## **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 15:05:07 2017

Program finished at Sat Aug 12 16:03:24 2017 [Runtime:0000:00:58:17]



### **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) 489844354

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.0.4

Haplotyping is turned on:

Output file: outfile\_0.4\_0.6

Posterior distribution raw histogram file: bayesfile
Raw data from the MCMC run: bayesallfile\_0.4\_0.6

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]
3	1	Jukes-Cantor	[Basefreq: =0.25]
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# Bayesian Analysis: Posterior distribution table

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Locus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
1	$\Theta_1$	0.00293	0.00907	0.01090	0.01280	0.03240	0.01397	0.01607
2	$\Theta_1$	0.00820	0.00993	0.01577	0.02373	0.02833	0.01930	0.02218
3	$\Theta_1$	0.00293	0.00920	0.01063	0.01227	0.03240	0.01390	0.01602
4	$\Theta_1$	0.00380	0.00693	0.01050	0.01600	0.02687	0.01390	0.01607
5	$\Theta_1$	0.00340	0.00800	0.01070	0.01407	0.02893	0.01390	0.01606
6	$\Theta_1$	0.00307	0.00493	0.01083	0.02193	0.03093	0.01390	0.01607
7	$\Theta_1$	0.00393	0.00940	0.01230	0.01593	0.03360	0.01577	0.01822
8	$\Theta_1$	0.01547	0.02793	0.03617	0.04380	0.05013	0.03350	0.04689
9	$\Theta_1$	0.01147	0.01733	0.02250	0.02813	0.04280	0.02630	0.03214
10	$\Theta_1$	0.00400	0.00740	0.01070	0.01520	0.02560	0.01390	0.01602
11	$\Theta_1$	0.00440	0.00660	0.01063	0.01687	0.02407	0.01397	0.01613
12	$\Theta_1$	0.00787	0.01320	0.01983	0.02673	0.04513	0.02383	0.02960
13	$\Theta_1$	0.00380	0.00760	0.01270	0.02027	0.03700	0.01603	0.01833
14	$\Theta_1$	0.00387	0.01040	0.01250	0.01480	0.03607	0.01603	0.01848
15	$\Theta_1$	0.00287	0.00707	0.01090	0.01607	0.03287	0.01390	0.01603
16	$\Theta_1$	0.00480	0.01033	0.01077	0.01113	0.02227	0.01397	0.01608
17	$\Theta_1$	0.00307	0.00600	0.01083	0.01867	0.03133	0.01397	0.01611
18	$\Theta_1$	0.00280	0.00840	0.01230	0.01713	0.04267	0.01570	0.01820
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19	$\Theta_1$	0.00513	0.00893	0.01650	0.02893	0.04633	0.02017	0.02337
20	$\Theta_1$	0.00360	0.00867	0.01063	0.01300	0.02787	0.01390	0.01602
21	$\Theta_1$	0.00453	0.00827	0.01063	0.01380	0.02313	0.01390	0.01602
22	$\Theta_1$	0.00693	0.00693	0.01457	0.03033	0.03033	0.01877	0.02165
23	$\Theta_1$	0.00220	0.00947	0.01063	0.01200	0.03867	0.01397	0.01603
24	$\Theta_1$	0.00240	0.00860	0.01070	0.01307	0.03580	0.01390	0.01607
25	$\Theta_1$	0.00213	0.00467	0.01083	0.02273	0.03887	0.01390	0.01605
26	$\Theta_1$	0.00393	0.00393	0.01070	0.02607	0.02607	0.01397	0.01609
27	$\Theta_1$	0.00220	0.00573	0.01083	0.01913	0.03820	0.01390	0.01601
28	$\Theta_1$	0.00293	0.00680	0.01077	0.01673	0.03260	0.01403	0.01611
29	$\Theta_1$	0.00253	0.00653	0.01077	0.01720	0.03520	0.01397	0.01605
30	$\Theta_1$	0.00500	0.00807	0.01230	0.01793	0.02713	0.01557	0.01787
31	$\Theta_1$	0.00240	0.00793	0.01090	0.01440	0.03587	0.01390	0.01601
32	$\Theta_1$	0.00300	0.00700	0.01077	0.01600	0.03193	0.01390	0.01599
33	$\Theta_1$	0.00460	0.00460	0.01077	0.02307	0.02307	0.01397	0.01614
34	$\Theta_1$	0.00453	0.00567	0.01077	0.01947	0.02340	0.01397	0.01603
35	$\Theta_1$	0.00287	0.00700	0.01057	0.01607	0.03293	0.01397	0.01599
36	$\Theta_1$	0.00333	0.00747	0.01357	0.02467	0.04733	0.01750	0.02017
37	$\Theta_1$	0.00407	0.01007	0.01417	0.01913	0.04247	0.01777	0.02052
38	$\Theta_1$	0.00393	0.00833	0.01050	0.01333	0.02593	0.01390	0.01600
39	$\Theta_1$	0.00340	0.00580	0.01063	0.01927	0.02927	0.01397	0.01607
40	$\Theta_1$	0.00427	0.00767	0.01070	0.01467	0.02453	0.01390	0.01606
41	$\Theta_1$	0.00273	0.00773	0.01057	0.01447	0.03407	0.01397	0.01606

