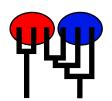
Preliminary migrate analysis of M. californianus

MIGRATION RATE AND POPULATION SIZE ESTIMATION using the coalescent and maximum likelihood or Bayesian inference Migrate-n version 3.7.2 [April-12-18]

Program started at Tue Jun 1 13:45:41 2021 Program finished at Tue Jun 1 19:03:01 2021



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) 4095239889

Start parameters:

Theta values were generated from guessed values

Theta = 0.01000

M values were generated from guessed values

M-matrix:

100000.00 [all are the same]

Connection type matrix:

where m = average (average over a group of Thetas or M,

s = symmetric M, S = symmetric 4Nm, 0 = zero, and not estimated,

* = free to vary, Thetas are on diagonal

Population	1	2	3	4	5	6	7	8	9	10	11	12
1 ElfinCo	m	m	m	m	m	m	m	m	m	m	m	m
2 Bamfiel	m	m	m	m	m	m	m	m	m	m	m	m
3 PortRen	m	m	m	m	m	m	m	m	m	m	m	m
4 WalkOnB	m	m	m	m	m	m	m	m	m	m	m	m
5 BodegaH	m	m	m	m	m	m	m	m	m	m	m	m
6 Davenpo	m	m	m	m	m	m	m	m	m	m	m	m
7 VistaDe	m	m	m	m	m	m	m	m	m	m	m	m
8 HazardR	m	m	m	m	m	m	m	m	m	m	m	m
9 Refugio	m	m	m	m	m	m	m	m	m	m	m	m
10 Carpint	m	m	m	m	m	m	m	m	m	m	m	m

11 WhitePo	* *	* * *	* * * * * *
	* *	* * *	* * * * * *
12 LaJolla	^ *	^	
Order of param		0	
1	$\Theta_1 =$	Θ_1 [m]	
2	Θ_2 =	Θ_1 [m]	
3	$\Theta_3^2 =$	Θ_1 [m]	
4	$\Theta_4 =$	Θ_1 [m]	
5	$\Theta_5^{T} =$	Θ_1 [m]	
6	Θ_6 =	Θ_1 [m]	
7	$\Theta_7 =$	Θ_1 [m]	
8	$\Theta_8 =$	Θ_1 [m]	
9	$\Theta_{0} =$	Θ_1 [m]	
10	Θ_{10} =	Θ_1 [m]	
11	Θ_{11}		<displayed></displayed>
12	Θ_{12}		<displayed></displayed>
13	$M_{2->1}^{12} =$	$M_{2->1}$ [m]	<displayed></displayed>
14	$M_{3->1} =$	$M_{2->1}$ [m]	
15	$M_{4->1}^{3} =$	$M_{2->1}^{2}$ [m]	
16	$M_{5->1}^{7-1} =$	M $_{2->1}^{2}$ [m]	
17	$M_{6->1}^{5->1} =$	$M_{2->1}^{2->1}$ [m]	
18	$M_{7->1} =$	$M_{2->1}$ [m]	
19	$M_{8->1}^{7->1} =$	$M_{2->1}$ [m]	
20	$M_{9->1}^{6->1} =$	M $_{2->1}^{2->1}$ [m]	
21	M	$M_{2->1}$ [m]	
22	10->1	M $_{2->1}^{2->1}$ [m]	
23	M	M $_{2->1}^{2->1}$ [m]	
24	12->1	$M_{2->1}$ [m]	
25	$M_{1->2} = M_{3->2} = M_{1->2}$	$M_{2->1}$ [m]	
26	$M_{4->2} = M_{4->2}$	$M_{2->1}$ [m]	
27	$M_{5->2} = M_{5->2}$	$M_{2->1}$ [m] $M_{2->1}$ [m]	
28	$M_{5->2} = M_{6->2} = M_{10}$	$M_{2->1}$ [m] $M_{2->1}$ [m]	
29	$M_{6\rightarrow 2} = M_{7\rightarrow 2} =$	M [m]	
30	1-22	$ \begin{array}{ccc} M & [m] \\ M & [m] \end{array} $	
31	M _{8->2} =	$M = \begin{bmatrix} M \\ 2->1 \end{bmatrix} $ [m]	
32	M _{9->2} =	$M_{2\rightarrow 1} [m]$	
33	$M_{10->2} = M_{10->2}$	M = [m] $M = [m]$	
34	$M_{11->2} = M_{11->2}$	$M = \begin{bmatrix} M \\ 2->1 \end{bmatrix} $ [m]	
35	$M_{12->2} =$	M = [m]	
	$M_{1->3} = M_{1->3}$	$M_{2\rightarrow 1} [m]$	
36	$M_{2->3} =$	$M_{2\rightarrow 1} [m]$	
37	$M_{4->3} =$	$M_{2->1}$ [m]	
38	$M_{5->3} =$	$M_{2->1}$ [m]	
39	IVI _{6->3} =	$M_{2->1}$ [m]	
40	$M_{7->3} =$	$M_{2->1}$ [m]	

