## **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 Sun Aug 13 02:14:37 2017

Program finished at Sun Aug 13 04:02:59 2017 [Runtime:0000:01:48:22]



### **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) 1391188356

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

**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

bayesallfile\_0.6\_0.6

Print options:

Data file: infile.0.6 NO

Haplotyping is turned on:

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

Posterior distribution raw histogram file: bayesfile

Print data: No

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

Raw data from the MCMC run:

# Data summary

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

model:			
	Mutationmodel	Mutationmodel parameters	
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[Basefreq: =0.25]

Jukes-Cantor

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# Bayesian Analysis: Posterior distribution table

Locus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
1	$\Theta_1$	0.00633	0.01047	0.01683	0.02633	0.04233	0.02077	0.02423
2	$\Theta_1$	0.01587	0.02667	0.03297	0.04160	0.05007	0.03337	0.04387
3	$\Theta_1$	0.00253	0.00753	0.01063	0.01473	0.03507	0.01390	0.01600
4	$\Theta_1$	0.01980	0.03500	0.04037	0.04787	0.05060	0.03690	0.05219
5	$\Theta_1$	0.01667	0.02793	0.03570	0.04347	0.05007	0.03397	0.04583
6	$\Theta_1$	0.00553	0.01140	0.01570	0.02147	0.04167	0.01997	0.02342
7	$\Theta_1$	0.00540	0.01133	0.01457	0.01887	0.03727	0.01863	0.02173
8	$\Theta_1$	0.01273	0.01713	0.02350	0.03520	0.04600	0.02797	0.03409
9	$\Theta_1$	0.00267	0.00900	0.01077	0.01267	0.03453	0.01397	0.01616
10	$\Theta_1$	0.01247	0.01813	0.02477	0.03440	0.04827	0.02863	0.03571
11	$\Theta_1$	0.01173	0.01747	0.02383	0.03187	0.04620	0.02723	0.03321
12	$\Theta_1$	0.01673	0.02787	0.03830	0.04313	0.05020	0.03410	0.04677
13	$\Theta_1$	0.00833	0.01560	0.01963	0.02473	0.04607	0.02390	0.02834
14	$\Theta_1$	0.00853	0.01013	0.01503	0.02447	0.02873	0.01983	0.02291
15	$\Theta_1$	0.00700	0.00847	0.01797	0.03640	0.04380	0.02223	0.02625
16	$\Theta_1$	0.00593	0.01160	0.01617	0.02260	0.04260	0.02037	0.02384
17	$\Theta_1$	0.01313	0.01947	0.02737	0.03520	0.04853	0.02943	0.03814
18	$\Theta_1$	0.00887	0.01293	0.01757	0.02340	0.03307	0.02150	0.02503

Migrate 5.0.0a: (http://popgen.sc.fsu.edu) [program run on 02:14:37]