_ocus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
42	$\Theta_1$	0.00453	0.01080	0.01430	0.01893	0.04093	0.01803	0.02070
43	$\Theta_1$	0.00293	0.00513	0.01070	0.02140	0.03240	0.01397	0.01609
44	$\Theta_1$	0.00220	0.00700	0.01077	0.01600	0.03813	0.01390	0.01599
45	$\Theta_1$	0.00373	0.00547	0.01083	0.02013	0.02767	0.01397	0.01614
46	$\Theta_1$	0.00273	0.00767	0.01083	0.01480	0.03373	0.01403	0.01612
47	$\Theta_1$	0.00353	0.00813	0.01063	0.01400	0.02827	0.01397	0.01607
48	$\Theta_1$	0.00287	0.00707	0.01063	0.01613	0.03293	0.01397	0.01605
49	$\Theta_1$	0.00327	0.00373	0.01070	0.02720	0.03047	0.01397	0.01612
50	$\Theta_1$	0.00347	0.00373	0.01083	0.02707	0.02873	0.01397	0.01600
51	$\Theta_1$	0.00453	0.00947	0.01370	0.01933	0.03727	0.01737	0.02001
52	$\Theta_1$	0.00367	0.00807	0.01070	0.01413	0.02780	0.01397	0.01608
53	$\Theta_1$	0.00193	0.00813	0.01083	0.01413	0.04073	0.01390	0.01602
54	$\Theta_1$	0.00473	0.01000	0.01357	0.01927	0.03740	0.01790	0.02080
55	$\Theta_1$	0.00267	0.00740	0.01070	0.01513	0.03433	0.01403	0.01611
56	$\Theta_1$	0.00547	0.01447	0.01537	0.01620	0.04073	0.01943	0.02238
57	$\Theta_1$	0.00373	0.00767	0.01057	0.01473	0.02707	0.01390	0.01599
58	$\Theta_1$	0.00880	0.01313	0.01957	0.02847	0.04233	0.02423	0.03011
59	$\Theta_1$	0.00727	0.01373	0.01930	0.02633	0.04673	0.02350	0.02833
60	$\Theta_1$	0.00360	0.00360	0.01097	0.02767	0.02767	0.01397	0.01608
61	$\Theta_1$	0.00500	0.00573	0.01317	0.02727	0.03073	0.01657	0.01915

