AUTO

POPULATION SIZE, MIGRATION, DIVERGENCE, ASSIGNMENT, HISTORY

Bayesian inference using the structured coalescent

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

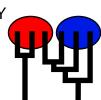
Using Intel AVX (Advanced Vector Extensions)

Compiled for PARALLEL computer architectures

One master and 39 compute nodes are available.

Program started at Mon Aug 14 13:04:13 2017

Program finished at Mon Aug 14 18:31:11 2017 [Runtime:0000:05:26:58]



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) 1181060031

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 chains

Recorded steps [a]

Increment (record every x step [b]

Number of concurrent chains (replicates) [c]

1
50000

200

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.0.8

Haplotyping is turned on:

Output file: outfile_0.8_0.6

Posterior distribution raw histogram file: bayesfile
Raw data from the MCMC run: bayesallfile_0.8_0.6

Print data:

Print genealogies [only some for some data type]:

Data summary

Data file: infile.0.8
Datatype: Sequence data
Number of loci: 100

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Mutation	model:			
Locus S		Mutationmodel	Mutationmodel parameters	
1	1	Jukes-Cantor	[Basefreq: =0.25]	
2	1	Jukes-Cantor	[Basefreq: =0.25]	
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Jukes-Cantor

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3	1 1	1.000	1.000	1.000	
4	1 1	1.000	1.000	1.000	
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Bayesian Analysis: Posterior distribution table

_ocus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
1	Θ_1	0.02853	0.04233	0.04757	0.04947	0.05140	0.04317	0.07133
2	Θ_1	0.02867	0.04300	0.04770	0.04927	0.05140	0.04317	0.07235
3	Θ_1	0.02780	0.04253	0.04763	0.04920	0.05140	0.04270	0.06998
4	Θ_1	0.02053	0.03820	0.04757	0.04873	0.05100	0.03857	0.06032
5	Θ_1	0.02007	0.03153	0.04257	0.04827	0.05053	0.03677	0.05037
6	Θ_1	0.02287	0.03940	0.04757	0.04880	0.05100	0.03957	0.05997
7	Θ_1	0.01887	0.03507	0.04237	0.04767	0.05047	0.03623	0.05197
8	Θ_1	0.01713	0.02707	0.03290	0.04200	0.05000	0.03390	0.04502
9	Θ_1	0.02293	0.03987	0.04757	0.04893	0.05107	0.04010	0.06397
10	Θ_1	0.02233	0.03860	0.04750	0.04873	0.05087	0.03903	0.05818
11	Θ_1	0.02080	0.03760	0.04717	0.04860	0.05087	0.03843	0.05737
12	Θ_1	0.02473	0.03953	0.04750	0.04920	0.05120	0.04103	0.06352
13	Θ_1	0.03107	0.04333	0.04770	0.04967	0.05147	0.04450	0.07921
14	Θ_1	0.02620	0.03920	0.04757	0.04967	0.05127	0.04197	0.06802
15	Θ_1	0.02100	0.03733	0.04550	0.04873	0.05093	0.03863	0.05847
16	Θ_1	0.02613	0.04080	0.04763	0.04953	0.05127	0.04210	0.06979
17	Θ_1	0.02180	0.03653	0.04530	0.04833	0.05080	0.03823	0.05498
18	Θ_1	0.02280	0.03967	0.04750	0.04893	0.05107	0.03983	0.06154

