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 17:27:00 2017

Program finished at Sat Aug 12 18:44:47 2017 [Runtime:0000:01:17:47]



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) 2065603377

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]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.0.6

Haplotyping is turned on:

Output file: outfile_0.6_0.5

Posterior distribution raw histogram file: bayesfile
Raw data from the MCMC run: bayesallfile_0.6_0.5

Raw data from the MCMC run: bayesallfile_0.6_0.5
Print data: No

Print genealogies [only some for some data type]:

Data summary

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

Mutationmodel:		model:				
	Locus S	ublocus	Mutationmodel	Mutationmodel parameters	Mutationmodel parameters	
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	3	1	Jukes-Cantor	[Basefreq: =0.25]	•	
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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	0.02380	0.03753	0.04750	0.04927	0.05107	0.04023	0.06264
2	Θ_1	0.02507	0.04047	0.04757	0.04933	0.05127	0.04170	0.06919
3	Θ_1	0.01780	0.03147	0.03983	0.04560	0.05047	0.03530	0.04964
4	Θ_1	0.02460	0.03920	0.04757	0.04920	0.05113	0.04063	0.06464
5	Θ_1	0.02227	0.03880	0.04757	0.04920	0.05120	0.04023	0.06549
6	Θ_1	0.02107	0.03787	0.04483	0.04827	0.05080	0.03810	0.05574
7	Θ_1	0.01793	0.03240	0.03943	0.04440	0.05040	0.03523	0.04945
8	Θ_1	0.02693	0.04140	0.04770	0.04947	0.05140	0.04270	0.07310
9	Θ_1	0.02520	0.04127	0.04763	0.04907	0.05133	0.04150	0.06912
10	Θ_1	0.02453	0.03973	0.04770	0.04940	0.05133	0.04110	0.06788
11	Θ_1	0.01647	0.02720	0.03137	0.04133	0.05007	0.03370	0.04622
12	Θ_1	0.02673	0.04187	0.04763	0.04927	0.05133	0.04217	0.07010
13	Θ_1	0.01793	0.03200	0.03703	0.04627	0.05040	0.03543	0.04947
14	Θ_1	0.01980	0.03607	0.04730	0.04840	0.05067	0.03730	0.05455
15	Θ_1	0.02440	0.04040	0.04750	0.04887	0.05107	0.04057	0.06301
16	Θ_1	0.02187	0.03713	0.04750	0.04873	0.05087	0.03883	0.05802
17	Θ_1	0.02367	0.03967	0.04757	0.04940	0.05133	0.04103	0.06895
18	Θ_1	0.02053	0.03773	0.04717	0.04840	0.05073	0.03790	0.05562

19	Θ_1	0.01967	0.03560	0.04137	0.04820	0.05060	0.03690	0.05273
20	Θ_1	0.02007	0.03560	0.04443	0.04833	0.05073	0.03743	0.05471
21	Θ_1	0.01973	0.03660	0.04717	0.04820	0.05073	0.03723	0.05452
22	Θ_1	0.02220	0.03820	0.04750	0.04880	0.05100	0.03923	0.06063
23	Θ_1	0.02567	0.04180	0.04763	0.04933	0.05133	0.04203	0.06849
24	Θ_1	0.02067	0.03460	0.04497	0.04880	0.05080	0.03790	0.05498
25	Θ_1	0.01647	0.02607	0.03543	0.04493	0.05007	0.03383	0.04598
26	Θ_1	0.02560	0.04053	0.04763	0.04947	0.05133	0.04183	0.06983
27	Θ_1	0.01740	0.03473	0.04150	0.04787	0.05107	0.03670	0.05255
28	Θ_1	0.02153	0.03700	0.04750	0.04873	0.05087	0.03857	0.05839
29	Θ_1	0.02420	0.03933	0.04757	0.04920	0.05127	0.04083	0.06432
30	Θ_1	0.02293	0.02467	0.04750	0.05080	0.05100	0.03963	0.06116
31	Θ_1	0.03027	0.04393	0.04783	0.04940	0.05153	0.04410	0.07835
32	Θ_1	0.02953	0.04233	0.04777	0.04967	0.05140	0.04357	0.07531
33	Θ_1	0.02447	0.04073	0.04750	0.04900	0.05120	0.04097	0.06557
34	Θ_1	0.02333	0.03873	0.04757	0.04913	0.05120	0.04023	0.06318
35	Θ_1	0.02913	0.04220	0.04777	0.04967	0.05147	0.04343	0.07575
36	Θ_1	0.02427	0.04053	0.04757	0.04900	0.05120	0.04070	0.06416
37	Θ_1	0.01893	0.03513	0.03783	0.04740	0.05053	0.03630	0.05135
38	Θ_1	0.02073	0.03640	0.04443	0.04827	0.05073	0.03777	0.05468
39	Θ_1	0.01880	0.03647	0.04290	0.04800	0.05067	0.03677	0.05307
40	Θ_1	0.02813	0.04280	0.04777	0.04933	0.05147	0.04303	0.07412
41	Θ_1	0.01993	0.03513	0.04250	0.04813	0.05067	0.03710	0.05363

