# 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 Mon May 31 15:26:49 2021 Program finished at Mon May 31 17:08:51 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) 412311115

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	1	2	2	3	3	4	4	4	4	4	4
1 ElfinCo	*	*	*	*	0	0	0	0	0	0	0	0
1 Bamfiel	*	*	*	*	0	0	0	0	0	0	0	0
2 PortRen	0	0	*	*	*	*	*	*	*	*	*	*
2 WalkOnB	0	0	*	*	*	*	*	*	*	*	*	*
3 BodegaH	0	0	0	0	0	0	*	*	*	*	*	*
3 Davenpo	0	0	0	0	0	0	*	*	*	*	*	*
4 VistaDe	*	*	*	*	0	0	0	0	0	0	0	0
4 HazardR	*	*	*	*	0	0	0	0	0	0	0	0
4 Refugio	*	*	*	*	0	0	0	0	0	0	0	0
4 Carpint	*	*	*	*	0	0	0	0	0	0	0	0

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4 White	Po	*	* *	k	*	0	0	0	0	0	0	0	0		
4 LaJol	la	*	* *	k	*	0	0	0	0	0	0	0	0		
Order o	f parameters	3:													
1	Θ	1						<di< td=""><td>ispla</td><td>ayed:</td><td>&gt;</td><td></td><td></td><td></td><td></td></di<>	ispla	ayed:	>				
2	Θ							<di< td=""><td>ispla</td><td>ayed:</td><td>&gt;</td><td></td><td></td><td></td><td></td></di<>	ispla	ayed:	>				
5	M							<di< td=""><td>ispla</td><td>ayed:</td><td>&gt;</td><td></td><td></td><td></td><td></td></di<>	ispla	ayed:	>				
9	M	$\frac{1}{3->2}$						<di< td=""><td>ispla</td><td>ayed:</td><td>&gt;</td><td></td><td></td><td></td><td></td></di<>	ispla	ayed:	>				
10	M	1 4->2						<di< td=""><td>ispla</td><td>ayed:</td><td>&gt;</td><td></td><td></td><td></td><td></td></di<>	ispla	ayed:	>				
13	M							<di< td=""><td>ispla</td><td>ayed:</td><td>&gt;</td><td></td><td></td><td></td><td></td></di<>	ispla	ayed:	>				
14	M							<di< td=""><td>ispla</td><td>ayed:</td><td>&gt;</td><td></td><td></td><td></td><td></td></di<>	ispla	ayed:	>				
15	M								-	yed:					
		2->4							•	•					
   Mutation	rate among	loci:												Mutatio	n rate is constant
	J														
l Analvsis	strategy:													В	ayesian inference
														_	.,
Proposal	distributions	s for pa	aramete	r											
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Prior diet	ribution for p	arama	otor												
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	of chains														1
	ded steps [a]			_											1000
	ent (record	-		-											100
	er of concurr														3
	l (sampled) p				-	-									300000
Numbe	er of discard	trees	per chai	n	(burn	-in)									1000
Multiple I	Markov chair	ns:													
Static	heating sche	eme												4 chains w	vith temperatures
											1	1000	00.00		1.50 1.00
														Swap	ping interval is 1
Print opti	ons:														
Data fi	le:													//mcalifornia	anus_210528.mig
I															

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

Locus	Gene copies
1	19
1	23
1	15
1	16
1	7
1	17
1	19
1	23
1	16
1	19
1	11
1	8
1	193
	Locus  1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1

### Bayesian Analysis: Posterior distribution table

Locus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
1	$\Theta_1$	0.00001	0.00001	0.01001	0.06001	0.16001	0.07001	0.01000
1	$\Theta_2$	0.00001	0.00001	0.01001	0.06001	0.16001	0.07001	0.01000
1	M <sub>2-&gt;1</sub>	0.0	0.0	1000.0	20000.0	890000.0	419000.0	459963.5
1	M <sub>3-&gt;2</sub>	0.0	0.0	1000.0	20000.0	210000.0	455000.0	503755.4
1	M <sub>4-&gt;2</sub>	0.0	0.0	1000.0	20000.0	674000.0	383000.0	465193.1
1	M <sub>4-&gt;3</sub>	0.0	0.0	1000.0	20000.0	522000.0	415000.0	504129.8
1	M <sub>1-&gt;4</sub>	0.0	0.0	1000.0	20000.0	722000.0	369000.0	465407.2
1	M <sub>2-&gt;4</sub>	0.0	0.0	1000.0	20000.0	160000.0	467000.0	522089.0