19	Θ_1	0.01767	0.03047	0.03790	0.04500	0.05040	0.03517	0.04778
20	Θ_1	0.01753	0.02827	0.03490	0.04380	0.05020	0.03457	0.04607
21	Θ_1	0.03020	0.04387	0.04777	0.04947	0.05147	0.04403	0.07671
22	Θ_1	0.02667	0.04187	0.04757	0.04913	0.05127	0.04203	0.06892
23	Θ_1	0.02540	0.04047	0.04763	0.04933	0.05127	0.04150	0.06729
24	Θ_1	0.02307	0.03860	0.04757	0.04913	0.05113	0.04010	0.06340
25	Θ_1	0.01733	0.02820	0.03570	0.04273	0.05020	0.03443	0.04569
26	Θ_1	0.02780	0.04273	0.04777	0.04947	0.05140	0.04297	0.07242
27	Θ_1	0.02687	0.04187	0.04763	0.04913	0.05127	0.04210	0.06871
28	Θ_1	0.02487	0.04013	0.04763	0.04940	0.05133	0.04150	0.06840
29	Θ_1	0.01853	0.03207	0.03983	0.04567	0.05033	0.03563	0.04870
30	Θ_1	0.01880	0.03180	0.03763	0.04720	0.05047	0.03590	0.04928
31	Θ_1	0.02287	0.03980	0.04757	0.04887	0.05120	0.04010	0.06244
32	Θ_1	0.02387	0.02760	0.04757	0.05053	0.05107	0.04010	0.06118
33	Θ_1	0.02927	0.04267	0.04777	0.04973	0.05147	0.04390	0.07749
34	Θ_1	0.02513	0.04007	0.04757	0.04927	0.05127	0.04150	0.06624
35	Θ_1	0.03180	0.04367	0.04777	0.04973	0.05153	0.04483	0.08274
36	Θ_1	0.02207	0.03833	0.04750	0.04847	0.05087	0.03857	0.05506
37	Θ_1	0.01420	0.02160	0.02850	0.03713	0.04940	0.03090	0.03927
38	Θ_1	0.02060	0.03800	0.04750	0.04860	0.05087	0.03817	0.05731
39	Θ_1	0.02320	0.03820	0.04750	0.04880	0.05100	0.03957	0.05961
40	Θ_1	0.02313	0.03867	0.04750	0.04887	0.05093	0.03970	0.06074
41	Θ_1	0.02127	0.03727	0.04757	0.04900	0.05107	0.03883	0.06102

Locus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
42	Θ_1	0.02940	0.04273	0.04777	0.04960	0.05153	0.04397	0.08207
43	Θ_1	0.01607	0.02447	0.03210	0.03873	0.04953	0.03230	0.04122
44	Θ_1	0.02400	0.04020	0.04757	0.04900	0.05107	0.04037	0.06196
45	Θ_1	0.02493	0.04067	0.04750	0.04887	0.05120	0.04103	0.06408
46	Θ_1	0.02713	0.04107	0.04770	0.04947	0.05133	0.04237	0.06991
47	Θ_1	0.02460	0.04120	0.04757	0.04907	0.05133	0.04143	0.06826
48	Θ_1	0.02540	0.04013	0.04757	0.04927	0.05127	0.04157	0.06713
49	Θ_1	0.02233	0.03600	0.04750	0.04940	0.05113	0.03977	0.06219
50	Θ_1	0.02053	0.03780	0.04757	0.04913	0.05107	0.03897	0.06297
51	Θ_1	0.01667	0.02680	0.03490	0.04187	0.05007	0.03363	0.04461
52	Θ_1	0.02440	0.04040	0.04763	0.04907	0.05120	0.04077	0.06396
53	Θ_1	0.02113	0.03740	0.04750	0.04873	0.05093	0.03850	0.05645
54	Θ_1	0.02487	0.04013	0.04757	0.04933	0.05127	0.04150	0.06738
55	Θ_1	0.01140	0.01833	0.02503	0.03260	0.04867	0.02823	0.03639
56	Θ_1	0.02380	0.04020	0.04757	0.04907	0.05120	0.04063	0.06324
57	Θ_1	0.02927	0.04220	0.04763	0.04960	0.05140	0.04343	0.07347
58	Θ_1	0.02633	0.04207	0.04770	0.04940	0.05133	0.04223	0.07136
59	Θ_1	0.02353	0.03900	0.04763	0.04913	0.05120	0.04057	0.06614
60	Θ_1	0.02673	0.04213	0.04770	0.04933	0.05140	0.04243	0.07177
61	Θ_1	0.02480	0.03960	0.04757	0.04927	0.05120	0.04103	0.06550