62	$\Theta_1$	0.00433	0.00747	0.01077	0.01533	0.02393	0.01397	0.01607
63	$\Theta_1$	0.00487	0.00700	0.01090	0.01613	0.02213	0.01397	0.01615
64	$\Theta_1$	0.00393	0.00853	0.01290	0.01947	0.03733	0.01670	0.01935
65	$\Theta_1$	0.00373	0.00767	0.01063	0.01480	0.02760	0.01397	0.01606
66	$\Theta_1$	0.01527	0.02433	0.02863	0.03920	0.04967	0.03217	0.04236
67	$\Theta_1$	0.00407	0.01100	0.01270	0.01440	0.03673	0.01643	0.01889
68	$\Theta_1$	0.00480	0.01020	0.01257	0.01513	0.03047	0.01603	0.01845
69	$\Theta_1$	0.00287	0.00667	0.01070	0.01667	0.03260	0.01390	0.01606
70	$\Theta_1$	0.00487	0.00693	0.01050	0.01613	0.02193	0.01390	0.01608
71	$\Theta_1$	0.00293	0.00933	0.01057	0.01187	0.03233	0.01397	0.01609
72	$\Theta_1$	0.00273	0.00480	0.01203	0.02873	0.04333	0.01577	0.01823
73	$\Theta_1$	0.00747	0.01220	0.01483	0.01827	0.02887	0.01897	0.02186
74	$\Theta_1$	0.00267	0.01007	0.01090	0.01167	0.03380	0.01390	0.01598
75	$\Theta_1$	0.00380	0.01067	0.01290	0.01587	0.04040	0.01670	0.01918
76	$\Theta_1$	0.00227	0.00787	0.01070	0.01420	0.03720	0.01390	0.01599
77	$\Theta_1$	0.00567	0.01007	0.01390	0.01887	0.03213	0.01750	0.02002
78	$\Theta_1$	0.00373	0.00840	0.01230	0.01733	0.03433	0.01563	0.01802
79	$\Theta_1$	0.00633	0.01067	0.01577	0.02280	0.03713	0.02003	0.02345
80	$\Theta_1$	0.00260	0.00880	0.01077	0.01313	0.03480	0.01397	0.01609
81	$\Theta_1$	0.00627	0.01093	0.01730	0.02573	0.04300	0.02097	0.02440
82	$\Theta_1$	0.00440	0.00747	0.01077	0.01520	0.02380	0.01390	0.01599
83	$\Theta_1$	0.00460	0.00887	0.01263	0.01793	0.03193	0.01617	0.01862
84	$\Theta_1$	0.00407	0.00700	0.01057	0.01613	0.02567	0.01397	0.01608

Locus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
85	$\Theta_1$	0.01087	0.01613	0.02123	0.02913	0.04140	0.02577	0.03141
86	$\Theta_1$	0.00380	0.00740	0.01043	0.01520	0.02687	0.01397	0.01606
87	$\Theta_1$	0.00273	0.00707	0.01083	0.01620	0.03340	0.01397	0.01607
88	$\Theta_1$	0.00827	0.01133	0.01930	0.03147	0.04267	0.02343	0.02791
89	$\Theta_1$	0.00480	0.00693	0.01057	0.01593	0.02220	0.01390	0.01602
90	$\Theta_1$	0.00333	0.00720	0.01083	0.01560	0.02947	0.01397	0.01604
91	$\Theta_1$	0.00467	0.00973	0.01070	0.01173	0.02307	0.01397	0.01606
92	$\Theta_1$	0.00540	0.01160	0.01630	0.02247	0.04507	0.02010	0.02329
93	$\Theta_1$	0.00500	0.00613	0.01077	0.01813	0.02147	0.01397	0.01601
94	$\Theta_1$	0.01367	0.02233	0.02963	0.03840	0.04967	0.03137	0.04304
95	$\Theta_1$	0.00607	0.01127	0.01650	0.02440	0.04240	0.02063	0.02390
96	$\Theta_1$	0.00440	0.01060	0.01077	0.01087	0.02407	0.01397	0.01610
97	$\Theta_1$	0.00467	0.00940	0.01190	0.01507	0.02833	0.01543	0.01767
98	$\Theta_1$	0.00933	0.01573	0.02110	0.02707	0.04380	0.02483	0.02945
99	$\Theta_1$	0.00773	0.01553	0.01930	0.02313	0.04600	0.02390	0.02944
100	$\Theta_1$	0.00273	0.00787	0.01057	0.01413	0.03340	0.01390	0.01600
All	$\Theta_1$	0.00980	0.01133	0.01237	0.01333	0.01487	0.01243	0.01240