41	M _{8->3} =	M _{2->1} [m]
42	$M_{9->3}^{8->3} =$	$ \begin{array}{cccccccccccccccccccccccccccccccccccc$
43	$M_{10->3} =$	$ \begin{array}{ccc} M & 2 & 2 & 2 \\ 2 & 2 & 2 & 2 \end{array} $ [m]
44	$M_{11->3}^{10->3} =$	$ \begin{array}{ccc} M & 2 & & \\ 2 & & & \\ 2 & & & \\ \end{array} $
45	$M_{12->3}^{11->3} =$	$ \begin{array}{ccc} M & 2 & 51 & 53 \\ 2 & 2 & 51 & 53 \\ \end{array} $
46	$M_{1->4}^{12->3} =$	$ \begin{array}{ccc} M & 2 \rightarrow 1 & [m] \\ 2 \rightarrow 1 & [m] \end{array} $
47	$M_{2->4}^{1->4} =$	$ \begin{array}{ccc} M & 2->1 & [m] \\ 2->1 & [m] \end{array} $
48	$M_{3->4}^{2->4} =$	$ \begin{array}{ccc} M & 2 & & & \\ 2 & & & & & \\ 2 & & & & & & \\ \end{array} $
49	$M_{5->4}^{5->4} =$	$ \begin{array}{ccc} M & 2 & 2 & 2 \\ 2 & 2 & 2 & 2 \end{array} $ [m]
50	$M_{6->4} =$	$ \begin{array}{ccc} M & 2 & > 1 \\ 2 & > > 1 \end{array} $ [m]
51	$M_{7->4}^{0->4} =$	$M_{2->1}^{2->1}$ [m]
52	$M_{8->4} =$	$ \begin{array}{ccc} M & 2 & > 1 \\ 2 & > > 1 \end{array} $ [m]
53	$M_{9->4}^{6->4} =$	$M_{2\to 1}$ [m]
54	$M_{10->4} =$	$ \begin{array}{ccc} M & 2 & > 1 \\ 2 & > > 1 \end{array} $ [m]
55	$M_{11->4}^{10->4} =$	$M_{2->1}^{2->1}$ [m]
56	$M_{12->4} =$	$ \begin{array}{ccc} M & 2 & > 1 \\ 2 & > > 1 \end{array} $ [m]
57	$M_{1->5} =$	M _{2->1} [m]
58	M _{2->5} =	$M_{2->1}^{2}$ [m]
59	$M_{3->5}^{2>3} =$	$M_{2->1}$ [m]
60	$M_{4->5}^{5} =$	$M_{2\to 1}$ [m]
61	$M_{6->5} =$	$M_{2\rightarrow 1}$ [m]
62	M _{7->5} =	$M_{2\rightarrow 1}$ [m]
63	M _{8->5} =	$M_{2\rightarrow 1}$ [m]
64	M _{9->5} =	$M_{2\rightarrow 1}$ [m]
65	$M_{10->5} =$	$M_{2\rightarrow 1}$ [m]
66	$M_{11->5} =$	$M_{2\rightarrow 1}$ [m]
67	$M_{12->5} =$	$M_{2\rightarrow 1}$ [m]
68	M _{1->6} =	$M_{2\rightarrow 1}$ [m]
69	M _{2->6} =	$M_{2\rightarrow 1}$ [m] $M_{2\rightarrow 1}$ [m]
70	$M_{3->6}^{2->0} =$	$M_{2\rightarrow 1}$ [m]
71	IVI _{4->6} =	M _{2->1} [m]
72	IVI _{5->6} =	$M_{2\rightarrow 1}$ [m]
73	IVI _{7->6} =	$M_{2\rightarrow 1}$ [m]
74	$ V _{8->6} =$	M _{2->1} [m]
75 70	$M_{9->6} =$	M _{2->1} [m]
76 77	$M_{10->6} =$	M _{2->1} [m]
77	$M_{11->6} =$	$ \begin{array}{c} M \\ 2->1 \end{array} [m] $
78	$M_{12->6} =$	M _{2->1} [m]
79	$M_{1->7} =$	M _{2->1} [m]
80	$M_{2->7} =$	M 2->1 [m]
81	$M_{3->7} =$	$ \begin{array}{c} M \\ 2->1 \end{array} [m] $
82	$M_{4->7}^{5-7} =$	M _{2->1} [m]
83	M _{5->7} =	$ \begin{array}{c} M \\ 2->1 \end{array} [m] $
84	M _{6->7} =	$ \begin{array}{c} M \\ 2->1 \end{array} [m] $
85	M _{8->7} =	$M_{2\rightarrow 1}$ [m]