19	$\Theta_1$	0.00553	0.01127	0.01610	0.02200	0.04247	0.01970	0.02259
20	$\Theta_1$	0.00547	0.01180	0.01537	0.02020	0.04213	0.01977	0.02296
21	$\Theta_1$	0.01447	0.02687	0.03123	0.04307	0.05013	0.03303	0.04809
22	$\Theta_1$	0.02160	0.03700	0.04750	0.04873	0.05087	0.03863	0.05743
23	$\Theta_1$	0.01013	0.01400	0.01943	0.02607	0.03600	0.02357	0.02812
24	$\Theta_1$	0.00520	0.01227	0.01677	0.02260	0.04853	0.02070	0.02390
25	$\Theta_1$	0.01127	0.01593	0.02290	0.03260	0.04600	0.02670	0.03206
26	$\Theta_1$	0.01900	0.03407	0.04057	0.04673	0.05047	0.03623	0.05125
27	$\Theta_1$	0.00507	0.00507	0.01210	0.02727	0.02727	0.01570	0.01813
28	$\Theta_1$	0.01093	0.01487	0.02223	0.03187	0.04213	0.02603	0.03250
29	$\Theta_1$	0.00660	0.01207	0.01703	0.02480	0.04387	0.02143	0.02476
30	$\Theta_1$	0.01107	0.01553	0.02223	0.03007	0.04180	0.02577	0.03121
31	$\Theta_1$	0.01093	0.01807	0.02423	0.03340	0.04913	0.02823	0.03484
32	$\Theta_1$	0.00493	0.00907	0.01377	0.02127	0.03593	0.01790	0.02080
33	$\Theta_1$	0.01333	0.01993	0.02637	0.03373	0.04867	0.02930	0.03643
34	$\Theta_1$	0.00667	0.01127	0.01423	0.01707	0.02820	0.01763	0.02026
35	$\Theta_1$	0.00833	0.01407	0.02090	0.03007	0.04793	0.02477	0.02951
36	$\Theta_1$	0.00820	0.01387	0.01923	0.02527	0.04127	0.02290	0.02686
37	$\Theta_1$	0.00287	0.00707	0.01063	0.01613	0.03313	0.01397	0.01610
38	$\Theta_1$	0.01527	0.02267	0.02770	0.03953	0.04947	0.03170	0.04061
39	$\Theta_1$	0.00300	0.00927	0.01083	0.01253	0.03153	0.01390	0.01604
40	$\Theta_1$	0.00260	0.00953	0.01217	0.01560	0.04447	0.01577	0.01828
41	$\Theta_1$	0.00720	0.00933	0.01730	0.03107	0.03907	0.02117	0.02449

Locus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
42	$\Theta_1$	0.01207	0.01907	0.02890	0.03633	0.04927	0.02977	0.04066
43	$\Theta_1$	0.01720	0.03407	0.04283	0.04793	0.05053	0.03570	0.05241
44	$\Theta_1$	0.00560	0.01100	0.01437	0.01900	0.03540	0.01837	0.02120
45	$\Theta_1$	0.00740	0.01193	0.01790	0.02560	0.04107	0.02183	0.02553
46	$\Theta_1$	0.00447	0.00987	0.01377	0.01880	0.03867	0.01750	0.02026
47	$\Theta_1$	0.00313	0.00827	0.01210	0.01773	0.03940	0.01577	0.01828
48	$\Theta_1$	0.01447	0.02167	0.02777	0.03573	0.04880	0.03037	0.03819
49	$\Theta_1$	0.00260	0.00533	0.01230	0.02580	0.04293	0.01557	0.01790
50	$\Theta_1$	0.01147	0.01507	0.01997	0.02973	0.03873	0.02517	0.02973
51	$\Theta_1$	0.00467	0.00907	0.01443	0.02213	0.03993	0.01797	0.02062
52	$\Theta_1$	0.00860	0.01220	0.01903	0.02827	0.03967	0.02297	0.02738
53	$\Theta_1$	0.01640	0.02433	0.03163	0.04040	0.04973	0.03277	0.04226
54	$\Theta_1$	0.01833	0.03460	0.04030	0.04827	0.05067	0.03657	0.05415
55	$\Theta_1$	0.00533	0.00707	0.01577	0.03427	0.04300	0.01997	0.02341
56	$\Theta_1$	0.00767	0.00933	0.01510	0.02373	0.02827	0.01877	0.02145
57	$\Theta_1$	0.01973	0.03753	0.04750	0.04873	0.05093	0.03817	0.06067
58	$\Theta_1$	0.01067	0.01527	0.01970	0.02740	0.03787	0.02450	0.02911
59	$\Theta_1$	0.00760	0.01173	0.01783	0.02580	0.03860	0.02143	0.02487
60	$\Theta_1$	0.00340	0.00640	0.01223	0.02253	0.03713	0.01583	0.01822
61	$\Theta_1$	0.00320	0.00807	0.01203	0.01740	0.03773	0.01543	0.01774

