# **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 22:25:36 2017

Program finished at Sun Aug 13 00:26:56 2017 [Runtime:0000:02:01:20]



### **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) 2597941120

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 chains 1
Recorded 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.0.4

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:

Print genealogies [only some for some data type]:

# Data summary

Data file: infile.0.4
Datatype: Sequence data
Number of loci: 100

Mutationmod	$\sim$ 1.	
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Mutationmodel:				
Locus Si	ublocus	Mutationmodel	Mutationmodel parameters	
1	1	Jukes-Cantor	[Basefreq: =0.25]	
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# Bayesian Analysis: Posterior distribution table

Locus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
1	$\Theta_1$	0.00453	0.00453	0.01050	0.02320	0.02320	0.01390	0.01600
2	$\Theta_1$	0.00980	0.01647	0.02150	0.02827	0.04647	0.02597	0.03189
3	$\Theta_1$	0.00307	0.00880	0.01063	0.01307	0.03147	0.01397	0.01612
4	$\Theta_1$	0.00467	0.00827	0.01097	0.01380	0.02260	0.01390	0.01603
5	$\Theta_1$	0.01507	0.02407	0.03037	0.03940	0.04960	0.03223	0.04294
6	$\Theta_1$	0.00293	0.00700	0.01070	0.01607	0.03233	0.01397	0.01606
7	$\Theta_1$	0.00393	0.00553	0.01357	0.03187	0.04220	0.01743	0.02001
8	$\Theta_1$	0.00367	0.00707	0.01077	0.01613	0.02773	0.01397	0.01602
9	$\Theta_1$	0.00347	0.00707	0.01063	0.01613	0.02847	0.01397	0.01606
10	$\Theta_1$	0.00260	0.00707	0.01070	0.01613	0.03500	0.01397	0.01613
11	$\Theta_1$	0.00353	0.00847	0.01077	0.01353	0.02840	0.01403	0.01613
12	$\Theta_1$	0.00433	0.00607	0.01050	0.01827	0.02420	0.01390	0.01608
13	$\Theta_1$	0.00593	0.01347	0.01617	0.01933	0.04153	0.02017	0.02335
14	$\Theta_1$	0.00720	0.00853	0.01523	0.02507	0.02933	0.01850	0.02126
15	$\Theta_1$	0.00240	0.00833	0.01250	0.01733	0.04667	0.01577	0.01830
16	$\Theta_1$	0.00360	0.00467	0.01210	0.02907	0.03573	0.01570	0.01809
17	$\Theta_1$	0.00273	0.00860	0.01070	0.01327	0.03400	0.01397	0.01611
18	$\Theta_1$	0.01253	0.02173	0.03103	0.03807	0.04960	0.03077	0.04185

