AUTO

POPULATION SIZE, MIGRATION, DIVERGENCE, ASSIGNMENT, HISTORY

Bayesian inference using the structured coalescent

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

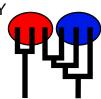
Using Intel AVX (Advanced Vector Extensions)

Compiled for PARALLEL computer architectures

One master and 39 compute nodes are available.

Program started at Fri Aug 11 08:03:19 2017

Program finished at Fri Aug 11 23:14:45 2017 [Runtime:0000:15:11:26]



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) 153332200

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

Exponential Distribution -Population size estimation:

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 Delta Prior Minimum Mean Maximum Bins UpdateFreq 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 50000 Recorded steps [a] 200 Increment (record every x step [b] Number of concurrent chains (replicates) [c]

20000000 Visited (sampled) parameter values [a*b*c] 10000 Number of discard trees per chain (burn-in)

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.0.6 NO

Haplotyping is turned on:

Output file: outfile_0.6_0.7

Posterior distribution raw histogram file: bayesfile Raw data from the MCMC run: bayesallfile_0.6_0.7

Print data: No

Print genealogies [only some for some data type]: None

Data summary

Data file: infile.0.6
Datatype: Sequence data
Number of loci: 100

Number	OI IOCI.			100
Mutation	model:			
Locus S		Mutationmodel	Mutationmodel parameters	
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Jukes-Cantor

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	97	10
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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	0.00133	0.00433	0.00663	0.00973	0.01940	0.00830	0.00926
2	Θ_1	0.00320	0.00993	0.01257	0.01573	0.03980	0.01577	0.01791
3	Θ_1	0.00140	0.00447	0.00683	0.00993	0.01987	0.00850	0.00950
4	Θ_1	0.02593	0.04220	0.04770	0.04960	0.05153	0.04257	0.07803
5	Θ_1	0.00307	0.00860	0.01137	0.01493	0.03520	0.01483	0.01699
6	Θ_1	0.00467	0.00740	0.01217	0.01913	0.02747	0.01497	0.01686
7	Θ_1	0.00080	0.00367	0.00583	0.00853	0.01733	0.00723	0.00813
8	Θ_1	0.00160	0.00267	0.00790	0.01887	0.02393	0.01023	0.01156
9	Θ_1	0.00440	0.00767	0.01223	0.01887	0.03000	0.01543	0.01754
10	Θ_1	0.00100	0.00393	0.00603	0.00893	0.01773	0.00757	0.00843
11	Θ_1	0.00180	0.00300	0.00863	0.02120	0.02767	0.01117	0.01259
12	Θ_1	0.00407	0.00960	0.01257	0.01667	0.03340	0.01610	0.01841
13	Θ_1	0.00460	0.00780	0.00997	0.01247	0.01953	0.01243	0.01396
14	Θ_1	0.00060	0.00320	0.00510	0.00753	0.01493	0.00630	0.00701
15	Θ_1	0.00433	0.00667	0.00983	0.01433	0.02027	0.01237	0.01381
16	Θ_1	0.00007	0.00247	0.00417	0.00620	0.01233	0.00510	0.00569
17	Θ_1	0.00080	0.00380	0.00517	0.00673	0.01413	0.00643	0.00718
18	Θ_1	0.00000	0.00207	0.00370	0.00547	0.01087	0.00450	0.00496