	Tremminary migrate analysis of M. Camornands COT haplotypes for Evolution 2
86	$M_{9->7} = M_{2->1} [m]$
87	$M_{10-7} = M_{2-1} [m]$
88	$M_{11->7} = M_{2->1} [m]$
89	$M_{12->7} = M_{2->1} [m]$
90	$M_{1->8} = M_{2->1} [m]$
91	$M_{2->8} = M_{2->1} [m]$
92	$M_{3-8} = M_{2-1} [m]$
93	$M_{4-8} = M_{2-1} [m]$
94	M = M = M = M = M = M = M = M = M = M =
95	$M_{6->8}^{3->6} = M_{2->1}^{2->1} [m]$
96	$M_{7-8}^{0-3} = M_{2-31}^{2-31} [m]$
97	$M_{9-8} = M_{2-1} [m]$
98	$M_{10-8} = M_{2-1} [m]$
99	$M_{11-8}^{10-8} = M_{2-1}^{2-1} [m]$
100	$M_{12->8}^{11->8} = M_{2->1}^{2->1} [m]$
101	$M \frac{12->8}{1->9} = M \frac{2->1}{2->1} [m]$
102	M = M = M = M = M = M = M = M = M = M =
103	$M_{3-9}^{2-99} = M_{2-1}^{2-1} [m]$
104	$M_{4\to9}^{3\to9} = M_{2\to1}^{2\to1} [m]$
105	$M_{5->9}^{4->9} = M_{2->1}^{2->1} [m]$
106	$M_{6->9}^{3->9} = M_{2->1}^{2->1} [m]$
107	$M_{7->9}^{6->9} = M_{2->1}^{2->1} [m]$
108	$M_{8-99}^{7-99} = M_{2-91}^{2-91} [m]$
109	$M_{10\to 9}^{8\to 9} = M_{2\to 1}^{2\to 1} [m]$
110	$M_{11->9} = M_{2->1} [m]$
111	$M_{12->9} = M_{2->1} [m]$
112	$M = \frac{12-99}{1->10} = M = \frac{2->1}{2->1} [m]$
113	$M_{2\rightarrow 10}^{1\rightarrow 10} = M_{2\rightarrow 1}^{2\rightarrow 1} [m]$
114	$M_{3\to 10} = M_{2\to 1} [m]$
115	M [ma]
116	$M_{4\rightarrow 10} = M_{2\rightarrow 1} [m]$ $M_{5\rightarrow 10} = M_{2\rightarrow 1} [m]$
117	$M = \frac{5-10}{6-10} = M = \frac{2-1}{2-1} [m]$
118	$M_{7\to 10} = M_{2\to 1} [m]$
119	$M_{8\to 10}^{7\to 10} = M_{2\to 1}^{2\to 1} [m]$
120	$M_{9\rightarrow 10} = M_{2\rightarrow 1} [m]$
121	$M_{11->10} = M_{2->1}$ [m]
122	$M_{12->10} = M_{2->1} [m]$
123	M
124	$M = \frac{1}{1->11}$ <alsplayed> <alsplayed> <alsplayed></alsplayed></alsplayed></alsplayed>
125	$M = \frac{2-311}{3-311}$ <displayed></displayed>
126	M diaplayed
127	M
128	$\begin{array}{c} \text{IVI} & \text{$<$}\text{displayed} \\ \text{M} & \text{6-$}\text{$>$}11 \end{array}$ "><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\displayed>"><\
129	$M_{7\rightarrow11} $ <displayed></displayed>
130	NA / 211
	IVI _{8->11} <asplayed></asplayed>

100000.00

3.00

1.50

Swapping interval is 1

1.00

		1 101111	illiary illigrate all	aryolo or ivi. camorri	ilarias oo i riapiotypi	68 101 EVOIUIION 2 3
131	M ₉₋	->11	<	displayed>		
132	N A)->11	<	displayed>		
133	N/I	2->11	<	displayed>		
134	N/I	->12	<	displayed>		
135	R A	->12 ->12	<	displayed>		
136	R A	->12 ->12	<	displayed>		
137	N/I	->12 ->12	<	displayed>		
138	R A	->12 ->12	<	displayed>		
139	R A	->12 ->12	<	displayed>		
140	R A	->12 ->12	<	displayed>		
141	N A	->12 ->12		displayed>		
142	ь л	->12 ->12		displayed>		
143	N/I	->12)->12		displayed>		
144	N/I)->12 1->12		displayed>		
	1	1->12		, ,		
Mutation	rate among loc	i:			Muta	tion rate is constant
Analysis	strategy:					Bayesian inference
Proposal	distributions fo	r naramatar				
Paramet		i parameter	Proposal			
Theta	GI	Mo	etropolis sampling			
M		IVIC	Slice sampling			
IVI			Slice sampling			
Prior dist	tribution for para	ameter				
Paramet	er Prior	Minimum	Mean*	Maximum	Delta	Bins
Theta	Exp window	0.000010	0.010000	10.000000	1.000000	500
М	Exp window	0.000100	100000.000000	1000000.000000	100000.000000	500
 Markov (chain settings:					Long chain
	of chains					1
Record	ded steps [a]					1000
	nent (record eve	ry x step [b]				100
	er of concurrent		cates) [c]			3
	l (sampled) para	` .	,			300000
	er of discard tre					1000
		: p 3: 2::8	/			
Multiple	Markov chains:					
	heating scheme)			4 chains	s with temperatures
				4000	000 00 2 00	4.50

Print options:

Data file:	//mcalifornianus_210528.mig
Output file:	outfile.txt
Posterior distribution raw histogram file:	bayesfile
Print data:	No
Print genealogies [only some for some data type]:	None

Data summary

Datatype: Sequence data
Number of loci: 1

Population	Locus	Gene copies	
1 ElfinCo	1	19	
2 Bamfiel	1	23	
3 PortRen	1	15	
4 WalkOnB	1	16	
5 BodegaH	1	7	
6 Davenpo	1	17	
7 VistaDe	1	19	
8 HazardR	1	23	
9 Refugio	1	16	
10 Carpint	1	19	
11 WhitePo	1	11	
12 LaJolla	1	8	
Total of all populations	1	193	
I and the second			

Bayesian Analysis: Posterior distribution table

Locus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
1	Θ_1	0.00001	0.00001	0.01001	0.06001	0.16001	0.07001	0.00071
1	Θ_2	0.00001	0.00001	0.01001	0.06001	0.16001	0.07001	0.00071
1	Θ_3	0.00001	0.00001	0.01001	0.06001	0.16001	0.07001	0.00071
1	Θ_4	0.00001	0.00001	0.01001	0.06001	0.16001	0.07001	0.00071
1	Θ_5	0.00001	0.00001	0.01001	0.06001	0.16001	0.07001	0.00071
1	Θ_6	0.00001	0.00001	0.01001	0.06001	0.16001	0.07001	0.00071
1	Θ_7	0.00001	0.00001	0.01001	0.06001	0.16001	0.07001	0.00071
1	$\Theta_8^{'}$	0.00001	0.00001	0.01001	0.06001	0.16001	0.07001	0.00071
1	Θ_9	0.00001	0.00001	0.01001	0.06001	0.16001	0.07001	0.00071
1	Θ_{10}	0.00001	0.00001	0.01001	0.06001	0.16001	0.07001	0.00071
1	Θ_{11}	0.00001	0.00001	0.01001	0.06001	0.16001	0.07001	0.00891
1	Θ_{12}^{11}	0.00001	0.00001	0.01001	0.06001	0.16001	0.07001	0.01072
1	M _{2->1}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{3->1}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{4->1}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{5->1}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{6->1}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{7->1}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{8->1}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{9->1}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{10->1}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{11->1}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{12->1}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{1->2}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{3->2}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{4->2}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{5->2}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{6->2}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{7->2}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{8->2}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{9->2}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{10->2}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{11->2}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{12->2}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{1->3}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{2->3}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5