19	$\Theta_1$	0.00387	0.00773	0.01110	0.01467	0.02640	0.01397	0.01605
20	$\Theta_1$	0.00287	0.00760	0.01077	0.01520	0.03300	0.01397	0.01609
21	$\Theta_1$	0.00227	0.00793	0.01057	0.01433	0.03733	0.01397	0.01598
22	$\Theta_1$	0.01353	0.02160	0.02883	0.03693	0.04927	0.03063	0.04126
23	$\Theta_1$	0.00347	0.00833	0.01190	0.01740	0.03673	0.01577	0.01828
24	$\Theta_1$	0.00367	0.01007	0.01077	0.01140	0.02773	0.01390	0.01609
25	$\Theta_1$	0.00413	0.00900	0.01357	0.02033	0.04007	0.01717	0.01968
26	$\Theta_1$	0.00433	0.01060	0.01363	0.01720	0.03893	0.01743	0.02017
27	$\Theta_1$	0.00660	0.00827	0.01557	0.02893	0.03533	0.01957	0.02256
28	$\Theta_1$	0.00287	0.00613	0.01090	0.01813	0.03273	0.01397	0.01609
29	$\Theta_1$	0.00467	0.01200	0.01423	0.01600	0.03773	0.01783	0.02088
30	$\Theta_1$	0.00513	0.00833	0.01077	0.01380	0.02107	0.01397	0.01613
31	$\Theta_1$	0.00440	0.00547	0.01083	0.02007	0.02393	0.01397	0.01608
32	$\Theta_1$	0.00427	0.00960	0.01183	0.01480	0.03100	0.01557	0.01796
33	$\Theta_1$	0.00300	0.00560	0.01083	0.01960	0.03207	0.01397	0.01613
34	$\Theta_1$	0.00193	0.00800	0.01083	0.01420	0.04153	0.01397	0.01615
35	$\Theta_1$	0.01187	0.01887	0.02650	0.03420	0.04907	0.02897	0.03699
36	$\Theta_1$	0.00293	0.00807	0.01070	0.01400	0.03220	0.01390	0.01602
37	$\Theta_1$	0.00173	0.00733	0.01083	0.01553	0.04293	0.01390	0.01602
38	$\Theta_1$	0.00413	0.01120	0.01203	0.01287	0.03200	0.01557	0.01785
39	$\Theta_1$	0.00880	0.01640	0.02337	0.03407	0.04987	0.02763	0.03532
40	$\Theta_1$	0.00240	0.00707	0.01083	0.01607	0.03700	0.01397	0.01607
41	$\Theta_1$	0.00580	0.01053	0.01437	0.01980	0.03353	0.01810	0.02070

ocus.	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
42	$\Theta_1$	0.00393	0.00393	0.01063	0.02600	0.02600	0.01397	0.01605
43	$\Theta_1$	0.00293	0.00773	0.01063	0.01460	0.03193	0.01397	0.01602
44	$\Theta_1$	0.00267	0.00927	0.01057	0.01220	0.03427	0.01397	0.01605
45	$\Theta_1$	0.00260	0.00907	0.01090	0.01287	0.03493	0.01397	0.01605
46	$\Theta_1$	0.00347	0.00927	0.01077	0.01240	0.02907	0.01397	0.01607
47	$\Theta_1$	0.00240	0.00987	0.01077	0.01153	0.03600	0.01397	0.01604
48	$\Theta_1$	0.00500	0.00813	0.01257	0.01813	0.02747	0.01570	0.01814
49	$\Theta_1$	0.00373	0.00700	0.01083	0.01607	0.02733	0.01397	0.01611
50	$\Theta_1$	0.01373	0.02427	0.03123	0.04120	0.04980	0.03190	0.04525
51	$\Theta_1$	0.00253	0.00533	0.01057	0.02027	0.03507	0.01390	0.01598
52	$\Theta_1$	0.00547	0.01040	0.01917	0.03553	0.05040	0.02397	0.02996
53	$\Theta_1$	0.00227	0.00593	0.01083	0.01867	0.03740	0.01390	0.01606
54	$\Theta_1$	0.00213	0.00873	0.01063	0.01280	0.03840	0.01390	0.01603
55	$\Theta_1$	0.00487	0.01033	0.01337	0.01660	0.03353	0.01690	0.01941
56	$\Theta_1$	0.00233	0.00833	0.01063	0.01353	0.03687	0.01390	0.01600
57	$\Theta_1$	0.00473	0.00787	0.01357	0.02427	0.03700	0.01757	0.02007
58	$\Theta_1$	0.00467	0.00767	0.01090	0.01493	0.02293	0.01397	0.01601
59	$\Theta_1$	0.00340	0.00673	0.01070	0.01667	0.02873	0.01397	0.01599
60	$\Theta_1$	0.00300	0.00427	0.01070	0.02473	0.03173	0.01397	0.01601
61	$\Theta_1$	0.01493	0.02360	0.02930	0.03867	0.04947	0.03170	0.04208