19	Θ_1	0.00213	0.00307	0.00570	0.00967	0.01187	0.00723	0.00810
20	Θ_1	0.00673	0.01360	0.01630	0.02000	0.03873	0.02037	0.02327
21	Θ_1	0.00253	0.00587	0.00717	0.00867	0.01600	0.00897	0.01003
22	Θ_1	0.00553	0.01067	0.01323	0.01573	0.02787	0.01657	0.01893
23	Θ_1	0.00133	0.00467	0.00770	0.01167	0.02527	0.01010	0.01158
24	Θ_1	0.00213	0.00387	0.00530	0.00700	0.01033	0.00663	0.00738
25	Θ_1	0.00173	0.00500	0.00757	0.01113	0.02133	0.00950	0.01064
26	Θ_1	0.00840	0.01160	0.01897	0.03087	0.04160	0.02283	0.02612
27	Θ_1	0.00233	0.00493	0.00823	0.01313	0.02287	0.01137	0.01365
28	Θ_1	0.00193	0.00420	0.00650	0.00947	0.01547	0.00803	0.00898
29	Θ_1	0.00087	0.00540	0.00830	0.01220	0.03253	0.01050	0.01190
30	Θ_1	0.00253	0.00493	0.00723	0.01053	0.01680	0.00937	0.01056
31	Θ_1	0.00140	0.00460	0.00710	0.01067	0.02207	0.00910	0.01036
32	Θ_1	0.00427	0.01027	0.01317	0.01660	0.03567	0.01650	0.01862
33	Θ_1	0.00020	0.00253	0.00430	0.00633	0.01253	0.00523	0.00583
34	Θ_1	0.00333	0.00973	0.01110	0.01273	0.03220	0.01430	0.01609
35	Θ_1	0.00247	0.00687	0.00743	0.00793	0.01707	0.00937	0.01057
36	Θ_1	0.00007	0.00247	0.00417	0.00627	0.01247	0.00517	0.00575
37	Θ_1	0.00500	0.00920	0.01330	0.01993	0.03280	0.01683	0.01897
38	Θ_1	0.00160	0.00580	0.00903	0.01313	0.03260	0.01190	0.01406
39	Θ_1	0.00887	0.01493	0.02130	0.02973	0.04700	0.02550	0.03205
40	Θ_1	0.00500	0.00887	0.01317	0.01940	0.03233	0.01637	0.01843
41	Θ_1	0.01000	0.01520	0.01930	0.02513	0.03720	0.02337	0.02697

_ocus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
42	Θ_1	0.00040	0.00287	0.00477	0.00707	0.01427	0.00597	0.00666
43	Θ_1	0.00233	0.00513	0.00883	0.01427	0.02480	0.01090	0.01224
44	Θ_1	0.00527	0.01113	0.01290	0.01440	0.02773	0.01590	0.01791
45	Θ_1	0.00167	0.00760	0.00983	0.01240	0.03447	0.01237	0.01395
46	Θ_1	0.00180	0.00387	0.00597	0.00873	0.01387	0.00737	0.00824
47	Θ_1	0.00000	0.00207	0.00370	0.00547	0.01067	0.00450	0.00492
48	Θ_1	0.00140	0.00440	0.00670	0.00973	0.01947	0.00830	0.00933
49	Θ_1	0.00113	0.00400	0.00623	0.00920	0.01860	0.00783	0.00879
50	Θ_1	0.00240	0.00540	0.00663	0.00800	0.01427	0.00823	0.00917
51	Θ_1	0.00000	0.00207	0.00370	0.00547	0.01073	0.00450	0.00494
52	Θ_1	0.00233	0.00720	0.00957	0.01267	0.03100	0.01303	0.01515
53	Θ_1	0.00033	0.00393	0.00610	0.00887	0.02253	0.00757	0.00840
54	Θ_1	0.00007	0.00247	0.00417	0.00620	0.01240	0.00517	0.00571
55	Θ_1	0.00407	0.00847	0.00990	0.01140	0.02133	0.01290	0.01489
56	Θ_1	0.01320	0.02020	0.02690	0.03720	0.04913	0.03017	0.04048
57	Θ_1	0.00100	0.00633	0.00790	0.00973	0.02933	0.01050	0.01201
58	Θ_1	0.00053	0.00320	0.00503	0.00740	0.01447	0.00617	0.00685
59	Θ_1	0.00487	0.00847	0.01063	0.01287	0.02067	0.01297	0.01453
60	Θ_1	0.00460	0.01293	0.01477	0.01713	0.04140	0.01890	0.02185
61	Θ_1	0.00040	0.00293	0.00470	0.00693	0.01360	0.00577	0.00637

