mexico

POPULATION SIZE, MIGRATION, DIVERGENCE, ASSIGNMENT, HISTORY

Bayesian inference using the structured coalescent

Migrate-n version 5.0.3a [January-08-2018]

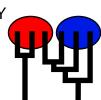
Using Intel AVX (Advanced Vector Extensions)

Compiled for PARALLEL computer architectures

One master and 40 compute nodes are available.

Program started at Sun Oct 28 13:49:27 2018

Program finished at Sun Oct 28 13:58:57 2018 [Runtime:0000:00:09:30]



Options

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) 1887979638

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

d = row population split off column population, D = split and then migration

Population 1

1 mexico *

Order of parameters:

1 Θ_1 <displayed>

Mutation rate among loci: Mutation rate is constant for all loci

Analysis strategy:
-Population size estimation:

Bayesian inference
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 -抵xponential 0.000000 0.100 0.900 - 1500 0.50000

[-1 -1 means priors were set globally]

Markov chain settings: Long chain

Number of chains 1

Recorded steps [a]10000Increment (record every x step [b]1000Number of concurrent chains (replicates) [c]5

Visited (sampled) parameter values [a*b*c] 50000000

Number of discard trees per chain (burn-in) 5000

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: influenza_h1n1_mexico_2014_infile

Haplotyping is turned on:

Output file: outfile_influenza_0.7

Posterior distribution raw histogram file: bayesfile

Raw data from the MCMC run: bayesallfile_influenza_0.7

Print data: No

Print genealogies [only some for some data type]:

Data summary

Data file: influenza_h1n1_mexico_2014_infile

Datatype: Haplotype data

Number of loci:

Mutationmodel:

| Locus Sublocus | | Mutationmodel | Mutationmodel parameters | |
|----------------|---|----------------|---|--|
| 1 | 1 | Felsenstein 84 | [Bf:0.34 0.19 0.26 0.22, t/t ratio=2.000] | |
| 2 | 1 | Felsenstein 84 | [Bf:0.36 0.19 0.23 0.22, t/t ratio=2.000] | |
| 3 | 1 | Felsenstein 84 | [Bf:0.33 0.20 0.24 0.23, t/t ratio=2.000] | |
| 4 | 1 | Felsenstein 84 | [Bf:0.35 0.19 0.22 0.24, t/t ratio=2.000] | |
| 5 | 1 | Felsenstein 84 | [Bf:0.33 0.20 0.26 0.21, t/t ratio=2.000] | |
| 6 | 1 | Felsenstein 84 | [Bf:0.32 0.18 0.24 0.26, t/t ratio=2.000] | |
| 7 | 1 | Felsenstein 84 | [Bf:0.29 0.21 0.26 0.24, t/t ratio=2.000] | |
| 8 | 1 | Felsenstein 84 | [Bf:0.32 0.19 0.24 0.25, t/t ratio=2.000] | |

Sites per locus

| Locus | Sites |
|-------|-------|
| 1 | 2280 |
| 2 | 2274 |
| 3 | 2151 |
| 4 | 1701 |
| 5 | 1497 |
| 6 | 1410 |
| 7 | 982 |
| 8 | 863 |

Site rate variation and probabilities:

Locus Sublocus Region type Rate of change Probability Patch size

| 1 | 1 | 1 | 1.000 | 1.000 | 1.000 |
|--------|-------|---|-------|-------|-------|
| 2 | 1 | 1 | 1.000 | 1.000 | 1.000 |
| 3 | 1 | 1 | 1.000 | 1.000 | 1.000 |
| 4 | 1 | 1 | 1.000 | 1.000 | 1.000 |
| 5 | 1 | 1 | 1.000 | 1.000 | 1.000 |
| 6 | 1 | 1 | 1.000 | 1.000 | 1.000 |
| 7 | 1 | 1 | 1.000 | 1.000 | 1.000 |
| 8 | 1 | 1 | 1.000 | 1.000 | 1.000 |
| Popula | ation | | | | Locus |