62	$\Theta_1$	0.00380	0.00847	0.01297	0.01947	0.03860	0.01637	0.01873
63	$\Theta_1$	0.00460	0.00760	0.01243	0.01873	0.02867	0.01550	0.01782
64	$\Theta_1$	0.00567	0.01153	0.01503	0.01933	0.03727	0.01877	0.02141
65	$\Theta_1$	0.00140	0.00593	0.01070	0.01873	0.04667	0.01397	0.01609
66	$\Theta_1$	0.00293	0.00653	0.01197	0.02180	0.04093	0.01563	0.01798
67	$\Theta_1$	0.00640	0.00967	0.01497	0.02293	0.03373	0.01890	0.02188
68	$\Theta_1$	0.00573	0.01240	0.01437	0.01687	0.03447	0.01810	0.02068
69	$\Theta_1$	0.00287	0.00787	0.01063	0.01447	0.03273	0.01390	0.01608
70	$\Theta_1$	0.01340	0.02207	0.02610	0.03660	0.04947	0.03083	0.04048
71	$\Theta_1$	0.00227	0.00673	0.01063	0.01653	0.03740	0.01390	0.01602
72	$\Theta_1$	0.00327	0.00860	0.01063	0.01327	0.02987	0.01397	0.01608
73	$\Theta_1$	0.00520	0.01173	0.01490	0.01913	0.04013	0.01883	0.02168
74	$\Theta_1$	0.00413	0.00853	0.01057	0.01327	0.02573	0.01397	0.01607
75	$\Theta_1$	0.00307	0.00840	0.01057	0.01340	0.03147	0.01390	0.01599
76	$\Theta_1$	0.00233	0.00840	0.01083	0.01353	0.03673	0.01397	0.01614
77	$\Theta_1$	0.00773	0.01193	0.01817	0.02740	0.04100	0.02217	0.02566
78	$\Theta_1$	0.00407	0.00847	0.01057	0.01333	0.02553	0.01397	0.01611
79	$\Theta_1$	0.00407	0.00933	0.01070	0.01233	0.02540	0.01397	0.01604
80	$\Theta_1$	0.00893	0.01320	0.02217	0.03820	0.04940	0.02697	0.03436
81	$\Theta_1$	0.00460	0.01207	0.01390	0.01573	0.03733	0.01737	0.01999
82	$\Theta_1$	0.00487	0.00713	0.01283	0.02300	0.03200	0.01650	0.01888
83	$\Theta_1$	0.00367	0.00840	0.01090	0.01387	0.02720	0.01397	0.01607
84	$\Theta_1$	0.00320	0.00853	0.01057	0.01307	0.03013	0.01390	0.01600

Locus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
85	$\Theta_1$	0.00327	0.00940	0.01210	0.01533	0.03753	0.01543	0.01778
86	$\Theta_1$	0.00453	0.00767	0.01230	0.01913	0.03020	0.01577	0.01814
87	$\Theta_1$	0.00293	0.00753	0.01083	0.01513	0.03180	0.01397	0.01603
88	$\Theta_1$	0.02013	0.03587	0.04450	0.04840	0.05080	0.03770	0.05550
89	$\Theta_1$	0.00340	0.00767	0.01063	0.01467	0.02907	0.01390	0.01612
90	$\Theta_1$	0.00673	0.01367	0.01750	0.02093	0.04027	0.02110	0.02464
91	$\Theta_1$	0.00253	0.00693	0.01090	0.01600	0.03533	0.01390	0.01604
92	$\Theta_1$	0.00507	0.00753	0.01063	0.01493	0.02107	0.01390	0.01597
93	$\Theta_1$	0.00333	0.00653	0.01030	0.01707	0.02960	0.01397	0.01610
94	$\Theta_1$	0.00287	0.01060	0.01217	0.01380	0.04193	0.01570	0.01812
95	$\Theta_1$	0.00447	0.01087	0.01390	0.01760	0.03907	0.01783	0.02071
96	$\Theta_1$	0.00207	0.00860	0.01077	0.01340	0.03967	0.01397	0.01600
97	$\Theta_1$	0.00387	0.00960	0.01070	0.01180	0.02667	0.01390	0.01601
98	$\Theta_1$	0.00293	0.00773	0.01083	0.01447	0.03193	0.01390	0.01604
99	$\Theta_1$	0.00433	0.00713	0.01443	0.02727	0.04193	0.01797	0.02063
100	$\Theta_1$	0.00233	0.00940	0.01070	0.01200	0.03687	0.01397	0.01606
All	$\Theta_1$	0.00973	0.01140	0.01250	0.01353	0.01520	0.01257	0.01250