62	Θ_1	0.02387	0.03893	0.04757	0.04907	0.05107	0.04023	0.06238
63	Θ_1	0.01960	0.03247	0.03670	0.04580	0.05033	0.03603	0.04839
64	Θ_1	0.02200	0.03880	0.04730	0.04860	0.05100	0.03903	0.05742
65	Θ_1	0.01840	0.03313	0.03923	0.04733	0.05053	0.03610	0.05058
66	Θ_1	0.02380	0.03947	0.04617	0.04880	0.05100	0.04003	0.06064
67	Θ_1	0.01953	0.03607	0.04717	0.04853	0.05080	0.03750	0.05704
68	Θ_1	0.01907	0.03580	0.04537	0.04787	0.05060	0.03663	0.05200
69	Θ_1	0.02460	0.04033	0.04763	0.04947	0.05133	0.04163	0.06928
70	Θ_1	0.02547	0.04007	0.04757	0.04913	0.05120	0.04117	0.06487
71	Θ_1	0.02947	0.04247	0.04770	0.04967	0.05153	0.04370	0.07753
72	Θ_1	0.02447	0.04067	0.04757	0.04907	0.05120	0.04083	0.06618
73	Θ_1	0.01920	0.03340	0.03843	0.04373	0.05047	0.03610	0.04934
74	Θ_1	0.02513	0.04013	0.04757	0.04927	0.05127	0.04157	0.06675
75	Θ_1	0.02087	0.03867	0.04750	0.04880	0.05100	0.03890	0.06003
76	Θ_1	0.02180	0.03820	0.04663	0.04860	0.05093	0.03883	0.05884
77	Θ_1	0.01933	0.03013	0.03690	0.04767	0.05033	0.03590	0.04890
78	Θ_1	0.01940	0.02960	0.04177	0.04873	0.05060	0.03657	0.05182
79	Θ_1	0.02940	0.04267	0.04777	0.04960	0.05153	0.04390	0.07608
80	Θ_1	0.01673	0.02533	0.03283	0.03927	0.04953	0.03277	0.04185
81	Θ_1	0.02253	0.03907	0.04757	0.04880	0.05100	0.03950	0.06058
82	Θ_1	0.02133	0.03733	0.04750	0.04887	0.05100	0.03903	0.06047
83	Θ_1	0.02247	0.03747	0.04637	0.04873	0.05087	0.03910	0.05790
84	Θ_1	0.02967	0.04273	0.04777	0.04967	0.05153	0.04397	0.07658

85 86 87	Parameter Θ_1 Θ_1 Θ_1 Θ_1	2.5% 0.02220 0.02293	25.0% 0.03847 0.03807	Mode 0.04750	75.0% 0.04867	97.5%	Median 0.03883	Mean
86	Θ_1	0.02293			0.04867	0.05087	0 03003	
87			0.03807				0.03003	0.05739
	Θ_1	0.0040=		0.04757	0.04893	0.05100	0.03950	0.05950
88		0.02187	0.03827	0.04763	0.04920	0.05120	0.03977	0.06451
	Θ_1	0.02847	0.04167	0.04770	0.04947	0.05140	0.04303	0.07256
89	Θ_1	0.01980	0.03753	0.04517	0.04840	0.05087	0.03783	0.05547
90	Θ_1	0.02400	0.03873	0.04763	0.04907	0.05113	0.04017	0.06008
91	Θ_1	0.02027	0.03607	0.03810	0.04747	0.05060	0.03710	0.05207
92	Θ_1	0.02840	0.04213	0.04770	0.04967	0.05153	0.04337	0.07536
93	Θ_1	0.03013	0.04300	0.04783	0.04967	0.05153	0.04423	0.07872
94	Θ_1	0.02127	0.03687	0.04570	0.04840	0.05080	0.03810	0.05515
95	Θ_1	0.01687	0.02867	0.03470	0.04253	0.05013	0.03417	0.04599
96	Θ_1	0.02880	0.04333	0.04777	0.04947	0.05147	0.04350	0.07725
97	Θ_1	0.02733	0.04153	0.04770	0.04953	0.05147	0.04283	0.07311
98	Θ_1	0.02700	0.04247	0.04763	0.04927	0.05147	0.04263	0.07364
99	Θ_1	0.02653	0.04100	0.04763	0.04947	0.05133	0.04230	0.07030
100	Θ_1	0.02333	0.03907	0.04750	0.04893	0.05107	0.04017	0.06177
All	Θ_1	0.04693	0.04833	0.04937	0.05033	0.05193	0.04943	0.05927