_ocus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
42	Θ_1	0.02160	0.03700	0.04337	0.04860	0.05087	0.03863	0.05778
43	Θ_1	0.02213	0.03893	0.04750	0.04867	0.05093	0.03910	0.05926
44	Θ_1	0.02413	0.04013	0.04763	0.04953	0.05133	0.04143	0.07023
45	Θ_1	0.02787	0.04207	0.04777	0.04980	0.05147	0.04323	0.07705
46	Θ_1	0.02053	0.03760	0.04377	0.04813	0.05073	0.03777	0.05456
47	Θ_1	0.02980	0.04353	0.04777	0.04967	0.05173	0.04483	0.08249
48	Θ_1	0.01733	0.03007	0.03797	0.04687	0.05040	0.03517	0.04958
49	Θ_1	0.01953	0.03487	0.04323	0.04720	0.05060	0.03677	0.05196
50	Θ_1	0.02240	0.03793	0.04750	0.04900	0.05100	0.03950	0.06206
51	Θ_1	0.02160	0.03787	0.04743	0.04860	0.05087	0.03863	0.05692
52	Θ_1	0.02507	0.04100	0.04763	0.04933	0.05127	0.04157	0.06888
53	Θ_1	0.02567	0.04060	0.04757	0.04920	0.05120	0.04143	0.06745
54	Θ_1	0.02813	0.04187	0.04777	0.04953	0.05147	0.04317	0.07483
55	Θ_1	0.01633	0.02660	0.03450	0.04340	0.05013	0.03370	0.04616
56	Θ_1	0.02253	0.03933	0.04757	0.04880	0.05100	0.03950	0.06077
57	Θ_1	0.01753	0.03180	0.03930	0.04513	0.05047	0.03510	0.04943
58	Θ_1	0.02887	0.04253	0.04770	0.04967	0.05147	0.04377	0.07806
59	Θ_1	0.02553	0.04087	0.04763	0.04920	0.05127	0.04150	0.06728
60	Θ_1	0.01940	0.03500	0.04083	0.04813	0.05060	0.03670	0.05303
61	Θ_1	0.02233	0.03760	0.04757	0.04887	0.05100	0.03930	0.06026

62	Θ_1	0.02293	0.03860	0.04757	0.04920	0.05113	0.04010	0.06158
63	Θ_1	0.01780	0.03187	0.03870	0.04427	0.05040	0.03523	0.04945
64	Θ_1	0.02033	0.03653	0.04750	0.04847	0.05087	0.03777	0.05595
65	Θ_1	0.01940	0.03480	0.04263	0.04813	0.05067	0.03683	0.05249
66	Θ_1	0.02027	0.03627	0.04457	0.04800	0.05060	0.03723	0.05406
67	Θ_1	0.01653	0.02793	0.03550	0.04220	0.05007	0.03383	0.04592
68	Θ_1	0.02200	0.03827	0.04630	0.04860	0.05087	0.03877	0.05824
69	Θ_1	0.02100	0.03793	0.04750	0.04840	0.05080	0.03810	0.05585
70	Θ_1	0.03000	0.04360	0.04777	0.04947	0.05153	0.04377	0.07743
71	Θ_1	0.02507	0.04013	0.04757	0.04940	0.05127	0.04150	0.06775
72	Θ_1	0.02613	0.04187	0.04777	0.04940	0.05147	0.04230	0.07315
73	Θ_1	0.02967	0.04280	0.04777	0.04960	0.05160	0.04410	0.07778
74	Θ_1	0.02527	0.04047	0.04763	0.04940	0.05133	0.04183	0.07177
75	Θ_1	0.02100	0.03707	0.04630	0.04853	0.05080	0.03830	0.05635
76	Θ_1	0.02293	0.03827	0.04750	0.04907	0.05100	0.03977	0.06169
77	Θ_1	0.02547	0.04120	0.04763	0.04927	0.05133	0.04163	0.06890
78	Θ_1	0.01773	0.03193	0.03730	0.04633	0.05033	0.03523	0.04962
79	Θ_1	0.02720	0.04167	0.04770	0.04947	0.05133	0.04263	0.07376
80	Θ_1	0.01920	0.03367	0.03870	0.04727	0.05053	0.03637	0.05132
81	Θ_1	0.02053	0.03633	0.04503	0.04827	0.05073	0.03763	0.05466
82	Θ_1	0.01580	0.02740	0.03423	0.04327	0.05020	0.03390	0.04618
83	Θ_1	0.02100	0.03753	0.04297	0.04833	0.05073	0.03797	0.05541
84	Θ_1	0.02553	0.04087	0.04763	0.04953	0.05140	0.04217	0.07303