Locus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
1	M _{4->3}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	$M_{5->3}$	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	$M_{6->3}$	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{7->3}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{8->3}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	$M_{9->3}$	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{10->3}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{11->3}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{12->3}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{1->4}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{2->4}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{3->4}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{5->4}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{6->4}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{7->4}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{8->4}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{9->4}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{10->4}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{11->4}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{12->4}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{1->5}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{2->5}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{3->5}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{4->5}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{6->5}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{7->5}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{8->5}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{9->5}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{10->5}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{11->5}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{12->5}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{1->6}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{2->6}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{3->6}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	3->6 M _{4->6}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{5->6}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	5->6 M _{7->6}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{8->6}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{9->6}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{10->6}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{11->6}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5

Locus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
1	M _{12->6}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{1->7}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{2->7}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{3->7}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{4->7}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{5->7}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{6->7}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{8->7}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{9->7}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{10->7}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{11->7}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{12->7}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{1->8}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{2->8}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	$M_{3->8}$	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{4->8}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{5->8}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{6->8}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{7->8}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	$M_{9->8}$	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{10->8}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{11->8}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{12->8}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{1->9}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{2->9}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	$M_{3->9}$	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{4->9}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{5->9}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	$M_{6->9}$	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{7->9}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	$M_{8->9}$	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{10->9}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{11->9}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{12->9}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{1->10}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{2->10}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{3->10}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{4->10}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{5->10}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{6->10}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{7->10}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5

Locus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
1	M _{8->10}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{9->10}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{11->10}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{12->10}	30000.0	40000.0	51000.0	58000.0	70000.0	53000.0	50263.5
1	M _{1->11}	12000.0	20000.0	39000.0	56000.0	80000.0	53000.0	66753.7
1	M _{2->11}	4000.0	40000.0	59000.0	78000.0	132000.0	65000.0	65561.3
1	M _{3->11}	34000.0	66000.0	91000.0	114000.0	136000.0	105000.0	118673.4
1	M _{4->11}	14000.0	100000.0	123000.0	154000.0	194000.0	119000.0	109576.1
1	M _{5->11}	72000.0	94000.0	111000.0	130000.0	164000.0	103000.0	85423.9
1	M _{6->11}	38000.0	144000.0	175000.0	194000.0	218000.0	139000.0	133833.2
1	M _{7->11}	52000.0	118000.0	143000.0	158000.0	190000.0	123000.0	120850.3
1	M _{8->11}	46000.0	62000.0	85000.0	108000.0	194000.0	153000.0	221476.9
1	M _{9->11}	14000.0	24000.0	45000.0	74000.0	178000.0	67000.0	82171.1
1	M _{10->11}	14000.0	24000.0	45000.0	68000.0	82000.0	177000.0	240562.9
1	M _{12->11}	78000.0	110000.0	139000.0	158000.0	202000.0	139000.0	139181.0
1	M _{1->12}	40000.0	102000.0	123000.0	142000.0	170000.0	111000.0	107008.5
1	M _{2->12}	66000.0	158000.0	203000.0	224000.0	262000.0	187000.0	177558.0
1	M _{3->12}	36000.0	58000.0	87000.0	110000.0	228000.0	101000.0	120530.0
1	M _{4->12}	10000.0	20000.0	43000.0	68000.0	98000.0	193000.0	252033.5
1	M _{5->12}	12000.0	28000.0	41000.0	52000.0	70000.0	151000.0	127302.9
1	M _{6->12}	0.0	2000.0	27000.0	54000.0	70000.0	225000.0	253111.1
1	M _{7->12}	22000.0	38000.0	57000.0	102000.0	194000.0	93000.0	101011.7
1	M _{8->12}	4000.0	14000.0	33000.0	50000.0	64000.0	125000.0	116122.6
1	M _{9->12}	0.0	4000.0	25000.0	94000.0	150000.0	97000.0	140391.7
1	M _{10->12}	0.0	0.0	15000.0	36000.0	220000.0	83000.0	89785.7
1	M _{11->12}	120000.0	136000.0	155000.0	184000.0	260000.0	153000.0	145010.2

Citation suggestions:

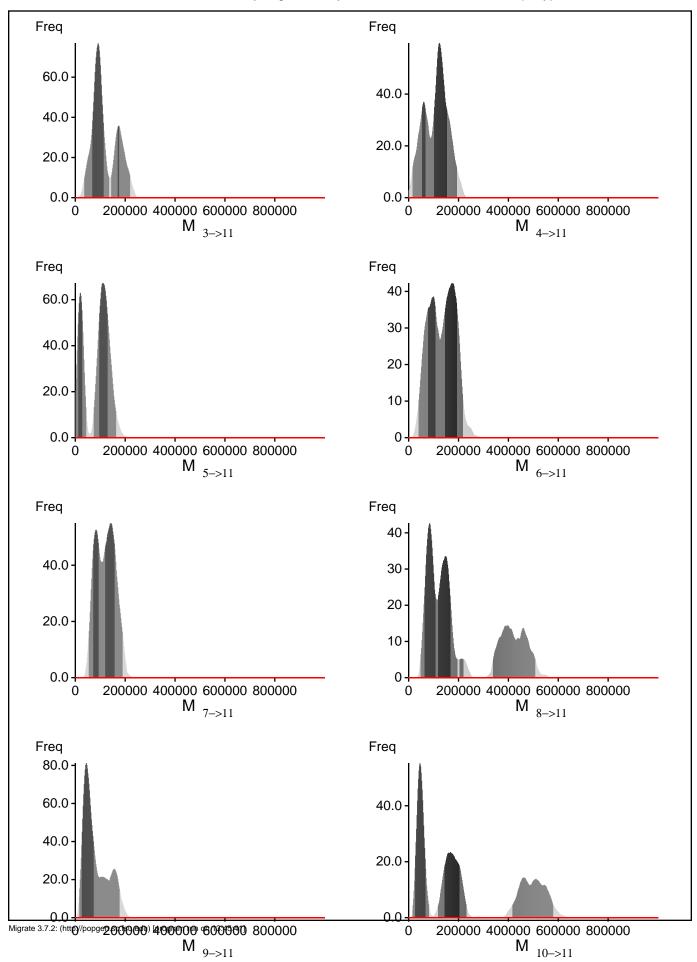
Beerli P., 2006. Comparison of Bayesian and maximum-likelihood inference of population genetic parameters. Bioinformatics 22:341-345

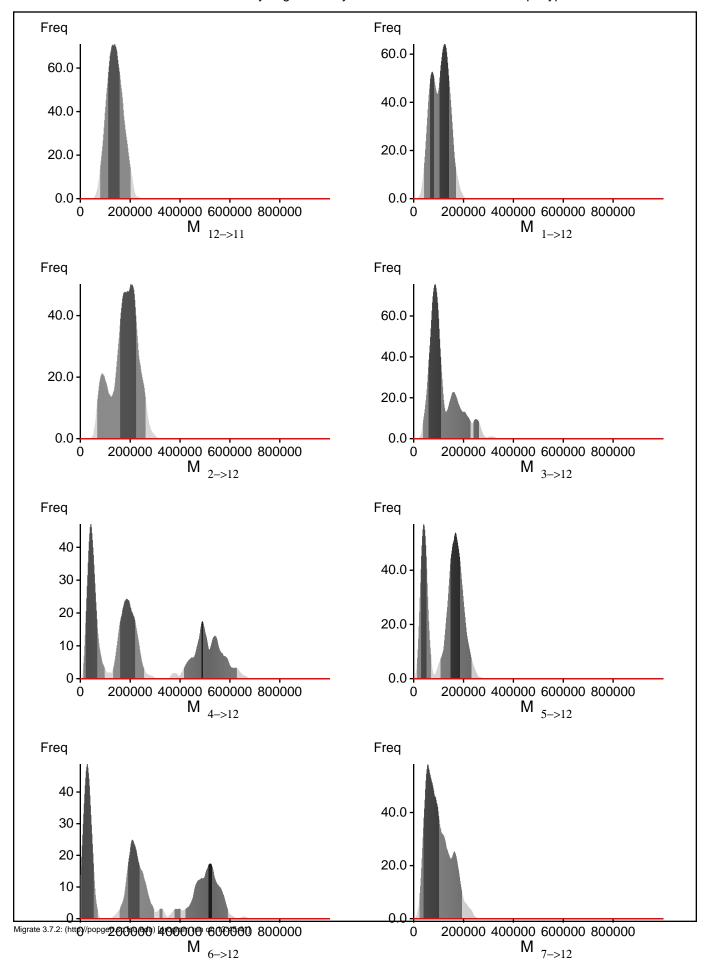
Beerli P., 2007. Estimation of the population scaled mutation rate from microsatellite data, Genetics, 177:1967-1968.

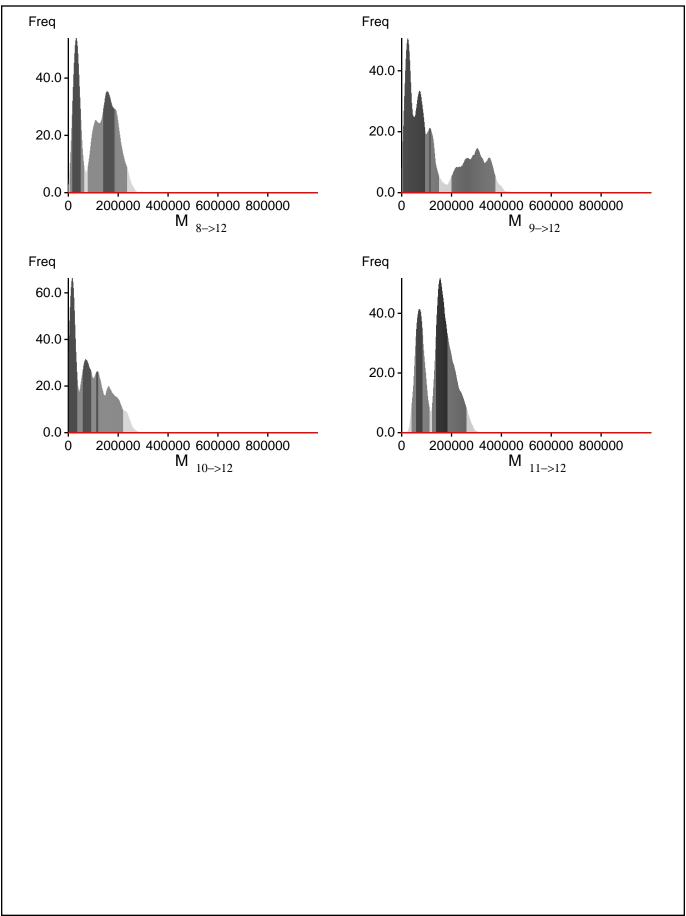
Beerli P., 2009. How to use MIGRATE or why are Markov chain Monte Carlo programs difficult to use? In Population Genetics for Animal Conservation, G. Bertorelle, M. W. Bruford, H. C. Hauffe, A. Rizzoli, and C. Vernesi, eds., vol. 17 of Conservation Biology, Cambridge University Press, Cambridge UK, pp. 42-79.

