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 15:05:09 2017

Program finished at Sat Aug 12 15:35:26 2017 [Runtime:0000:00:30:17]



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) 1704398911

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 Θ_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.0.4 Haplotyping is turned on: NO

Output file: outfile_0.4_1.0

Posterior distribution raw histogram file: bayesfile

Raw data from the MCMC run: bayesallfile_0.4_1.0

Print data: No

Print genealogies [only some for some data type]:

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]
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Bayesian Analysis: Posterior distribution table

Locus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
1	Θ_1	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
2	Θ_1	0.00000	0.00000	0.00043	0.00113	0.00253	0.00117	0.00044
3	Θ_1	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
4	Θ_1	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
5	Θ_1	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
6	Θ_1	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
7	Θ_1	0.00000	0.00000	0.00003	0.00100	0.00233	0.00103	0.00028
8	Θ_1	0.00000	0.00067	0.00163	0.00247	0.00407	0.00190	0.00169
9	Θ_1	0.00000	0.00000	0.00057	0.00120	0.00267	0.00123	0.00057
10	Θ_1	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
11	Θ_1	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
12	Θ_1	0.00600	0.01000	0.01203	0.01460	0.02553	0.01410	0.01531
13	Θ_1	0.00000	0.00000	0.00003	0.00100	0.00233	0.00103	0.00027
14	Θ_1	0.00000	0.00000	0.00003	0.00100	0.00233	0.00103	0.00027
15	Θ_1	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
16	Θ_1	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
17	Θ_1	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
18	Θ_1	0.00000	0.00000	0.00003	0.00093	0.00227	0.00097	0.00026

19	Θ_1	0.00000	0.00000	0.00057	0.00120	0.00260	0.00123	0.00054
20	Θ_1	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
21	Θ_1	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
22	Θ_1	0.00000	0.00000	0.00017	0.00107	0.00240	0.00110	0.00033
23	Θ_1	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
24	Θ_1	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
25	Θ_1	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
26	Θ_1	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
27	Θ_1	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
28	Θ_1	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
29	Θ_1	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
30	Θ_1	0.00000	0.00000	0.00003	0.00093	0.00227	0.00097	0.00025
31	Θ_1	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
32	Θ_1	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
33	Θ_1	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
34	Θ_1	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
35	Θ_1	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
36	Θ_1	0.00000	0.00000	0.00003	0.00100	0.00233	0.00103	0.00030
37	Θ_1	0.00000	0.00000	0.00003	0.00100	0.00233	0.00103	0.00030
38	Θ_1	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
39	Θ_1	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
40	Θ_1	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
41	Θ_1	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023

_ocus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
42	Θ_1	0.00000	0.00000	0.00030	0.00107	0.00240	0.00110	0.00034
43	Θ_1	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
44	Θ_1	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
45	Θ_1	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
46	Θ_1	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
47	Θ_1	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
48	Θ_1	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
49	Θ_1	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
50	Θ_1	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
51	Θ_1	0.00000	0.00000	0.00003	0.00100	0.00233	0.00103	0.00030
52	Θ_1	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
53	Θ_1	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
54	Θ_1	0.00000	0.00000	0.00010	0.00107	0.00240	0.00110	0.00033
55	Θ_1	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
56	Θ_1	0.00000	0.00000	0.00023	0.00107	0.00240	0.00110	0.00034
57	Θ_1	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
58	Θ_1	0.00647	0.01040	0.01343	0.01767	0.03053	0.01570	0.01711
59	Θ_1	0.00000	0.00127	0.00237	0.00333	0.00513	0.00250	0.00246
60	Θ_1	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
61	Θ_1	0.00000	0.00000	0.00003	0.00100	0.00233	0.00103	0.00030

62	Θ_1	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
63	Θ_1	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
64	Θ_1	0.00000	0.00000	0.00003	0.00100	0.00233	0.00103	0.00030
65	Θ_1	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
66	Θ_1	0.00000	0.00020	0.00097	0.00167	0.00313	0.00143	0.00096
67	Θ_1	0.00000	0.00000	0.00003	0.00100	0.00233	0.00103	0.00028
68	Θ_1	0.00000	0.00000	0.00003	0.00100	0.00233	0.00103	0.00027
69	Θ_1	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
70	Θ_1	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
71	Θ_1	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
72	Θ_1	0.00000	0.00000	0.00003	0.00100	0.00233	0.00103	0.00027
73	Θ_1	0.00000	0.00000	0.00017	0.00107	0.00240	0.00110	0.00034
74	Θ_1	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
75	Θ_1	0.00000	0.00000	0.00003	0.00100	0.00233	0.00103	0.00029
76	Θ_1	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
77	Θ_1	0.00000	0.00000	0.00023	0.00107	0.00240	0.00110	0.00032
78	Θ_1	0.00000	0.00000	0.00003	0.00093	0.00227	0.00097	0.00025
79	Θ_1	0.00000	0.00000	0.00030	0.00107	0.00247	0.00110	0.00039
80	Θ_1	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
81	Θ_1	0.00000	0.00000	0.00030	0.00107	0.00247	0.00110	0.00038
82	Θ_1	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
83	Θ_1	0.00000	0.00000	0.00003	0.00100	0.00233	0.00103	0.00027
84	Θ_1	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023