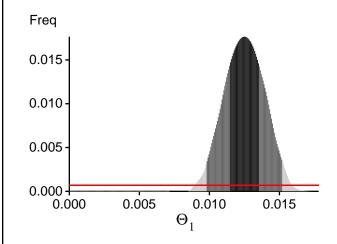
Citation suggestions:

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	-13885.37	-13732.02	-13772.67	-13866.46
2	-14486.48	-14165.90	-14192.74	-14274.35
3	-13886.29	-13733.15	-13774.20	-13868.79
4	-13886.77	-13733.38	-13774.82	-13867.97
5	-14041.57	-13876.38	-13932.64	-14009.31
6	-13885.86	-13732.73	-13773.35	-13866.94
7	-13918.20	-13760.25	-13803.59	-13892.30
8	-13882.47	-13729.06	-13770.05	-13863.50
9	-13884.62	-13731.33	-13771.71	-13865.62
10	-13885.56	-13732.20	-13773.35	-13866.36
11	-13885.26	-13731.99	-13772.95	-13866.46
12	-13886.38	-13733.40	-13774.47	-13867.70
13	-14042.83	-13879.43	-13928.88	-14013.50
14	-14015.59	-13821.98	-13863.10	-13949.59
15	-13899.76	-13746.14	-13788.82	-13880.66
16	-13897.75	-13744.55	-13788.40	-13878.77
17	-13886.35	-13733.17	-13772.01	-13868.07
18	-14765.14	-14497.33	-14542.83	-14618.90
19	-13884.76	-13731.67	-13772.79	-13866.81
20	-13886.32	-13732.99	-13774.20	-13867.44
21	-13886.65	-13733.15	-13773.79	-13868.22
22	-16472.58	-15739.55	-15709.23	-15788.41
23	-13896.85	-13743.70	-13786.51	-13878.37
24	-13886.17	-13732.75	-13774.08	-13867.43
25	-13955.14	-13782.62	-13825.56	-13913.86
26	-13916.24	-13759.77	-13803.83	-13892.30
27	-14071.59	-13869.16	-13910.13	-13995.94
28	-13886.52	-13733.10	-13773.92	-13870.93
29	-13910.01	-13756.41	-13801.83	-13890.79

Migrate 5.0.0a: (http://popgen.sc.fsu.edu) [program run on 22:25:36]