Bayesian Analysis: Posterior distribution over all loci Freq Freq 600.00 -600.00 400.00 400.00 -200.00 -200.00 0.00 -0.00 - Θ_1 8 2 8 2 6 6 Θ_{11} Freq Freq 200.0 600.00 150.0 400.00 100.0 200.00 50.0 0.00 -0.0 Θ_{12} 200000 400000 600000 800000 8 M _{2->1} Freq Freq 100.0 80.0 75.0 60.0 50.0 40.0 25.0 20.0 0.0 0.0 200000 400000 600000 800000 200000 400000 600000 800000 M _{2->11} $\mathsf{M}_{1 \to 11}$

Migrate 3.7.2: (http://popgen.sc.fsu.edu) [program run on 13:45:41]







Log-Probability of the data given the model (marginal likelihood)

Use this value for Bayes factor calculations:

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

Method	In(Prob(D Model))	Notes
Thermodynamic integration	-2378.875145	(1a)
	-2263.785700	(1b)
Harmonic mean	-1945.014335	(2)

(1a, 1b and 2) are approximations to the marginal likelihood, make sure that the program run long enough! (1a, 1b) and (2) should give similar results, in principle.

But (2) is overestimating the likelihood, it is presented for historical reasons and should not be used (1a, 1b) needs heating with chains that span a temperature range of 1.0 to at least 100,000.

(1b) is using a Bezier-curve to get better approximations for runs with low number of heated chains

Citation suggestions:

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

Acceptance ratios for all parameters and the genealogies

Parameter	Accepted changes	Ratio
Θ_1	13/1066	0.01220
Θ_2	13/1066	0.01220
Θ_3^2	13/1066	0.01220
Θ_4°	13/1066	0.01220
Θ_5^{\cdot}	13/1066	0.01220
Θ_6°	13/1066	0.01220
Θ_7°	13/1066	0.01220
$\Theta_8^{'}$	13/1066	0.01220
$\mathbf{p}_{\mathbf{q}}$	13/1066	0.01220
9 ₁₀	13/1066	0.01220
211	618/1083	0.57064
12	693/1055	0.65687
1 2->1	1040/1040	1.00000
1 3->1 3->1	1040/1040	1.00000
1 4->1	1040/1040	1.00000
1 5->1	1040/1040	1.00000
1 6->1	1040/1040	1.00000
7->1	1040/1040	1.00000
1 8->1	1040/1040	1.00000
9->1	1040/1040	1.00000
10->1	1040/1040	1.00000
Λ	1040/1040	1.00000
11->1 1	1040/1040	1.00000
1 12->1	1040/1040	1.00000
1->2 1	1040/1040	1.00000
3->2 1	1040/1040	1.00000
4->2 1	1040/1040	1.00000
J->2 1	1040/1040	1.00000
0->2 1	1040/1040	1.00000
/->2 A	1040/1040	1.00000
0->2 1	1040/1040	1.00000
9->2 1	1040/1040	1.00000
10->2 1	1040/1040	1.00000
11->2 1	1040/1040	1.00000
12->2 1	1040/1040	1.00000
1->5 1	1040/1040	1.00000
7 2->3 1 4 2	1040/1040	1.00000

M _{5->3}	1040/1040	1.00000
M _{6->3}	1040/1040	1.00000
M _{7->3}	1040/1040	1.00000
M _{8->3}	1040/1040	1.00000
$M_{9->3}$	1040/1040	1.00000
$M_{10->3}$	1040/1040	1.00000
M 11->3	1040/1040	1.00000
$M_{12->3}$	1040/1040	1.00000
M _{1->4}	1040/1040	1.00000
$M_{2\rightarrow 4}$	1040/1040	1.00000
$M_{3\rightarrow 4}$	1040/1040	1.00000
M _{5->4}	1040/1040	1.00000
M _{6->4}	1040/1040	1.00000
M _{7->4}	1040/1040	1.00000
M _{8->4}	1040/1040	1.00000
M _{9->4}	1040/1040	1.00000
M 10->4	1040/1040	1.00000
M 11->4	1040/1040	1.00000
M _{12->4}	1040/1040	1.00000
M _{1->5}	1040/1040	1.00000
M _{2->5}	1040/1040	1.00000
M _{3->5}	1040/1040	1.00000
M _{4->5}	1040/1040	1.00000
M _{6->5}	1040/1040	1.00000
M _{7->5}	1040/1040	1.00000
M _{8->5}	1040/1040	1.00000
M _{9->5}	1040/1040	1.00000
M _{10->5}	1040/1040	1.00000
M _{11->5}	1040/1040	1.00000
M _{12->5}	1040/1040	1.00000
M _{1->6}	1040/1040	1.00000
M _{2->6}	1040/1040	1.00000
M _{3->6}	1040/1040	1.00000
M _{4->6}	1040/1040	1.00000
M 5->6	1040/1040	1.00000
M 7->6	1040/1040	1.00000
M 8->6	1040/1040	1.00000
M 9->6	1040/1040	1.00000
M 10->6	1040/1040	1.00000
M 11->6	1040/1040	1.00000
M 12->6	1040/1040	1.00000
M 1->7	1040/1040	1.00000
M 2->7	1040/1040	1.00000
M 3->7	1040/1040	1.00000
M _{4->7}	1040/1040	1.00000

