Current protocols example dataset

POPULATION SIZE, MIGRATION, DIVERGENCE, ASSIGNMENT, HISTORY

Bayesian inference using the structured coalescent

Migrate-n version 4.4.3(git:) [March-21-2019]

Using Intel AVX (Advanced Vector Extensions)

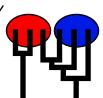
Compiled for PARALLEL computer architectures

One master and 8 compute nodes are available.

Compiled for a SYMMETRIC multiprocessors (Grandcentral)

Program started at Sun Jun 2 00:01:40 2019

Program finished at Sun Jun 2 00:17:03 2019 [Runtime:0000:00:15:23]



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) 2612977134

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 2 3 1 Arbon_1 * 0 0 2 Berg_2 * * 0 3 Chur_3 0 * *

Order of parameters:

infile

| | | | | | Current | protocols exa | mple dataset 2 |
|-----------------------------|-------------|----------------|--|--------|---------|----------------|--------------------|
| 1 Θ_1 | | | <displa< td=""><td>yed></td><td></td><td></td><td></td></displa<> | yed> | | | |
| Θ_2 | | | <displa< td=""><td>yed></td><td></td><td></td><td></td></displa<> | yed> | | | |
| Θ_3 | | | <displa< td=""><td>•</td><td></td><td></td><td></td></displa<> | • | | | |
| 4 M ₁₋₃ | >2 | | <displa< td=""><td>-</td><td></td><td></td><td></td></displa<> | - | | | |
| 5 M ₂₋₃ | >3 | | <displa< td=""><td>yed></td><td></td><td></td><td></td></displa<> | yed> | | | |
| Mutation rate among loc | i: | | | | Mutati | on rate is con | stant for all loci |
| Analysis strategy: | | | | | | Baye | esian inference |
| -Population size estima | tion: | | | | | Exponen | tial Distribution |
| -Geneflow estimation: | | | | | | Exponen | tial Distribution |
| Proposal distributions fo | r paramete | er | | | | | |
| Parameter | | Prop | oosal | | | | |
| Theta | M | letropolis sam | pling | | | | |
| M | M | letropolis sam | pling | | | | |
| Divergence | M | letropolis sam | pling | | | | |
| Divergence Spread | N | letropolis sam | pling | | | | |
| Genealogy | N | letropolis-Has | tings | | | | |
| Prior distribution for para | ameter | | | | | | |
| Parameter | Prior | Minimum | MeanMa | aximum | Delta | Bins U | JpdateFreq |
| 1 Theta ** | Uniform | 0.000000 | 0.050 | 0.100 | 0.010 | 1500 | 0.10000 |
| 2 Theta ** | Uniform | 0.000000 | 0.050 | 0.100 | 0.010 | 1500 | 0.10000 |
| 3 Theta ** | Uniform | 0.000000 | 0.050 | 0.100 | 0.010 | 1500 | 0.10000 |
| 4 M ** | Uniform | 0.000000 | 500.0 | 1000. | 100.0 | 1500 | 0.10000 |
| 5 M ** | Uniform | 0.000000 | 500.0 | 1000. | 100.0 | 1500 | 0.10000 |
| [* * means priors were s | et globally | 1 | | | | | |
| Markov chain settings: | | | | | | | Long chain |
| Number of chains | | | | | | | 1 |
| Recorded steps [a] | | | | | | | 5000 |
| Increment (record eve | ry x step [| b] | | | | | 100 |
| Number of concurrent | chains (re | plicates) [c] | | | | | 1 |
| Visited (sampled) para | ameter val | ues [a*b*c] | | | | | 500000 |
| Number of discard tre- | es per cha | in (burn-in) | | | | | 5000 |
| Multiple Markov chains: | | | | | | | |
| Static heating scheme | • | | | | | 4 chains with | temperatures |
| | | | | 1000 | 00.000 | | 1.50 1.00 |
| | | | | | | Swappii | ng interval is 1 |
| Print options: | | | | | | | |
| 1 5 | | | | | | | |

Data file:

| Haplotyping is turned on: | NC |
|---|-----------------|
| Output file: | outfile_model1 |
| Posterior distribution raw histogram file: | bayesfile |
| Raw data from the MCMC run: | bayesallfile.gz |
| Print data: | No |
| Print genealogies [only some for some data type]: | None |
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Data summary

| Data file: | infile |
|-----------------|----------------|
| Datatype: | Haplotype data |
| Number of loci: | 10 |