30	-13886.00	-13732.96	-13774.17	-13868.01
31	-13886.59	-13733.14	-13774.52	-13867.62
32	-13902.68	-13747.14	-13788.57	-13880.51
33	-13886.38	-13732.93	-13774.30	-13867.25
34	-13886.50	-13732.96	-13773.63	-13867.18
35	-13989.09	-13834.82	-13890.79	-13971.32
36	-13886.36	-13733.39	-13774.03	-13867.79
37	-13886.27	-13733.12	-13774.30	-13867.38
38	-13900.66	-13744.99	-13786.77	-13878.07
39	-17327.65	-16117.31	-15999.29	-16077.02
40	-13882.43	-13729.46	-13770.42	-13863.96
41	-13995.98	-13810.20	-13851.75	-13939.19
42	-13886.01	-13732.78	-13773.97	-13867.11
43	-13886.31	-13732.84	-13773.68	-13867.08
44	-13884.36	-13731.11	-13772.09	-13866.72
45	-13885.69	-13732.31	-13773.55	-13866.91
46	-13886.35	-13733.00	-13774.13	-13868.78
47	-13883.64	-13730.49	-13771.35	-13865.23
48	-13900.73	-13746.37	-13787.93	-13880.69
49	-13883.50	-13730.01	-13771.15	-13865.21
50	-26023.69	-23777.89	-23555.96	-23628.04
51	-13886.69	-13733.54	-13774.91	-13867.76
52	-27159.88	-21598.83	-20707.15	-20784.67
53	-13883.50	-13730.28	-13771.51	-13865.86
54	-13885.04	-13731.67	-13773.17	-13866.15
55	-13928.06	-13773.91	-13817.66	-13908.74
56	-13885.89	-13732.78	-13773.96	-13867.39
57	-13956.42	-13792.09	-13836.78	-13925.22
58	-13886.39	-13732.95	-13773.68	-13867.57
59	-13884.92	-13731.52	-13772.31	-13865.92
60	-13886.24	-13732.84	-13774.37	-13867.69
61	-15744.99	-15046.70	-15018.46	-15091.59
62	-13929.68	-13764.86	-13807.59	-13896.90
63	-13902.94	-13746.41	-13788.05	-13879.17
64	-14037.17	-13834.79	-13874.87	-13960.89
65	-13884.58	-13730.98	-13771.92	-13865.47
66	-13901.31	-13745.55	-13787.58	-13880.19
67	-13934.99	-13775.50	-13820.90	-13908.08
68	-13994.90	-13809.33	-13851.35	-13939.18
69	-13886.51	-13733.18	-13772.16	-13867.30
70	-14157.11	-13974.76	-14028.06	-14105.85
71	-13885.75	-13732.25	-13773.51	-13866.89
72	-13886.05	-13732.75	-13773.89	-13867.09
73	-13974.68	-13816.80	-13864.46	-13950.68
74	-13885.74	-13732.38	-13773.64	-13866.88

75	-13885.65	-13732.20	-13773.27	-13867.08
76	-13885.17	-13731.96	-13772.53	-13866.17
77	-14113.16	-13915.47	-13959.83	-14043.24
78	-13883.83	-13730.79	-13771.97	-13865.08
79	-13884.92	-13731.71	-13772.39	-13866.52
80	-18934.95	-17003.88	-16756.99	-16834.30
81	-13915.90	-13757.45	-13801.34	-13891.19
82	-13923.32	-13764.70	-13808.00	-13899.50
83	-13884.62	-13731.37	-13773.17	-13869.02
84	-13885.58	-13732.40	-13772.53	-13867.23
85	-13900.83	-13744.52	-13785.49	-13877.23
86	-13898.42	-13744.20	-13786.28	-13878.26
87	-13882.94	-13729.78	-13771.11	-13867.45
88	-14237.65	-14064.36	-14128.15	-14198.78
89	-13886.82	-13733.57	-13774.81	-13867.80
90	-13935.60	-13777.70	-13825.13	-13911.41
91	-13887.19	-13733.55	-13774.39	-13868.05
92	-13886.62	-13733.19	-13774.68	-13867.66
93	-13884.46	-13731.10	-13772.51	-13865.55
94	-13896.16	-13741.99	-13784.12	-13876.37
95	-13912.84	-13758.38	-13803.56	-13893.44
96	-13885.74	-13732.30	-13773.60	-13867.23
97	-13884.36	-13731.27	-13771.99	-13865.79
98	-13885.75	-13732.33	-13772.90	-13866.73
99	-13929.39	-13768.80	-13814.56	-13907.23
100	-13885.55	-13732.21	-13773.79	-13866.48
All	-1431004.12	-1403511.70	-1405959.20	-1414990.05

- (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 = 134.678604]

#### 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
$\Theta_1$	336498903/400008005	0.84123
Genealogies	985586980/1599991995	0.61599

# MCMC-Autocorrelation and Effective MCMC Sample Size

Parameter	Autocorrelation	Effective Sampe Size
$\Theta_1$	0.44799	3871679.23
Genealogies	0.06851	8743887.01

# 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

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