Gene copies

data (missing)

Migrate 5.0.3a: (http://popgen.sc.fsu.edu) [program run on 13:49:27]

| 1 mexico | 1 | 6 | |
|--|---|---|-----|
| | 2 | 6 | |
| | 3 | 6 | |
| | 4 | 6 | |
| | 5 | 6 | |
| | 6 | 6 | |
| | 7 | 6 | |
| | 8 | 6 | |
| Total of all populations | 1 | 6 | (0) |
| | 2 | 6 | (0) |
| | 3 | 6 | (0) |
| | 4 | 6 | (0) |
| | 5 | 6 | (0) |
| | 6 | 6 | (0) |
| | 7 | 6 | (0) |
| | 8 | 6 | (0) |
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| Migrate 5.0 3a: (http://popgen.sc.fsu.edu) [program run on 13:49:27] | | | |

Bayesian Analysis: Posterior distribution table

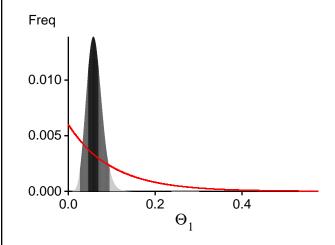
| Locus | Parameter | 2.5% | 25.0% | Mode | 75.0% | 97.5% | Median | Mean |
|-------|------------|---------|---------|---------|---------|---------|---------|---------|
| 1 | Θ_1 | 0.00900 | 0.03600 | 0.05670 | 0.08460 | 0.17520 | 0.07230 | 0.08145 |
| 2 | Θ_1 | 0.00180 | 0.02760 | 0.04890 | 0.07860 | 0.17160 | 0.06630 | 0.07615 |
| 3 | Θ_1 | 0.00720 | 0.04200 | 0.05550 | 0.07140 | 0.17520 | 0.07470 | 0.08508 |
| 4 | Θ_1 | 0.00000 | 0.03240 | 0.05550 | 0.09000 | 0.27420 | 0.07590 | 0.08678 |
| 5 | Θ_1 | 0.00000 | 0.02580 | 0.04770 | 0.07680 | 0.16740 | 0.06450 | 0.07390 |
| 6 | Θ_1 | 0.00180 | 0.02880 | 0.05130 | 0.08100 | 0.17580 | 0.06810 | 0.07813 |
| 7 | Θ_1 | 0.00000 | 0.01380 | 0.02910 | 0.04620 | 0.10560 | 0.03870 | 0.04385 |
| 8 | Θ_1 | 0.00720 | 0.03420 | 0.05430 | 0.08400 | 0.17340 | 0.07110 | 0.08005 |
| All | Θ_1 | 0.02640 | 0.04560 | 0.05790 | 0.06960 | 0.09540 | 0.05970 | 0.06034 |

Citation suggestions:

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

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

| Locus | TI(1a) | BTI(1b) | SS(2) | HS(3) |
|-------|-----------|-----------|----------|-----------|
| 1 | -3375.30 | -3315.19 | -3314.16 | -3343.55 |
| 2 | -3360.20 | -3299.11 | -nan | -3327.90 |
| 3 | -3235.54 | -3176.03 | -nan | -3203.35 |
| 4 | -2636.36 | -2582.50 | -nan | -2603.07 |
| 5 | -2232.16 | -2192.35 | -nan | -2211.35 |
| 6 | -2125.95 | -2085.44 | -nan | -2103.55 |
| 7 | -1413.04 | -1390.46 | -nan | -1403.73 |
| 8 | -1333.99 | -1306.16 | -nan | -1317.91 |
| All | -19716.52 | -19351.23 | -nan | -19518.40 |

- (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 = -3.977822]

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 |
|-------------|---------------------|---------|
| Θ_1 | 108465964/199993872 | 0.54235 |
| Genealogies | 33017252/200006128 | 0.16508 |

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 |
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