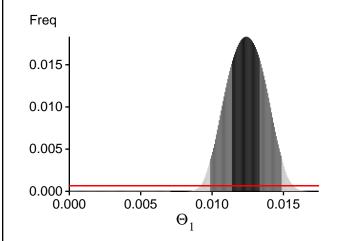
#### Citation suggestions:

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	-13884.30	-13730.99	-13772.42	-13865.17
2	-14052.81	-13854.98	-13896.77	-13983.53
3	-13885.09	-13731.68	-13772.62	-13866.62
4	-13886.05	-13732.86	-13773.84	-13867.19
5	-13885.13	-13731.65	-13772.97	-13866.29
6	-13886.28	-13732.71	-13773.27	-13867.24
7	-13897.70	-13744.29	-13786.95	-13878.85
8	-14144.64	-13971.87	-14026.07	-14104.27
9	-13972.38	-13811.40	-13864.14	-13947.10
10	-13885.47	-13732.12	-13773.56	-13866.38
11	-13886.61	-13733.51	-13774.50	-13868.70
12	-25683.52	-19520.92	-18486.24	-18565.28
13	-13917.00	-13755.31	-13797.81	-13888.14
14	-13918.44	-13756.92	-13799.45	-13891.31
15	-13883.66	-13730.40	-13771.39	-13864.67
16	-13884.48	-13731.35	-13772.03	-13866.31
17	-13884.49	-13731.33	-13771.78	-13865.49
18	-13900.82	-13746.69	-13789.01	-13882.08
19	-14134.47	-13912.10	-13950.80	-14035.43
20	-13886.51	-13733.10	-13773.96	-13867.49
21	-13886.81	-13733.63	-13774.77	-13868.08
22	-13927.59	-13768.88	-13813.88	-13902.18
23	-13886.64	-13733.15	-13774.01	-13867.16
24	-13885.68	-13732.21	-13773.07	-13867.04
25	-13886.92	-13733.48	-13774.90	-13867.57
26	-13883.58	-13730.15	-13771.40	-13864.43
27	-13886.99	-13733.58	-13774.34	-13867.86
28	-13883.10	-13730.06	-13770.66	-13864.52
29	-13887.02	-13733.61	-13773.84	-13868.58

Migrate 5.0.0a: (http://popgen.sc.fsu.edu) [program run on 15:05:07]