Mutationmodel parameters

| Mutationmodel: | | | |
|----------------|--|--|--|
| Locus Sublocus | | | |

| 1 | 1 | Jukes-Cantor | [Basefreq: =0.25] |
|----|---|--------------|-------------------|
| 2 | 1 | Jukes-Cantor | [Basefreq: =0.25] |
| 3 | 1 | Jukes-Cantor | [Basefreq: =0.25] |
| 4 | 1 | Jukes-Cantor | [Basefreq: =0.25] |
| 5 | 1 | Jukes-Cantor | [Basefreq: =0.25] |
| 6 | 1 | Jukes-Cantor | [Basefreq: =0.25] |
| 7 | 1 | Jukes-Cantor | [Basefreq: =0.25] |
| 8 | 1 | Jukes-Cantor | [Basefreq: =0.25] |
| 9 | 1 | Jukes-Cantor | [Basefreq: =0.25] |
| 10 | 1 | Jukes-Cantor | [Basefreq: =0.25] |

Mutationmodel

Sites per locus

| Locus | Sites |
|-------|-------|
| 1 | 1000 |
| 2 | 1000 |
| 3 | 1000 |
| 4 | 1000 |
| 5 | 1000 |
| 6 | 1000 |
| 7 | 1000 |
| 8 | 1000 |
| 9 | 1000 |
| 10 | 1000 |

Site rate variation and probabilities:

Locus Sublocus Region type Rate of change Probability Patch size

| ı | | | | | | |
|---|---|---|---|-------|-------|-------|
| | 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 | | |
| 9 1 | 1 | 1.000 | 1.000 | 1.000 | | |
| 10 1 | 1 | 1.000 | 1.000 | 1.000 | | |
| Population | · | 1.000 | 11000 | Locus | Gene c | opies |
| | | | | | data | (missing) |
| 1 Arbon_1 | | | | 1 | 10 | (11 3) |
| _ | | | | 2 | 10 | |
| | | | | 3 | 10 | |
| | | | | 4 | 10 | |
| | | | | 5 | 10 | |
| | | | | 6 | 10 | |
| | | | | 7 | 10 | |
| | | | | 8 | 10 | |
| | | | | 9 | 10 | |
| | | | | 10 | 10 | |
| 2 Berg_2 | | | | 1 | 10 | |
| <u> </u> | | | | 2 | 10 | |
| | | | | 3 | 10 | |
| | | | | 4 | 10 | |
| | | | | 5 | 10 | |
| | | | | 6 | 10 | |
| | | | | 7 | 10 | |
| | | | | 8 | 10 | |
| | | | | 9 | 10 | |
| | | | | 10 | 10 | |
| 3 Chur_3 | | | | 1 | 10 | |
| | | | | 2 | 10 | |
| | | | | 3 | 10 | |
| | | | | 4 | 10 | |
| | | | | 5 | 10 | |
| | | | | 6 | 10 | |
| | | | | 7 | 10 | |
| | | | | 8 | 10 | |
| | | | | 9 | 10 | |
| | | | | 10 | 10 | |
| Total of all popu | lations | | | 1 | 30 | (0) |
| | | | | 2 | 30 | (0) |
| | | | | 3 | 30 | (0) |
| | | | | 4 | 30 | (0) |
| | | | | 5 | 30 | (0) |
| | | | | 6 | 30 | (0) |
| | | | | 7 | 30 | (0) |
| | | | | 8 | 30 | (0) |
| | | | | 9 | 30 | (0) |
| | | | | | | . , |

| Curre | mple dataset 6 | |
|-------|----------------|-----|
| 10 | 30 | (0) |
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Bayesian Analysis: Posterior distribution table