62	Θ_1	0.00140	0.00647	0.00743	0.00853	0.02320	0.00923	0.01030
63	Θ_1	0.00000	0.00207	0.00370	0.00547	0.01073	0.00450	0.00493
64	Θ_1	0.00033	0.00280	0.00450	0.00667	0.01313	0.00557	0.00615
65	Θ_1	0.00220	0.00807	0.00970	0.01140	0.03187	0.01323	0.01579
66	Θ_1	0.00500	0.00860	0.01297	0.01893	0.03100	0.01690	0.01999
67	Θ_1	0.00607	0.01100	0.01583	0.02060	0.03487	0.01863	0.02115
68	Θ_1	0.00127	0.00400	0.00657	0.01033	0.02033	0.00890	0.01026
69	Θ_1	0.00180	0.00513	0.00783	0.01147	0.02307	0.00983	0.01106
70	Θ_1	0.00247	0.00407	0.00650	0.00993	0.01407	0.00850	0.00958
71	Θ_1	0.00313	0.00600	0.00763	0.00967	0.01587	0.00970	0.01091
72	Θ_1	0.00807	0.01467	0.01903	0.02920	0.04853	0.02570	0.03378
73	Θ_1	0.00087	0.00527	0.00837	0.01240	0.03240	0.01070	0.01207
74	Θ_1	0.00480	0.00767	0.01503	0.02647	0.03973	0.01930	0.02367
75	Θ_1	0.00180	0.00507	0.00763	0.01100	0.02167	0.00943	0.01054
76	Θ_1	0.00047	0.00307	0.00490	0.00727	0.01433	0.00603	0.00673
77	Θ_1	0.00100	0.00487	0.00630	0.00807	0.01913	0.00817	0.00919
78	Θ_1	0.00540	0.00540	0.01130	0.02127	0.02127	0.01417	0.01602
79	Θ_1	0.00033	0.00287	0.00477	0.00713	0.01453	0.00597	0.00669
80	Θ_1	0.00133	0.00753	0.00883	0.01013	0.03467	0.01210	0.01458
81	Θ_1	0.00687	0.00813	0.01950	0.04200	0.04800	0.02310	0.02705
82	Θ_1	0.00093	0.00380	0.00590	0.00873	0.01753	0.00737	0.00825
83	Θ_1	0.00367	0.01067	0.01197	0.01347	0.03287	0.01497	0.01689
84	Θ_1	0.00007	0.00247	0.00417	0.00620	0.01233	0.00510	0.00569

Locus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
85	Θ_1	0.00020	0.00267	0.00443	0.00660	0.01327	0.00550	0.00611
86	Θ_1	0.00293	0.00567	0.01123	0.02080	0.03347	0.01383	0.01557
87	Θ_1	0.00107	0.00207	0.00370	0.00553	0.00720	0.00450	0.00496
88	Θ_1	0.00033	0.00287	0.00477	0.00713	0.01460	0.00597	0.00671
89	Θ_1	0.00580	0.00960	0.01603	0.02693	0.04333	0.02077	0.02543
90	Θ_1	0.00713	0.00853	0.01710	0.03227	0.03760	0.02063	0.02349
91	Θ_1	0.00900	0.01187	0.01850	0.02940	0.03800	0.02303	0.02739
92	Θ_1	0.00087	0.00360	0.00563	0.00820	0.01627	0.00690	0.00772
93	Θ_1	0.00600	0.01520	0.01577	0.01620	0.03673	0.01950	0.02239
94	Θ_1	0.00280	0.00480	0.00717	0.01040	0.01547	0.00923	0.01039
95	Θ_1	0.00760	0.01513	0.01930	0.02400	0.04507	0.02370	0.02988
96	Θ_1	0.00073	0.00360	0.00563	0.00833	0.01667	0.00703	0.00784
97	Θ_1	0.00073	0.00280	0.00463	0.00687	0.01193	0.00570	0.00636
98	Θ_1	0.00153	0.00487	0.00743	0.01087	0.02180	0.00930	0.01043
99	Θ_1	0.00440	0.01047	0.01143	0.01240	0.02700	0.01430	0.01606
100	Θ_1	0.00060	0.00320	0.00510	0.00747	0.01480	0.00630	0.00699
All	Θ_1	0.00567	0.00700	0.00797	0.00887	0.01013	0.00803	0.00794