62	$\Theta_1$	0.01060	0.01527	0.02197	0.03127	0.04367	0.02577	0.03073
63	$\Theta_1$	0.00640	0.01540	0.01690	0.01880	0.04287	0.02130	0.02499
64	$\Theta_1$	0.00760	0.00987	0.01410	0.02027	0.02593	0.01803	0.02089
65	$\Theta_1$	0.01053	0.01540	0.02277	0.03067	0.04493	0.02630	0.03332
66	$\Theta_1$	0.01487	0.02640	0.03163	0.04100	0.05007	0.03283	0.04517
67	$\Theta_1$	0.02087	0.03653	0.04377	0.04853	0.05080	0.03823	0.05638
68	$\Theta_1$	0.01593	0.02407	0.03090	0.03747	0.04940	0.03203	0.04089
69	$\Theta_1$	0.00407	0.00880	0.01303	0.01853	0.03660	0.01657	0.01912
70	$\Theta_1$	0.00587	0.00967	0.01437	0.02033	0.03240	0.01783	0.02048
71	$\Theta_1$	0.00580	0.01180	0.01530	0.02053	0.04013	0.01957	0.02265
72	$\Theta_1$	0.01120	0.01873	0.02337	0.02973	0.04827	0.02770	0.03501
73	$\Theta_1$	0.01387	0.01720	0.02590	0.03847	0.04727	0.02903	0.03554
74	$\Theta_1$	0.01773	0.02920	0.03523	0.04393	0.05013	0.03463	0.04666
75	$\Theta_1$	0.01680	0.03013	0.03803	0.04240	0.05020	0.03443	0.04747
76	$\Theta_1$	0.01987	0.03680	0.04310	0.04667	0.05060	0.03697	0.05182
77	$\Theta_1$	0.00507	0.01220	0.01510	0.01840	0.04193	0.01903	0.02207
78	$\Theta_1$	0.00480	0.00973	0.01210	0.01487	0.02767	0.01543	0.01772
79	$\Theta_1$	0.00660	0.01500	0.01770	0.02040	0.04387	0.02163	0.02523
80	$\Theta_1$	0.01053	0.01593	0.02143	0.02867	0.04200	0.02537	0.03046
81	$\Theta_1$	0.00440	0.00940	0.01450	0.02367	0.04640	0.01910	0.02227
82	$\Theta_1$	0.01313	0.01900	0.02663	0.03473	0.04860	0.02917	0.03783
83	$\Theta_1$	0.00540	0.00993	0.01250	0.01607	0.02813	0.01623	0.01860
84	$\Theta_1$	0.01347	0.01973	0.02603	0.03647	0.04873	0.02970	0.03674

Locus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
85	$\Theta_1$	0.00553	0.01147	0.01783	0.02480	0.04847	0.02123	0.02490
86	$\Theta_1$	0.01560	0.02447	0.03143	0.03793	0.04973	0.03230	0.04207
87	$\Theta_1$	0.00387	0.00967	0.01297	0.01740	0.03893	0.01670	0.01923
88	$\Theta_1$	0.01027	0.01520	0.01823	0.02333	0.03440	0.02310	0.02686
89	$\Theta_1$	0.00887	0.01520	0.01917	0.02380	0.04013	0.02323	0.02733
90	$\Theta_1$	0.01033	0.01473	0.02330	0.03273	0.04633	0.02603	0.03180
91	$\Theta_1$	0.00747	0.01267	0.01743	0.02553	0.04140	0.02197	0.02565
92	$\Theta_1$	0.01460	0.02280	0.03397	0.04387	0.04987	0.03243	0.04471
93	$\Theta_1$	0.01387	0.02033	0.02663	0.03573	0.04833	0.02970	0.03697
94	$\Theta_1$	0.00940	0.01333	0.02117	0.03287	0.04613	0.02510	0.03008
95	$\Theta_1$	0.01107	0.01587	0.02030	0.02587	0.03687	0.02450	0.02867
96	$\Theta_1$	0.02173	0.03813	0.04670	0.04853	0.05080	0.03863	0.05796
97	$\Theta_1$	0.00940	0.02153	0.02790	0.03373	0.05087	0.03003	0.03796
98	$\Theta_1$	0.01327	0.01793	0.02637	0.03560	0.04787	0.02877	0.03495
99	$\Theta_1$	0.00447	0.01080	0.01417	0.01860	0.04120	0.01803	0.02073
100	$\Theta_1$	0.01573	0.02687	0.03623	0.04220	0.05007	0.03343	0.04690
All	$\Theta_1$	0.01753	0.01947	0.02063	0.02173	0.02360	0.02070	0.02064

