for mac roscopic species there is rarely the need to increase the prior for Theta beyond 0.1; but if you use microsatellites it is rather common that your prior distribution for Theta should have a range from 0.0 to 100 or more. With many populations (>3) it is also very common that some migration rou tes are estimated poorly because the data contains little or no information for that route. Increasing the range will not help in such situations, reducing number of parameters may help in such situations.		
No warning was recorded during the run		