M _{5->7}	1040/1040	1.00000
M _{6->7}	1040/1040	1.00000
M _{8->7}	1040/1040	1.00000
M _{9->7}	1040/1040	1.00000
M 10->7	1040/1040	1.00000
M 11->7	1040/1040	1.00000
$M_{12->7}$	1040/1040	1.00000
M 1->8	1040/1040	1.00000
$M_{2->8}$	1040/1040	1.00000
$M_{3->8}$	1040/1040	1.00000
$M_{4->8}$	1040/1040	1.00000
$M_{5->8}$	1040/1040	1.00000
M _{6->8}	1040/1040	1.00000
M _{7->8}	1040/1040	1.00000
$M_{9->8}$	1040/1040	1.00000
M 10->8	1040/1040	1.00000
$M_{11->8}^{10->8}$	1040/1040	1.00000
$M_{12->8}^{11->6}$	1040/1040	1.00000
M 1->9	1040/1040	1.00000
M _{2->9}	1040/1040	1.00000
M _{3->9}	1040/1040	1.00000
M _{4->9}	1040/1040	1.00000
M _{5->9}	1040/1040	1.00000
M _{6->9}	1040/1040	1.00000
M _{7->9}	1040/1040	1.00000
M _{8->9}	1040/1040	1.00000
M 10->9	1040/1040	1.00000
M 11->9	1040/1040	1.00000
M _{12->9}	1040/1040	1.00000
M _{1->10}	1040/1040	1.00000
M _{2->10}	1040/1040	1.00000
$M_{3->10}$	1040/1040	1.00000
M _{4->10}	1040/1040	1.00000
M _{5->10}	1040/1040	1.00000
M _{6->10}	1040/1040	1.00000
M 7->10	1040/1040	1.00000
M _{8->10}	1040/1040	1.00000
M _{9->10}	1040/1040	1.00000
M 11->10	1040/1040	1.00000
M _{12->10}	1040/1040	1.00000
M 1->11	1020/1020	1.00000
$M_{2->11}$	1061/1061	1.00000
$M_{3->11}$	1087/1087	1.00000
M _{4->11}	1085/1085	1.00000
M _{5->11}	967/967	1.00000
,		

M _{6->11}	1043/1043	1.00000
M _{7->11}	1031/1031	1.00000
M _{8->11}	1023/1023	1.00000
M _{9->11}	1039/1039	1.00000
$M_{10->11}$	1003/1003	1.00000
M _{12->11}	974/974	1.00000
$M_{1->12}$	1047/1047	1.00000
$M_{2->12}$	1062/1062	1.00000
$M_{3->12}$	1020/1020	1.00000
$M_{4\rightarrow 12}$	995/995	1.00000
$M_{5->12}$	1032/1032	1.00000
M _{6->12}	1091/1091	1.00000
$M_{7->12}$	1025/1025	1.00000
$M_{8->12}$	1086/1086	1.00000
$M_{9->12}$	1007/1007	1.00000
$M_{10->12}$	975/975	1.00000
M 11->12	1008/1008	1.00000
Genealogies	22885/150581	0.15198

MCMC-Autocorrelation and Effective MCMC Sample Size

Parameter	Autocorrelation	Effective Sampe Size
Θ_1	0.99012	14.89
$\Theta_2^{^1}$	0.99012	14.89
Θ_3^2	0.99012	14.89
Θ_4°	0.99012	14.89
Θ_{5}	0.99012	14.89
96	0.99012	14.89
\mathbf{p}_7°	0.99012	14.89
) ₈	0.99012	14.89
09	0.99012	14.89
9 ₁₀	0.99012	14.89
) ₁₁	0.82298	291.40
12	0.82091	295.09
1 2->1	0.98933	16.07
1 3->1	0.98933	16.07
A 4->1	0.98933	16.07
1 5->1	0.98933	16.07
1 6->1	0.98933	16.07
1 _{7->1}	0.98933	16.07
8->1	0.98933	16.07
1 9->1	0.98933	16.07
10->1	0.98933	16.07
11->1	0.98933	16.07
11->1	0.98933	16.07
1 1->2	0.98933	16.07
1 3->2	0.98933	16.07
1 4->2	0.98933	16.07
1 _{5->2}	0.98933	16.07
1 6->2	0.98933	16.07
1 7->2	0.98933	16.07
1 8->2	0.98933	16.07
1 9->2	0.98933	16.07
1 10->2	0.98933	16.07
11->2	0.98933	16.07
1 11->2 12->2	0.98933	16.07
1->3	0.98933	16.07
$A = \begin{cases} 1->3 \\ 2->3 \end{cases}$	0.98933	16.07
1 4->3	0.98933	16.07