Locus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
85	Θ_1	0.00000	0.00033	0.00110	0.00187	0.00333	0.00157	0.00114
86	Θ_1	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
87	Θ_1	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
88	Θ_1	0.00000	0.00000	0.00043	0.00113	0.00260	0.00117	0.00048
89	Θ_1	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
90	Θ_1	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
91	Θ_1	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00022
92	Θ_1	0.00000	0.00000	0.00030	0.00107	0.00247	0.00110	0.00037
93	Θ_1	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
94	Θ_1	0.00040	0.00247	0.00383	0.00513	0.00847	0.00417	0.00434
95	Θ_1	0.00000	0.00000	0.00057	0.00120	0.00267	0.00123	0.00058
96	Θ_1	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00022
97	Θ_1	0.00000	0.00000	0.00003	0.00093	0.00227	0.00097	0.00025
98	Θ_1	0.00000	0.00000	0.00070	0.00133	0.00280	0.00130	0.00069
99	Θ_1	0.00207	0.00460	0.00630	0.00827	0.01413	0.00710	0.00761
100	Θ_1	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
All	Θ_1	0.00000	0.00000	0.00017	0.00040	0.00040	0.00103	0.00017

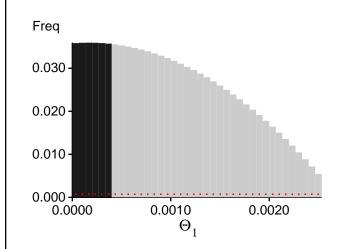
Citation suggestions:

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	-14119.10	-13771.47	-13774.76	-13868.11
2	-14262.80	-13891.79	-13907.08	-13985.96
3	-14125.47	-13773.04	-13776.07	-13868.36
4	-14124.59	-13773.92	-13777.20	-13877.10
5	-14117.91	-13771.79	-13775.54	-13868.32
6	-14122.50	-13773.40	-13776.33	-13869.62
7	-14132.00	-13784.38	-13790.56	-13880.80
8	-14358.25	-14007.15	-14039.71	-14106.22
9	-14199.03	-13849.40	-13867.51	-13948.79
10	-14123.37	-13773.12	-13776.23	-13869.03
11	-14123.51	-13774.32	-13777.33	-13870.04
12	-22912.40	-19086.71	-18527.54	-18574.72
13	-14148.64	-13794.71	-13800.70	-13890.54
14	-14149.60	-13796.24	-13802.21	-13891.97
15	-14121.15	-13771.26	-13774.56	-13867.33
16	-14119.96	-13771.92	-13775.58	-13868.86
17	-14121.97	-13772.20	-13775.01	-13868.15
18	-14133.71	-13786.59	-13791.82	-13883.02
19	-14333.36	-13947.64	-13963.25	-14038.82
20	-14124.92	-13774.13	-13776.57	-13869.77
21	-14123.66	-13774.42	-13777.48	-13870.10
22	-14157.85	-13808.27	-13815.13	-13903.91
23	-14122.61	-13773.79	-13776.88	-13869.80
24	-14120.93	-13772.70	-13776.49	-13868.98
25	-14124.86	-13774.41	-13777.39	-13871.67
26	-14119.39	-13770.74	-13774.33	-13867.47
27	-14123.96	-13774.41	-13777.90	-13870.75
28	-14120.88	-13770.98	-13774.39	-13868.76
29	-14121.00	-13773.96	-13777.67	-13870.43

Migrate 5.0.0a: (http://popgen.sc.fsu.edu) [program run on 15:05:09]