| Locus | Parameter | 2.5% | 25.0% | Mode | 75.0% | 97.5% | Median | Mean |
|-------|----------------------|---------|---------|---------|---------|---------|---------|---------|
| 1 | Θ_1 | 0.00360 | 0.00713 | 0.00877 | 0.00993 | 0.01620 | 0.00937 | 0.01037 |
| 1 | Θ_2 | 0.00140 | 0.00373 | 0.00437 | 0.00513 | 0.01013 | 0.00503 | 0.00565 |
| 1 | Θ_3^- | 0.00393 | 0.00820 | 0.00937 | 0.01067 | 0.02427 | 0.01363 | 0.01552 |
| 1 | M _{1->2} | 11.333 | 88.667 | 91.000 | 95.333 | 204.667 | 107.000 | 143.048 |
| 1 | $M_{2->3}$ | 0.000 | 48.000 | 57.000 | 60.667 | 170.667 | 71.667 | 92.589 |
| 2 | Θ_1 | 0.00247 | 0.00460 | 0.00550 | 0.00620 | 0.01127 | 0.00650 | 0.00702 |
| 2 | Θ_2 | 0.00080 | 0.00220 | 0.00350 | 0.00440 | 0.00913 | 0.00430 | 0.00499 |
| 2 | Θ_3 | 0.00133 | 0.00287 | 0.00323 | 0.00540 | 0.00960 | 0.00510 | 0.00578 |
| 2 | M _{1->2} | 32.667 | 125.333 | 132.333 | 137.333 | 230.000 | 148.333 | 162.097 |
| 2 | $M_{2->3}$ | 4.000 | 38.667 | 42.333 | 48.000 | 208.667 | 95.667 | 130.782 |
| 3 | Θ_1 | 0.00253 | 0.00513 | 0.00543 | 0.00553 | 0.01167 | 0.00670 | 0.00707 |
| 3 | Θ_2 | 0.00147 | 0.00393 | 0.00643 | 0.00920 | 0.02460 | 0.01583 | 0.02423 |
| 3 | Θ_3^- | 0.00540 | 0.00940 | 0.00957 | 0.01020 | 0.02240 | 0.01323 | 0.01470 |
| 3 | M _{1->2} | 337.333 | 412.000 | 413.000 | 413.333 | 462.667 | 499.000 | 519.034 |
| 3 | $M_{2->3}$ | 0.667 | 12.667 | 19.667 | 24.000 | 97.333 | 37.000 | 44.511 |
| 4 | Θ_1 | 0.00187 | 0.00353 | 0.00523 | 0.00653 | 0.01247 | 0.00630 | 0.00795 |
| 4 | Θ_2 | 0.00073 | 0.00173 | 0.00310 | 0.00473 | 0.01380 | 0.00537 | 0.00703 |
| 4 | Θ_3 | 0.00347 | 0.01080 | 0.01130 | 0.01287 | 0.01900 | 0.01143 | 0.01282 |
| 4 | M _{1->2} | 21.333 | 83.333 | 90.333 | 98.000 | 268.000 | 199.000 | 248.163 |
| 4 | $M_{2->3}$ | 0.000 | 8.667 | 16.333 | 40.667 | 158.000 | 55.667 | 83.091 |
| 5 | Θ_1 | 0.00140 | 0.00227 | 0.00377 | 0.00553 | 0.01660 | 0.00743 | 0.00977 |
| 5 | Θ_2 | 0.00067 | 0.00340 | 0.00377 | 0.00500 | 0.01147 | 0.00537 | 0.00723 |
| 5 | Θ_3 | 0.00500 | 0.01267 | 0.01323 | 0.01413 | 0.02900 | 0.01763 | 0.01974 |
| 5 | M _{1->2} | 132.000 | 133.333 | 135.000 | 142.667 | 150.000 | 409.667 | 449.100 |
| 5 | $M_{2->3}$ | 0.000 | 17.333 | 36.333 | 45.333 | 170.000 | 64.333 | 105.032 |
| 6 | Θ_1 | 0.00113 | 0.00200 | 0.00250 | 0.00367 | 0.00593 | 0.00330 | 0.00348 |
| 6 | Θ_2 | 0.00067 | 0.00167 | 0.00410 | 0.00540 | 0.01687 | 0.00577 | 0.01001 |
| 6 | Θ_3 | 0.00520 | 0.01173 | 0.01197 | 0.01273 | 0.02447 | 0.01510 | 0.01705 |
| 6 | M _{1->2} | 826.667 | 977.333 | 981.000 | 982.667 | 999.333 | 813.000 | 769.159 |
| 6 | $M_{2->3}$ | 12.000 | 64.000 | 65.000 | 70.000 | 204.000 | 100.333 | 116.489 |