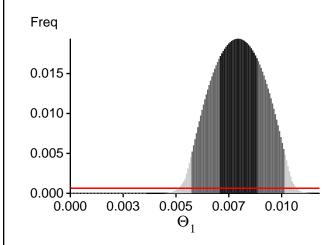
Citation suggestions:

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

ocus	TI(1a)	BTI(1b)	SS(2)	HS(3)
1	-14005.62	-13819.91	-13866.71	-13950.31
2	-14133.40	-13940.47	-13994.13	-14067.39
3	-13999.69	-13825.13	-13872.21	-13954.19
4	-113814.40	-79583.69	-73887.55	-73942.82
5	-14407.59	-14145.74	-14189.89	-14263.71
6	-14092.60	-13898.14	-13954.62	-14027.56
7	-13960.47	-13784.75	-13828.92	-13913.86
8	-14224.20	-13995.97	-14038.56	-14118.42
9	-14059.31	-13881.29	-13939.90	-14011.74
10	-13999.96	-13810.90	-13855.05	-13938.88
11	-14915.82	-14409.77	-14407.32	-14484.19
12	-14264.54	-14034.02	-14084.21	-14157.34
13	-14535.81	-14167.67	-14190.51	-14267.37
14	-13960.73	-13779.73	-13821.16	-13907.97
15	-14517.82	-14163.23	-14188.05	-14263.10
16	-13923.20	-13750.19	-13787.80	-13879.46
17	-13948.07	-13773.68	-13814.48	-13903.72
18	-13909.86	-13737.53	-13774.14	-13868.04
19	-13960.06	-13784.83	-13828.44	-13914.18
20	-14826.69	-14385.89	-14404.26	-14472.96
21	-14067.83	-13872.21	-13919.33	-13999.38
22	-14422.65	-14207.12	-14263.32	-14336.95
23	-15528.72	-14751.43	-14698.96	-14778.00
24	-13962.22	-13789.59	-13832.16	-13919.06
25	-14042.24	-13850.31	-13898.33	-13978.32
26	-14470.66	-14194.39	-14245.31	-14310.17
27	-18131.57	-16935.88	-16822.94	-16907.02
28	-13978.36	-13797.35	-13842.93	-13925.85
29	-13991.99	-13817.59	-13868.21	-13946.74

Migrate 5.0.0a: (http://popgen.sc.fsu.edu) [program run on 08:03:19]