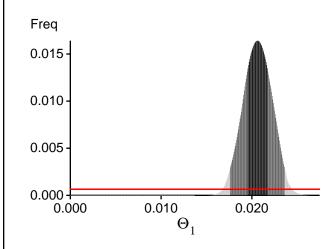
#### Citation suggestions:

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	-13940.68	-13780.14	-13826.83	-13913.07
2	-14735.67	-14403.52	-14439.79	-14511.54
3	-13886.82	-13733.34	-13773.79	-13867.58
4	-14280.61	-14093.80	-14155.85	-14224.42
5	-14439.24	-14186.19	-14235.20	-14306.36
6	-13920.23	-13766.94	-13816.31	-13901.53
7	-13923.53	-13766.75	-13813.08	-13900.57
8	-14207.98	-13970.57	-14013.99	-14091.88
9	-13881.66	-13728.56	-13770.06	-13862.84
10	-14548.31	-14238.29	-14271.14	-14348.22
11	-13994.55	-13825.02	-13878.15	-13957.21
12	-14439.79	-14171.69	-14215.84	-14290.93
13	-13968.63	-13809.27	-13859.16	-13941.68
14	-13965.45	-13801.35	-13848.94	-13934.34
15	-13935.07	-13780.59	-13830.61	-13915.66
16	-13942.45	-13780.29	-13825.24	-13912.72
17	-16261.73	-15749.01	-15762.51	-15837.65
18	-14245.91	-14015.16	-14054.46	-14137.92
19	-13957.46	-13786.95	-13833.75	-13918.46
20	-13928.05	-13771.31	-13817.63	-13903.24
21	-14646.93	-14436.05	-14485.32	-14564.80
22	-17199.89	-15695.56	-15526.09	-15593.66
23	-13949.55	-13793.19	-13843.59	-13926.44
24	-14015.22	-13832.00	-13876.82	-13960.53
25	-14046.95	-13864.88	-13916.12	-13995.34
26	-15331.07	-14774.38	-14774.67	-14843.33
27	-13899.36	-13745.38	-13787.70	-13879.39
28	-17206.10	-15650.89	-15458.64	-15540.18
29	-13998.37	-13818.58	-13865.14	-13948.59

Migrate 5.0.0a: (http://popgen.sc.fsu.edu) [program run on 02:14:37]