Locus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
85	Θ_1	0.02520	0.04007	0.04763	0.04933	0.05127	0.04143	0.06833
86	Θ_1	0.01767	0.03173	0.03830	0.04620	0.05047	0.03550	0.04988
87	Θ_1	0.02093	0.03807	0.04697	0.04853	0.05087	0.03830	0.05703
88	Θ_1	0.02367	0.03947	0.04750	0.04900	0.05113	0.04043	0.06508
89	Θ_1	0.02373	0.03887	0.04750	0.04913	0.05113	0.04037	0.06309
90	Θ_1	0.02713	0.04133	0.04770	0.04953	0.05140	0.04263	0.07289
91	Θ_1	0.02320	0.04007	0.04757	0.04900	0.05107	0.04023	0.06403
92	Θ_1	0.02067	0.03587	0.04230	0.04833	0.05073	0.03770	0.05484
93	Θ_1	0.03133	0.04340	0.04790	0.04980	0.05160	0.04457	0.08311
94	Θ_1	0.02600	0.04167	0.04770	0.04927	0.05133	0.04183	0.06833
95	Θ_1	0.03080	0.04307	0.04783	0.04967	0.05153	0.04430	0.07894
96	Θ_1	0.02273	0.03967	0.04750	0.04887	0.05107	0.03983	0.06200
97	Θ_1	0.02720	0.04227	0.04770	0.04933	0.05140	0.04263	0.07292
98	Θ_1	0.01987	0.03580	0.04517	0.04840	0.05080	0.03750	0.05466
99	Θ_1	0.02427	0.04040	0.04757	0.04900	0.05120	0.04063	0.06434
100	Θ_1	0.02580	0.04060	0.04757	0.04947	0.05133	0.04197	0.06943
All	Θ_1	0.04660	0.04807	0.04910	0.05007	0.05153	0.04917	0.05954

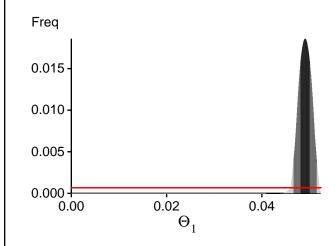
Citation suggestions:

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	-14011.20	-13858.74	-13912.56	-13994.73
2	-14030.58	-13863.32	-13918.66	-13996.37
3	-13885.43	-13741.23	-13787.27	-13878.28
4	-14019.42	-13842.91	-13893.90	-13974.71
5	-16970.38	-15918.06	-15818.94	-15911.84
6	-13956.34	-13798.67	-13847.28	-13932.73
7	-13891.38	-13743.58	-13786.96	-13878.97
8	-14850.07	-14459.75	-14483.65	-14557.04
9	-17109.51	-15829.61	-15695.73	-15771.69
10	-16401.54	-15454.38	-15378.97	-15456.02
11	-13874.07	-13730.03	-13773.35	-13866.48
12	-14156.18	-13939.51	-13989.24	-14064.43
13	-13885.95	-13740.39	-13785.24	-13876.67
14	-13910.17	-13762.40	-13810.79	-13898.24
15	-14063.20	-13878.44	-13928.27	-14008.06
16	-14055.06	-13861.83	-13906.89	-13990.19
17	-24915.02	-20885.17	-20262.41	-20366.33
18	-13921.30	-13769.93	-13817.75	-13906.09
19	-13974.86	-13795.26	-13838.32	-13925.43
20	-13923.59	-13773.35	-13821.86	-13908.86
21	-13912.10	-13764.26	-13812.77	-13900.28
22	-13935.97	-13788.24	-13839.33	-13923.22
23	-14246.61	-13988.02	-14027.91	-14105.95
24	-13938.44	-13782.25	-13829.06	-13915.54
25	-13873.70	-13729.53	-13773.22	-13865.96
26	-14114.93	-13924.67	-13975.03	-14054.00
27	-13905.83	-13756.20	-13801.86	-13891.37
28	-13924.48	-13776.78	-13825.34	-13911.41
29	-14221.86	-13979.57	-14019.06	-14100.11