30	-14133.56	-13916.38	-13958.48	-14040.35
31	-13983.11	-13808.22	-13857.62	-13938.67
32	-14101.23	-13918.24	-13973.68	-14048.24
33	-13937.14	-13760.33	-13799.07	-13889.37
34	-14076.02	-13893.94	-13950.91	-14023.95
35	-14438.08	-14124.35	-14152.43	-14232.91
36	-13922.95	-13750.74	-13788.97	-13880.75
37	-14159.43	-13972.58	-14031.17	-14101.94
38	-18680.99	-17072.38	-16891.62	-16965.76
39	-14437.45	-14242.58	-14305.32	-14372.68
40	-14132.18	-13929.20	-13841.84	-14056.30
41	-15133.32	-14711.63	-13851.68	-14804.89
42	-13934.98	-13762.47	-13803.23	-13892.54
43	-14149.17	-13930.78	-13849.92	-14054.47
44	-14196.90	-13987.41	-13842.73	-14113.73
45	-14687.73	-14301.57	-13815.89	-14397.53
46	-14041.78	-13842.78	-13867.48	-13969.10
47	-13906.85	-13734.57	-13770.81	-13864.40
48	-14005.62	-13827.43	-13860.05	-13956.58
49	-13991.24	-13808.88	-13854.45	-13938.09
50	-14043.02	-13846.20	-13837.95	-13973.66
51	-13910.22	-13738.19	-13775.15	-13868.08
52	-15032.59	-14618.37	-13829.62	-14713.97
53	-14034.06	-13845.15	-13873.91	-13974.04
54	-13925.37	-13751.60	-13788.36	-13880.48
55	-16204.08	-15439.48	-13797.84	-15475.29
56	-18611.38	-17805.29	-14012.69	-17861.13
57	-14226.67	-14031.16	-13805.91	-14160.55
58	-14018.66	-13816.05	-13855.12	-13944.53
59	-14063.29	-13871.90	-13925.26	-14000.06
60	-14212.78	-14008.75	-13980.65	-14136.32
61	-13965.17	-13785.90	-13826.88	-13915.27
62	-14033.18	-13841.26	-13887.83	-13969.02
63	-13908.66	-13735.82	-13772.77	-13865.60
64	-13963.51	-13778.24	-13817.88	-13906.62
65	-17467.64	-16495.38	-14084.22	-16505.24
66	-15349.42	-14854.03	-14166.45	-14934.83
67	-14627.16	-14246.14	-14051.07	-14341.35
68	-14844.41	-14587.03	-14630.62	-14711.38
69	-14188.89	-13972.45	-14019.26	-14097.08
70	-14358.83	-14155.76	-14203.33	-14286.47
71	-14047.47	-13853.44	-13900.77	-13981.48
72	-14895.06	-14650.89	-14408.88	-14776.97
73	-14108.33	-13903.29	-13949.57	-14029.74
74	-20864.82	-19634.44	-14268.67	-19629.89

75	-14106.27	-13894.98	-13940.86	-14019.42
76	-13952.20	-13771.88	-13813.10	-13900.15
77	-14464.65	-14139.63	-14163.14	-14248.07
78	-14138.03	-13933.33	-13819.64	-14060.41
79	-13929.87	-13758.21	-13799.62	-13889.01
80	-19791.11	-18027.86	-13867.26	-17901.39
81	-14900.56	-14504.57	-13804.48	-14601.43
82	-13968.12	-13794.77	-13839.09	-13924.71
83	-14354.94	-14086.54	-13801.55	-14203.39
84	-13923.26	-13750.05	-13788.37	-13879.19
85	-13933.16	-13760.31	-13799.68	-13890.70
86	-14102.18	-13914.23	-13897.02	-14043.34
87	-13907.72	-13735.56	-13771.18	-13865.73
88	-13933.58	-13761.64	-13804.10	-13892.49
89	-40088.88	-28785.72	-13937.88	-27118.35
90	-14206.14	-13998.46	-13877.72	-14124.60
91	-14549.72	-14305.44	-13844.25	-14428.48
92	-13987.59	-13800.51	-13783.47	-13928.51
93	-14192.97	-13988.97	-13900.80	-14118.35
94	-14475.81	-14150.24	-14175.59	-14255.78
95	-14921.66	-14638.66	-13851.92	-14757.35
96	-14042.03	-13864.00	-13910.63	-13994.75
97	-13938.21	-13762.25	-13801.02	-13891.10
98	-14031.97	-13841.43	-13889.16	-13970.59
99	-14178.30	-13972.34	-14025.80	-14098.20
100	-13984.99	-13810.88	-13855.24	-13940.45
All	-1576863.34	-1502052.47	-1460268.89	-1505684.90

- (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 = 151.077401]

Citation suggestions:

Beerli P. and M. Palczewski, 2010. Unified framework to evaluate panmixia and migration direction among multiple sampling locations, Genetics, 185: 313-326.

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Acceptance ratios for all parameters and the genealogies

Parameter	Accepted changes	Ratio
Θ_1	296315456/400017262	0.74076
Genealogies	479497071/1599982738	0.29969

MCMC-Autocorrelation and Effective MCMC Sample Size

Parameter	Autocorrelation	Effective Sampe Size
Θ_1	0.27545	16441553.94
Genealogies	0.10130	22932730.93

Average temperatures during the run

Chain Temperatures 1 0.00000 2 0.00000 3 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

4

0.00000

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
The Warning was resorted during the run