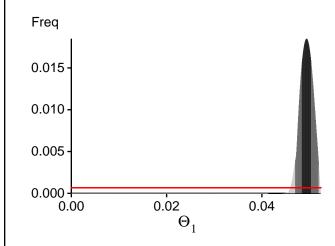
Citation suggestions:

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	-15820.59	-15023.50	-14990.09	-15048.66
2	-15315.80	-14854.26	-14884.46	-14942.47
3	-14823.41	-14499.45	-14551.66	-14609.31
4	-14445.37	-14245.08	-14305.57	-14374.97
5	-14328.54	-14097.71	-14152.95	-14221.04
6	-15986.45	-15499.33	-15525.68	-15591.49
7	-14518.53	-14243.99	-14292.92	-14360.74
8	-14416.18	-14112.46	-14150.06	-14223.60
9	-23982.07	-20984.56	-20594.88	-20657.16
10	-14534.27	-14216.13	-14257.14	-14325.13
11	-14182.65	-13996.37	-14059.09	-14125.95
12	-15137.69	-14816.75	-14869.50	-14930.92
13	-16945.12	-15832.47	-15753.53	-15805.65
14	-15899.07	-15262.08	-15260.94	-15321.72
15	-14364.20	-14142.68	-14201.36	-14268.37
16	-14900.51	-14599.29	-14655.90	-14714.89
17	-14395.62	-14118.87	-14167.68	-14234.19
18	-14262.57	-14075.07	-14141.25	-14205.63
19	-14146.50	-13951.69	-14008.73	-14080.31
20	-14177.90	-13977.55	-14035.23	-14106.60
21	-15509.93	-15078.08	-15118.65	-15174.98
22	-18995.64	-17208.25	-17009.57	-17069.33
23	-15548.39	-15019.56	-15034.27	-15095.73
24	-15417.17	-14886.90	-14897.86	-14963.99
25	-14263.24	-14016.14	-14064.12	-14135.16
26	-15931.76	-15333.29	-15344.56	-15400.82
27	-15472.15	-14869.37	-14869.42	-14929.90
28	-14696.96	-14401.36	-14455.24	-14515.13
29	-14211.72	-14018.32	-14077.84	-14147.84

Migrate 5.0.0a: (http://popgen.sc.fsu.edu) [program run on 13:04:13]