30	-13964.77	-13801.72	-13852.46	-13935.11
31	-14797.69	-14436.61	-14465.13	-14537.89
32	-14407.22	-14143.00	-14189.12	-14260.91
33	-14179.04	-14008.79	-14065.16	-14143.09
34	-14024.82	-13846.03	-13895.35	-13976.64
35	-17783.36	-16341.54	-16189.00	-16258.50
36	-14056.61	-13875.54	-13925.92	-14006.49
37	-13914.06	-13768.47	-13814.97	-13905.50
38	-13953.25	-13784.43	-13829.69	-13916.06
39	-13902.04	-13756.13	-13803.41	-13892.16
40	-14748.87	-14361.07	-14385.70	-14458.11
41	-13922.67	-13768.06	-13814.35	-13903.09
42	-13939.85	-13779.90	-13828.69	-13913.28
43	-13939.94	-13786.32	-13835.42	-13920.70
44	-14118.27	-13963.85	-14013.71	-14097.13
45	-14368.79	-14171.24	-14224.78	-14298.43
46	-14043.30	-13841.76	-13882.35	-13970.52
47	-14830.32	-14472.57	-14511.15	-14575.69
48	-13887.22	-13742.86	-13788.93	-13879.76
49	-13931.90	-13777.16	-13823.32	-13912.58
50	-20529.69	-17140.92	-16611.10	-16692.13
51	-13983.74	-13819.44	-13867.99	-13952.65
52	-14035.40	-13874.42	-13929.76	-14008.35
53	-14058.58	-13871.39	-13922.12	-14000.19
54	-15508.73	-14921.58	-14914.72	-14986.62
55	-13872.48	-13728.47	-13770.61	-13864.92
56	-13943.05	-13790.33	-13840.97	-13925.98
57	-13886.53	-13742.24	-13788.39	-13878.88
58	-14597.32	-14338.14	-14392.00	-14460.00
59	-14042.19	-13882.56	-13939.87	-14017.02
60	-13904.49	-13756.28	-13802.51	-13892.99
61	-13950.51	-13797.28	-13847.91	-13932.87
62	-14168.04	-13929.37	-13968.72	-14050.09
63	-13885.93	-13741.60	-13787.44	-13878.02
64	-13915.83	-13768.83	-13817.42	-13904.08
65	-13961.46	-13789.05	-13832.96	-13920.20
66	-13960.82	-13810.99	-13860.40	-13947.06
67	-13873.39	-13729.23	-13772.68	-13866.49
68	-14029.33	-13854.64	-13902.49	-13986.17
69	-14113.42	-13896.40	-13935.12	-14020.82
70	-14731.13	-14399.33	-14439.02	-14507.79
71	-14023.08	-13852.03	-13905.69	-13984.14
72	-14328.28	-14126.06	-14177.37	-14253.48
73	-14461.82	-14161.57	-14201.89	-14271.88
74	-14427.44	-14244.56	-14291.64	-14373.61

75	-14064.95	-13903.55	-13953.98	-14038.77
76	-13959.73	-13809.45	-13861.64	-13944.72
77	-14342.01	-14089.29	-14132.37	-14208.52
78	-13886.01	-13741.71	-13786.92	-13878.22
79	-14461.03	-14229.63	-14278.50	-14352.42
80	-13933.08	-13771.15	-13816.06	-13906.28
81	-14030.09	-13840.90	-13883.59	-13969.78
82	-13874.29	-13730.19	-13773.78	-13866.56
83	-13937.88	-13782.73	-13829.89	-13916.89
84	-15704.00	-15375.54	-15395.19	-15478.26
85	-14007.83	-13842.80	-13897.35	-13975.71
86	-13888.26	-13742.79	-13787.33	-13880.33
87	-13937.10	-13780.65	-13829.87	-13914.68
88	-13965.13	-13811.99	-13865.59	-13946.86
89	-13981.82	-13821.27	-13873.06	-13954.69
90	-18956.40	-16836.84	-16556.26	-16629.28
91	-15008.99	-14494.98	-14486.54	-14570.25
92	-13949.36	-13784.80	-13831.60	-13917.26
93	-14974.28	-14624.80	-14670.18	-14730.68
94	-14109.09	-13918.94	-13971.09	-14048.79
95	-14554.83	-14322.03	-14380.59	-14447.52
96	-13955.40	-13798.47	-13849.48	-13933.09
97	-14236.33	-14022.01	-14074.69	-14149.24
98	-13912.96	-13764.98	-13812.00	-13900.18
99	-14008.49	-13836.94	-13888.95	-13969.40
100	-14914.18	-14478.44	-14491.29	-14567.48
All	-1447313.07	-1414660.43	-1416972.51	-1425238.64

- (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 = 26.292218]

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	384108904/400022158 505675462/1599977842	0.96022 0.31605

MCMC-Autocorrelation and Effective MCMC Sample Size

Parameter	Autocorrelation	Effective Sampe Size
Θ_1	0.67366	1956716.33
Genealogies	0.10589	8365140.92

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