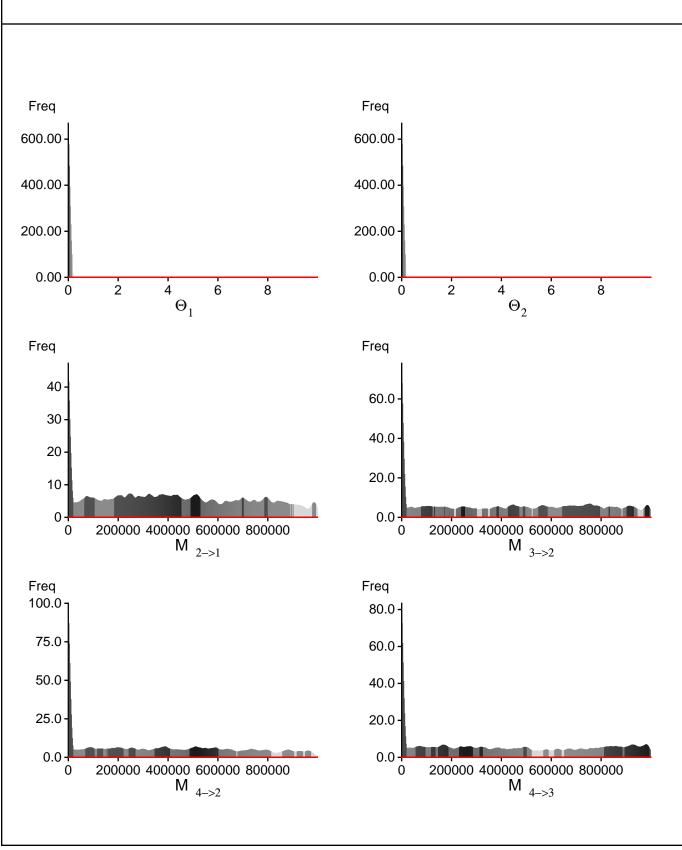
### Citation suggestions:

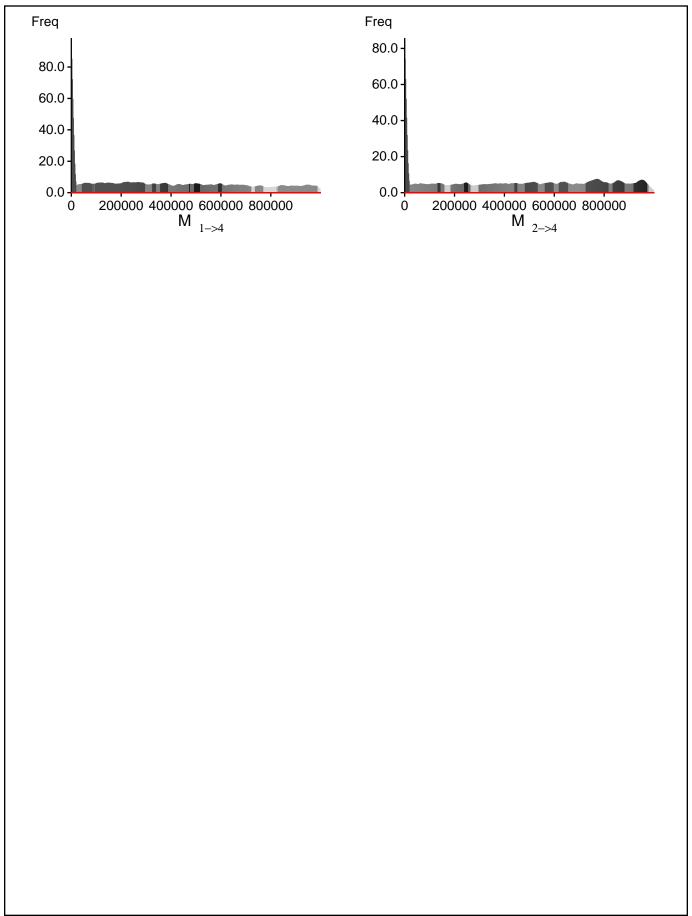
Beerli P., 2006. Comparison of Bayesian and maximum-likelihood inference of population genetic parameters. Bioinformatics 22:341-345

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 | 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	0.000000	(1a)
	0.000000	(1b)
Harmonic mean	inf	(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
$\Theta_1$	0/18718	0.00000
$\Theta_2^{^1}$	0/18793	0.00000
$M^{2}_{2->1}$	18744/18744	1.00000
$M_{3\rightarrow 2}^{2\rightarrow 1}$	18785/18785	1.00000
$M_{4\rightarrow 2}^{3\rightarrow 2}$	18662/18662	1.00000
M $_{4->3}^{4->2}$	18809/18809	1.00000
M 1->4	18847/18847	1.00000
$M_{2\rightarrow 4}$	18684/18684	1.00000
Genealogies	0/149958	0.00000

## MCMC-Autocorrelation and Effective MCMC Sample Size

Parameter	Autocorrelation	Effective Sampe Size		
$\Theta_1$	1.00000	0.00		
$\Theta_2^{-1}$	1.00000	0.00		
$\Theta_3^2$	0.00000	0.00		
$\Theta_4^{\circ}$	0.00000	0.00		
$M_{2->1}$	0.93828	95.44		
$M_{3->2}^{2->1}$	0.94613	82.97		
M $_{4->2}^{3->2}$	0.94164	90.09		
$M_{4->3}$	0.95443	69.89		
M 1->4	0.94355	87.06		
M $_{2->4}^{1->4}$	0.94713	81.40		
Ln[Prob(D G)]	1.00000	0.00		

### 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.

Param 1: Effective sample size of run seems too short!

Param 2: Effective sample size of run seems too short!