30	-14135.87	-13787.09	-13790.73	-13882.73
31	-14122.77	-13774.03	-13777.31	-13871.88
32	-14122.49	-13773.87	-13777.54	-13870.63
33	-14121.30	-13772.24	-13775.85	-13869.02
34	-14122.37	-13773.50	-13775.44	-13870.91
35	-14124.82	-13773.35	-13776.45	-13870.20
36	-14146.18	-13798.38	-13804.96	-13894.17
37	-14149.30	-13799.56	-13805.05	-13894.44
38	-14123.31	-13773.14	-13776.19	-13871.50
39	-14123.81	-13773.80	-13777.30	-13870.21
40	-14124.90	-13774.11	-13776.83	-13871.04
41	-14123.55	-13773.87	-13777.45	-13869.95
42	-14186.07	-13837.63	-13850.53	-13934.17
43	-14123.46	-13773.14	-13776.57	-13869.95
44	-14122.41	-13773.24	-13777.03	-13869.71
45	-14123.00	-13773.33	-13776.76	-13869.60
46	-14120.78	-13773.38	-13776.86	-13870.25
47	-14122.45	-13773.58	-13777.30	-13869.69
48	-14124.34	-13773.55	-13776.82	-13869.55
49	-14123.40	-13773.79	-13776.48	-13869.83
50	-14121.98	-13773.78	-13777.54	-13870.30
51	-14147.56	-13799.03	-13805.02	-13896.31
52	-14118.77	-13770.22	-13773.81	-13868.45
53	-14124.77	-13774.36	-13777.30	-13871.33
54	-14142.52	-13794.82	-13803.34	-13893.60
55	-14122.64	-13774.16	-13778.28	-13870.30
56	-14170.95	-13816.06	-13826.48	-13911.56
57	-14121.07	-13772.85	-13776.39	-13869.10
58	-19927.24	-18920.94	-18904.50	-18950.75
59	-15707.04	-14890.68	-14846.60	-14912.73
60	-14119.83	-13770.82	-13774.29	-13868.12
61	-14139.59	-13791.13	-13797.60	-13887.85
62	-14124.97	-13774.26	-13777.30	-13870.53
63	-14123.40	-13774.09	-13777.68	-13870.91
64	-14145.02	-13796.64	-13803.06	-13893.11
65	-14124.32	-13773.63	-13776.68	-13869.23
66	-14263.09	-13920.49	-13944.84	-14018.30
67	-14153.94	-13802.44	-13810.10	-13898.90
68	-14147.40	-13795.11	-13801.51	-13891.15
69	-14117.25	-13768.77	-13772.01	-13864.95
70	-14123.46	-13773.76	-13777.46	-13871.12
71	-14121.47	-13772.28	-13775.63	-13868.31
72	-14136.74	-13786.07	-13791.85	-13882.03
73	-14161.91	-13809.81	-13816.38	-13905.09
74	-14119.25	-13771.87	-13774.79	-13869.54
L				

All	-1435764.19	-1394350.98	-1394098.42	-1402953.61
100	-14124.00	-13//4.40	-13777.08	-130/0./0
99 100	-18910.26 -14124.88	-16865.25 -13774.40	-16617.35 -13777.59	-16670.51 -13870.78
98	-14314.33	-13951.38	-13973.14	-14046.99
97	-14135.97	-13786.72	-13789.83	-13882.22
				-13869.24
95 96	-14335.10 -14120.62	-13964.62 -13771.99	-13964.29 -13775.81	
94 95	-14799.03 -14335.10	-14422.67 -13964.62	-13984.29	-14520.12 -14059.08
93 94	-14121.57 -14799.03	-13772.93 -14422.87	-13776.52 -14463.17	-13669.19 -14520.12
92	-14175.49 -14121.57	-13623.50 -13772.93	-13634.79 -13776.52	-13869.19
91	-14121.97 -14175.49	-13772.57 -13823.50	-13834.79	-13919.43
90	-14121.84 -14121.97	-13773.74 -13772.57	-13777.12 -13776.28	-13868.80
90	-14121.84	-13773.65 -13773.74	-13777.57 -13777.12	-13871.61
89	-14179.27 -14121.68	-13033.11 -13773.85	-13647.33 -13777.57	-13930.42
88	-14123.15 -14179.27	-13774.23 -13833.11	-13777.70 -13847.33	-13930.42
87	-14123.15	-13774.12	-13777.70	-13872.46
86	-14466.45 -14124.60	-14079.39 -13774.12	-13777.81	-13870.98
85	-14120.74 -14466.45	-14079.39	-13774.35 -14103.47	-13007.30 -14172.34
84	-14120.74	-13770.70	-13774.35	-13867.30
83	-14123.40 -14147.39	-13773.01	-13800.59	-13890.73
82	-14171.92	-13773.01	-13776.42	-13869.02
81	-14176.04 -14171.92	-13821.40	-13832.55	-13916.37
79 80	-14154.64 -14118.64	-13607.45 -13771.11	-13774.83	-13904.26 -13867.37
78 79	-14154.29 -14154.84	-13764.63 -13807.45	-13818.40	-13904.28
77 78	-14194.70 -14134.29	-13784.63	-13788.05	-13879.69
70 77	-14123.92	-13833.32	-13844.03	-13927.47
75 76	-14166.32 -14125.92	-13812.51 -13774.42	-13820.98 -13777.66	-13908.05 -13870.82

- (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!
- (3) HS: Harmonic mean approximation: Overestimates the marginal likelihood, poor variance [Scaling factor = 512.353751]

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.

(2) SS: Steppingstone Sampling (Xie et al 2011)

Acceptance ratios for all parameters and the genealogies

Parameter	Accepted changes	Ratio
Θ_1 Genealogies	37090548/400034760 1013791446/1599965240	0.09272 0.63363

MCMC-Autocorrelation and Effective MCMC Sample Size

Parameter	Autocorrelation	Effective Sampe Size
Θ_1	0.08280	8482092.95
Genealogies	0.10768	8124237.91

Average temperatures during the run

Chain Temperatures 1 0.00000

3 0.000004 0.00000

0.00000

2

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 aged inspect the tables carefully and judge wether an action is required. For example, if you run a Rayesian

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