M _{5->3}	0.98933	16.07
M _{6->3}	0.98933	16.07
M _{7->3}	0.98933	16.07
M _{8->3}	0.98933	16.07
$M_{9->3}$	0.98933	16.07
M 10->3	0.98933	16.07
M 11->3	0.98933	16.07
$M_{12->3}$	0.98933	16.07
M 1->4	0.98933	16.07
M _{2->4}	0.98933	16.07
$M_{3->4}$	0.98933	16.07
M _{5->4}	0.98933	16.07
M _{6->4}	0.98933	16.07
M 7->4	0.98933	16.07
M _{8->4}	0.98933	16.07
M _{9->4}	0.98933	16.07
M 10->4	0.98933	16.07
M 11->4	0.98933	16.07
M _{12->4}	0.98933	16.07
M 1->5	0.98933	16.07
$M_{2->5}$	0.98933	16.07
$M_{3->5}$	0.98933	16.07
M _{4->5}	0.98933	16.07
M _{6->5}	0.98933	16.07
M _{7->5}	0.98933	16.07
M _{8->5}	0.98933	16.07
M _{9->5}	0.98933	16.07
M _{10->5}	0.98933	16.07
M _{11->5}	0.98933	16.07
M _{12->5}	0.98933	16.07
M _{1->6}	0.98933	16.07
M _{2->6}	0.98933	16.07
M _{3->6}	0.98933	16.07
M _{4->6}	0.98933	16.07
M _{5->6}	0.98933	16.07
M _{7->6}	0.98933	16.07
M _{8->6}	0.98933	16.07
M _{9->6}	0.98933	16.07
M _{10->6}	0.98933	16.07
M _{11->6}	0.98933	16.07
M 12->6	0.98933	16.07
M 1->7	0.98933	16.07
M 2->7	0.98933	16.07
M 3->7	0.98933	16.07
M _{4->7}	0.98933	16.07

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M _{5->7}	0.98933	16.07
M _{6->7}	0.98933	16.07
M _{8->7}	0.98933	16.07
M _{9->7}	0.98933	16.07
M 10->7	0.98933	16.07
M 11->7	0.98933	16.07
M 12->7	0.98933	16.07
$M_{1->8}$	0.98933	16.07
$M_{2->8}$	0.98933	16.07
$M_{3->8}^{2->6}$	0.98933	16.07
$M_{4->8}^{3->6}$	0.98933	16.07
M 5->8	0.98933	16.07
I NA	0.98933	16.07
M 6->8 M 7->8	0.98933	16.07
$M_{9->8}^{7->8}$	0.98933	16.07
$M_{10->8}^{9->8}$	0.98933	16.07
M 11->8	0.98933	16.07
$M_{12->8}^{11->8}$	0.98933	16.07
M 1->9	0.98933	16.07
$M_{2\rightarrow 9}$	0.98933	16.07
M _{3->9}	0.98933	16.07
M _{4->9}	0.98933	16.07
M _{5->9}	0.98933	16.07
M _{6->9}	0.98933	16.07
M _{7->9}	0.98933	16.07
M _{8->9}	0.98933	16.07
M 10->9	0.98933	16.07
M 11->9	0.98933	16.07
M _{12->9}	0.98933	16.07
M _{1->10}	0.98933	16.07
M _{2->10}	0.98933	16.07
M _{3->10}	0.98933	16.07
M _{4->10}	0.98933	16.07
M _{5->10}	0.98933	16.07
M _{6->10}	0.98933	16.07
M _{7->10}	0.98933	16.07
M _{8->10}	0.98933	16.07
M _{9->10}	0.98933	16.07
M _{11->10}	0.98933	16.07
M _{12->10}	0.98933	16.07
M 1->11	0.85194	240.23
M 2->11	0.91872	128.59
M _{3->11}	0.87265	205.29
M _{4->11}	0.90976	142.24
M _{5->11}	0.87955	192.26

173.80 173.80 173.80 173.80 173.81 183.81 193.81 194.66 193.81 196.74 196.74 196.74 196.74 196.71 10.89338 169.27 12.311 10.91018 140.98 140.98 143.41 12.312 143.41 13.312 143.41 13.312 143.41 13.312 143.41 13.312 143.41 143.81	6->11	0.93956	93.77
S->11			
9->11 0.87784 196.74 10->11 0.89338 169.27 12->11 0.91018 140.98 1->12 0.89839 162.16 2->12 0.90930 143.41 3->12 0.88408 186.90 4->12 0.91595 131.69 5->12 0.86635 216.10 6->12 0.91936 125.96 7->12 0.92929 110.19 8->12 0.89651 164.11 9->12 0.94008 92.67 10->12 0.92430 118.83 11->12 0.89143 173.03			144.66
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$ \begin{array}{cccccccccccccccccccccccccccccccccccc$		0.89338	169.27
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$ \begin{array}{cccccccccccccccccccccccccccccccccccc$		0.89839	162.16
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$ \begin{array}{cccccccccccccccccccccccccccccccccccc$		0.88408	186.90
$ \begin{array}{cccccccccccccccccccccccccccccccccccc$		0.91595	131.69
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$\begin{array}{cccccccccccccccccccccccccccccccccccc$			
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	/->12	0.89651	
0.92430 118.83 173.03		0.94008	
$11 \rightarrow 12$ 0.89143 173.03			
11-212			
	11->12 Prob(DIG)1		

Potential Problems

This section reports potential problems with your run, but such reporting is often not very accurate. Whith many parameters in a multilocus analysis, it is very common that some parameters for some loci will not be very informative, triggering suggestions (for example to increase the prior range) that are not sensible. This suggestion tool will improve with time, therefore do not blindly follow its suggestions. If some parameters are flagged, inspect the tables carefully and judge wether an action is required. For example, if you run a Bayesian inference with sequence data, for macroscopic 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 routes 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.

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