30	-13987.26	-13825.56	-13875.46	-13959.53
31	-14002.03	-13837.91	-13891.34	-13970.49
32	-13910.04	-13756.84	-13802.99	-13891.75
33	-14018.49	-13859.92	-13917.66	-13996.18
34	-13945.36	-13789.95	-13835.93	-13924.55
35	-14269.06	-14083.65	-14135.55	-14216.05
36	-14174.98	-13943.00	-13983.17	-14064.74
37	-13885.45	-13732.19	-13772.42	-13867.39
38	-14267.74	-14033.65	-14081.66	-14155.78
39	-13886.06	-13732.88	-13773.86	-13867.32
40	-13898.53	-13745.15	-13788.13	-13879.51
41	-13976.21	-13802.45	-13847.96	-13933.47
42	-26661.04	-22827.19	-22268.87	-22364.25
43	-14354.94	-14148.60	-14203.11	-14276.60
44	-13935.03	-13776.71	-13822.65	-13910.75
45	-13992.24	-13815.10	-13863.00	-13945.68
46	-13915.31	-13758.79	-13802.86	-13891.53
47	-13899.71	-13746.09	-13789.32	-13880.45
48	-14104.09	-13911.85	-13964.34	-14040.63
49	-13900.77	-13745.11	-13786.42	-13878.39
50	-14041.19	-13859.94	-13911.31	-13990.32
51	-13950.24	-13794.28	-13840.85	-13928.37
52	-13948.24	-13793.63	-13843.80	-13927.51
53	-14180.78	-13972.38	-14026.87	-14098.84
54	-14739.46	-14481.38	-14535.36	-14603.91
55	-13922.12	-13768.66	-13816.84	-13903.46
56	-14021.24	-13832.98	-13874.95	-13961.77
57	-16609.61	-16241.75	-16286.38	-16353.02
58	-14002.48	-13826.30	-13875.83	-13957.49
59	-13965.20	-13800.36	-13848.12	-13932.09
60	-13898.40	-13745.15	-13787.22	-13879.42
61	-13902.46	-13746.15	-13787.30	-13879.96
62	-14001.24	-13836.04	-13889.03	-13968.37
63	-13947.06	-13788.58	-13838.03	-13922.49
64	-13930.25	-13767.45	-13812.64	-13901.77
65	-27275.72	-20809.76	-19735.43	-19813.09
66	-14129.37	-13953.71	-14007.55	-14084.54
67	-15062.62	-14575.67	-14589.32	-14656.46
68	-14203.26	-13968.87	-14016.03	-14091.67
69	-13906.48	-13753.16	-13796.63	-13887.59
70	-13978.04	-13802.87	-13846.48	-13933.84
71	-13974.25	-13802.77	-13849.41	-13934.40
72	-14967.07	-14554.01	-14570.19	-14651.53
73	-14392.07	-14073.13	-14102.64	-14179.67
74	-14114.56	-13929.34	-13989.56	-14059.75
L				

All	-1459647.32	-1423405.18	-1425206.34	-1433388.69
100	-16257.61	-15478.39	-15439.64	-15516.55
99	-13926.65	-13767.39	-13813.00	-13902.44
98	-14340.81	-14045.63	-14080.00	-14155.51
97	-15433.54	-14691.11	-14647.76	-14723.07
96	-17962.28	-16263.68	-16065.04	-16130.15
95	-14129.63	-13914.85	-13958.45	-14040.03
94	-13997.07	-13824.50	-13876.70	-13957.07
93	-14007.23	-13847.13	-13902.75	-13980.42
92	-14947.88	-14649.97	-14693.49	-14767.01
91	-13968.53	-13799.35	-13847.72	-13931.10
90	-13988.73	-13831.78	-13884.68	-13967.38
89	-13977.39	-13813.98	-13864.40	-13946.46
88	-13999.55	-13822.49	-13870.69	-13953.05
87	-13917.88	-13764.19	-13808.08	-13898.26
86	-14084.14	-13908.81	-13966.72	-14039.90
85	-13950.21	-13791.29	-13839.42	-13926.54
84	-14056.90	-13894.14	-13951.37	-14027.36
83	-13915.83	-13756.90	-13799.73	-13890.07
82	-20002.35	-18564.23	-18433.24	-18507.26
81	-13925.68	-13770.79	-13815.70	-13904.90
80	-13979.40	-13813.42	-13864.86	-13946.63
79	-13951.35	-13790.63	-13840.39	-13924.04
78	-13897.53	-13741.31	-13781.65	-13874.67
77	-13952.62	-13786.84	-13832.33	-13919.16
76	-14316.27	-14108.68	-14167.27	-14237.15
75	-14094.40	-13922.30	-13979.50	-14055.14

- (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 = 79.133306]

#### 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
$\Theta_1$	357923238/400018036	0.89477
Genealogies	505330883/1599981964	0.31584

## MCMC-Autocorrelation and Effective MCMC Sample Size

Parameter	Autocorrelation	Effective Sampe Size
$\Theta_1$ Genealogies	0.58204 0.08076	2691201.16 8658752.21

## 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