30	-13902.17	-13746.93	-13788.32	-13881.11
31	-13886.67	-13733.39	-13774.32	-13867.77
32	-13886.73	-13733.26	-13774.42	-13867.78
33	-13884.98	-13731.54	-13771.81	-13866.13
34	-13885.98	-13732.83	-13774.57	-13867.39
35	-13885.53	-13732.17	-13773.11	-13868.29
36	-13916.07	-13758.98	-13802.37	-13895.60
37	-13916.63	-13759.61	-13802.93	-13893.15
38	-13885.75	-13732.22	-13773.29	-13866.44
39	-13886.36	-13732.91	-13772.57	-13866.92
40	-13886.47	-13733.07	-13772.40	-13867.40
41	-13886.24	-13733.01	-13774.20	-13867.85
42	-13953.10	-13797.06	-13843.19	-13931.06
43	-13885.54	-13732.18	-13771.67	-13866.38
44	-13885.71	-13732.50	-13773.61	-13867.90
45	-13885.66	-13732.47	-13773.06	-13866.92
46	-13886.42	-13733.01	-13774.14	-13867.15
47	-13886.59	-13732.98	-13774.14	-13867.22
48	-13885.83	-13732.48	-13773.44	-13867.64
49	-13886.40	-13733.00	-13774.27	-13867.73
50	-13886.36	-13733.19	-13773.38	-13867.62
51	-13913.36	-13759.07	-13803.22	-13892.95
52	-13883.05	-13729.60	-13770.29	-13863.78
53	-13886.95	-13733.43	-13774.67	-13870.54
54	-13908.02	-13755.02	-13801.54	-13889.75
55	-13886.93	-13733.58	-13774.65	-13867.89
56	-13940.58	-13776.91	-13823.27	-13912.91
57	-13885.58	-13732.30	-13773.25	-13867.57
58	-20659.68	-19028.74	-18862.53	-18941.81
59	-15921.38	-14917.66	-14819.57	-14903.52
60	-13883.81	-13730.20	-13771.27	-13864.95
61	-13904.48	-13750.87	-13793.54	-13885.09
62	-13885.95	-13733.13	-13773.86	-13867.53
63	-13886.72	-13733.34	-13774.44	-13867.60
64	-13910.09	-13756.42	-13799.78	-13891.66
65	-13885.85	-13732.59	-13773.44	-13867.18
66	-14040.51	-13883.58	-13940.80	-14018.39
67	-13925.11	-13763.34	-13806.35	-13896.14
68	-13916.80	-13755.87	-13798.52	-13888.96
69	-13881.38	-13728.14	-13769.10	-13862.53
70	-13885.95	-13732.85	-13774.13	-13867.61
71	-13884.79	-13731.55	-13772.44	-13865.82
72	-13899.05	-13745.40	-13789.16	-13881.49
73	-13926.50	-13769.64	-13814.78	-13902.54
74	-13884.90	-13731.45	-13771.96	-13865.46

75	-13938.80	-13773.55	-13816.29	-13906.01
76	-13886.66	-13733.25	-13774.61	-13868.89
77	-13973.27	-13795.29	-13837.95	-13925.68
78	-13899.38	-13744.26	-13785.91	-13877.56
79	-13921.25	-13768.08	-13815.92	-13902.99
80	-13884.03	-13730.65	-13771.80	-13864.81
81	-13942.54	-13782.64	-13830.36	-13917.65
82	-13885.44	-13732.06	-13773.06	-13867.94
83	-13913.76	-13754.81	-13797.87	-13888.09
84	-13882.98	-13729.78	-13770.59	-13864.46
85	-14289.12	-14046.45	-14085.84	-14167.45
86	-13886.52	-13733.16	-13774.22	-13867.65
87	-13886.84	-13733.53	-13774.71	-13868.03
88	-13950.60	-13794.52	-13844.06	-13928.10
89	-13886.51	-13733.33	-13774.17	-13868.18
90	-13886.51	-13733.18	-13774.28	-13867.40
91	-13884.85	-13731.74	-13772.07	-13866.17
92	-13946.06	-13784.20	-13831.49	-13916.82
93	-13885.51	-13732.28	-13772.58	-13867.04
94	-14646.18	-14392.35	-14439.57	-14514.54
95	-14129.99	-13927.92	-13970.99	-14055.09
96	-13884.30	-13731.32	-13771.42	-13865.85
97	-13902.78	-13746.62	-13788.02	-13879.50
98	-14107.37	-13915.04	-13963.46	-14045.71
99	-20205.43	-17063.70	-16581.09	-16661.49
100	-13887.08	-13733.49	-13774.13	-13868.15
All	-1418863.35	-1391652.38	-1393935.51	-1403042.07

- (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 = 143.212986]

#### 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
$\Theta_1$	335133787/400018015	0.83780
Genealogies	1019117984/1599981985	0.63696

## MCMC-Autocorrelation and Effective MCMC Sample Size

Parameter	Autocorrelation	Effective Sampe Size
$\Theta_1$	0.44061	3928921.66
Genealogies	0.06126	8857652.66

## 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