30	-14214.98	-14008.84	-14067.50	-14136.14
31	-17105.80	-15811.05	-15685.84	-15749.48
32	-15723.65	-14908.44	-14864.33	-14929.88
33	-15333.33	-14866.09	-14898.35	-14952.42
34	-14789.07	-14428.20	-14469.82	-14531.66
35	-20023.19	-18818.31	-18764.79	-18809.07
36	-14353.04	-14119.86	-14177.52	-14242.58
37	-14031.67	-13865.28	-13922.41	-13997.62
38	-14224.67	-14032.09	-14094.19	-14162.55
39	-15016.40	-14521.87	-14535.72	-14599.76
40	-14406.84	-14176.78	-13938.41	-14300.02
41	-16037.24	-15511.81	-14099.91	-15595.45
42	-44177.23	-35291.63	-14108.05	-33996.40
43	-14152.31	-13962.97	-14019.49	-14095.15
44	-15011.61	-14516.14	-14161.72	-14593.16
45	-14506.62	-14227.97	-14164.90	-14344.02
46	-14866.54	-14516.51	-14050.41	-14621.88
47	-15440.57	-14981.31	-14267.30	-15070.62
48	-14394.63	-14157.88	-14074.44	-14282.28
49	-14417.12	-14194.45	-14021.21	-14323.46
50	-26492.57	-23259.20	-14150.96	-22918.28
51	-14059.54	-13888.72	-13948.39	-14022.77
52	-14726.82	-14394.18	-14182.30	-14500.58
53	-14649.35	-14318.15	-14308.52	-14424.82
54	-14572.49	-14321.60	-14208.92	-14444.51
55	-14591.89	-14314.57	-14073.07	-14433.52
56	-14308.32	-14093.75	-14157.86	-14221.33
57	-14950.10	-14541.10	-14577.78	-14635.47
58	-15694.56	-15292.84	-14307.41	-15394.76
59	-15227.84	-14974.95	-14541.70	-15099.04
60	-16130.44	-15320.32	-14993.57	-15348.31
61	-15080.29	-14768.36	-14822.79	-14883.95
62	-14279.63	-14080.74	-14147.32	-14209.70
63	-14283.60	-14043.72	-14095.56	-14165.09
64	-14321.47	-14110.59	-14173.10	-14238.13
65	-14133.46	-13950.35	-14010.57	-14080.36
66	-14493.75	-14193.47	-14240.87	-14304.51
67	-14311.66	-14117.49	-14177.19	-14246.78
68	-14308.01	-14068.94	-14121.34	-14190.43
69	-70619.32	-37782.28	-14884.80	-30597.65
70	-14347.90	-14132.90	-14197.67	-14258.85
71	-21562.64	-18534.03	-14561.08	-18171.33
72	-15199.29	-14883.56	-14936.13	-14999.62
73	-14255.71	-14022.54	-14076.01	-14144.99
74	-14505.25	-14238.26	-14295.69	-14356.23

75	-14447.92	-14205.75	-14263.62	-14329.35
76	-14260.90	-14052.03	-14112.83	-14179.12
77	-14297.13	-14057.21	-14109.57	-14178.35
78	-14184.12	-13983.47	-14043.93	-14111.32
79	-15698.23	-15035.40	-14249.68	-15086.40
80	-14315.26	-14042.02	-13948.37	-14156.55
81	-14645.00	-14329.66	-14375.70	-14441.56
82	-21470.12	-20181.66	-14037.61	-20172.46
83	-14449.64	-14159.44	-14206.82	-14272.03
84	-14980.02	-14609.28	-14233.64	-14711.04
85	-14219.05	-14018.91	-14014.45	-14147.16
86	-14416.21	-14150.16	-14188.26	-14267.56
87	-14404.88	-14197.37	-14105.12	-14326.38
88	-14834.51	-14569.32	-14566.89	-14690.96
89	-14421.96	-14176.20	-14228.74	-14297.17
90	-14814.80	-14411.86	-14181.35	-14505.29
91	-14213.02	-14026.26	-14088.90	-14156.89
92	-26078.02	-21044.21	-14167.82	-20324.95
93	-23421.54	-19948.60	-14264.93	-19517.53
94	-14527.98	-14278.27	-14144.34	-14400.45
95	-15548.25	-14957.21	-14433.03	-15021.52
96	-14997.21	-14711.54	-14398.06	-14830.72
97	-15313.94	-14872.23	-14375.73	-14963.46
98	-14652.12	-14437.66	-14509.64	-14566.44
99	-17799.27	-16961.29	-15019.48	-17001.69
100	-14655.40	-14345.52	-14393.58	-14457.18
All	-1631002.67	-1535680.92	-1450942.92	-1534293.60

- (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 = 26.224801]

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
Θ_1 Genealogies	383941169/400042638 156208834/1599957362	0.95975 0.09763

MCMC-Autocorrelation and Effective MCMC Sample Size

Parameter	Autocorrelation	Effective Sampe Size
Θ_1	0.66450	5355411.94
Genealogies	0.17821	19047216.92

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

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