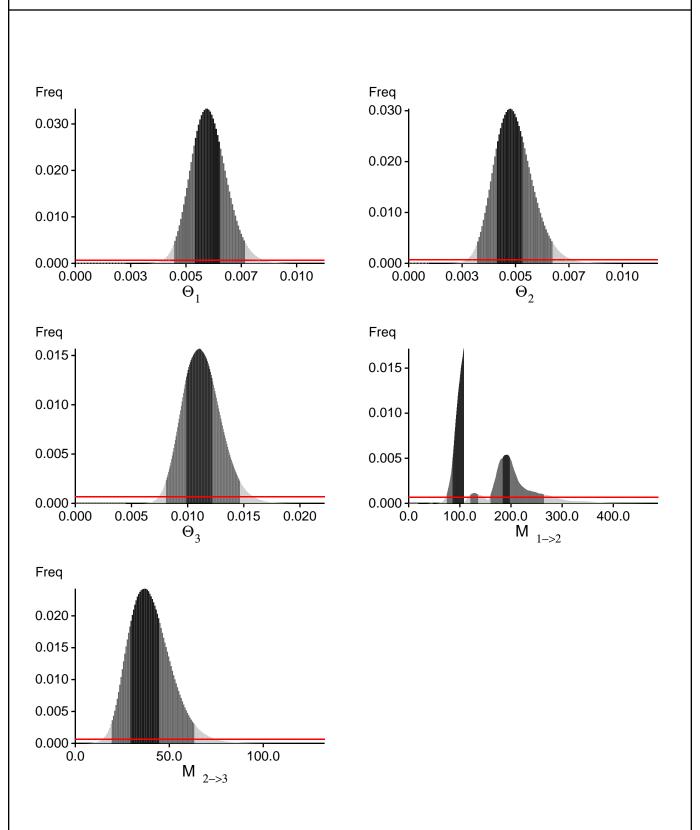
| 7 | Θ_1 | 0.00427 | 0.00713 | 0.00803 | 0.00973 | 0.01547 | 0.00930 | 0.00973 |
|-----|----------------------|---------|---------|---------|---------|---------|---------|---------|
| 7 | Θ_2 | 0.00287 | 0.00553 | 0.00570 | 0.00607 | 0.02333 | 0.02663 | 0.03432 |
| 7 | Θ_3 | 0.00653 | 0.01273 | 0.01317 | 0.01360 | 0.02753 | 0.01623 | 0.01794 |
| 7 | M _{1->2} | 537.333 | 548.667 | 549.667 | 553.333 | 560.667 | 645.000 | 633.997 |
| 7 | $M_{2->3}$ | 0.000 | 17.333 | 21.000 | 27.333 | 102.667 | 38.333 | 50.415 |
| | | | | | | | | |
| 8 | Θ_1 | 0.00287 | 0.00467 | 0.00563 | 0.00727 | 0.01513 | 0.00723 | 0.00889 |
| 8 | Θ_2 | 0.00153 | 0.00493 | 0.00543 | 0.00560 | 0.01707 | 0.00910 | 0.01121 |
| 8 | Θ_3 | 0.00293 | 0.00647 | 0.00710 | 0.00727 | 0.01573 | 0.00917 | 0.01032 |
| 8 | M _{1->2} | 22.667 | 56.000 | 69.000 | 70.000 | 272.000 | 128.333 | 185.033 |
| 8 | $M_{2->3}$ | 1.333 | 53.333 | 56.333 | 71.333 | 203.333 | 114.333 | 147.983 |
| | | | | | | | | |
| 9 | Θ_1 | 0.00367 | 0.00660 | 0.00830 | 0.00967 | 0.01867 | 0.01017 | 0.01136 |
| 9 | Θ_2 | 0.00113 | 0.00253 | 0.00337 | 0.00440 | 0.00940 | 0.00470 | 0.00517 |
| 9 | Θ_3 | 0.01440 | 0.01513 | 0.01530 | 0.01533 | 0.02167 | 0.04097 | 0.04578 |
| 9 | M _{1->2} | 16.667 | 59.333 | 83.000 | 92.000 | 179.333 | 94.333 | 115.036 |
| 9 | $M_{2->3}$ | 618.667 | 625.333 | 626.333 | 626.667 | 642.667 | 473.667 | 483.016 |
| | | | | | | | | |
| 10 | Θ_1 | 0.00267 | 0.00400 | 0.00523 | 0.00620 | 0.01087 | 0.00617 | 0.00658 |
| 10 | Θ_2 | 0.00167 | 0.00620 | 0.00650 | 0.00793 | 0.02133 | 0.00963 | 0.01253 |
| 10 | Θ_3 | 0.00687 | 0.01287 | 0.01437 | 0.01500 | 0.02807 | 0.01763 | 0.01965 |
| 10 | M _{1->2} | 342.000 | 402.000 | 403.000 | 404.667 | 428.000 | 413.667 | 427.101 |
| 10 | $M_{2->3}$ | 0.000 | 15.333 | 20.333 | 32.000 | 96.667 | 37.000 | 50.401 |
| | | 0.00115 | 0.00=00 | 0.00=0= | 0.000=0 | 0.00=0= | 0.00040 | 0.0000= |
| All | Θ_1 | 0.00440 | 0.00533 | 0.00597 | 0.00653 | 0.00767 | 0.00610 | 0.00605 |
| All | Θ_2 | 0.00313 | 0.00407 | 0.00477 | 0.00533 | 0.00673 | 0.00490 | 0.00492 |
| All | Θ_3 | 0.00800 | 0.00980 | 0.01110 | 0.01220 | 0.01467 | 0.01123 | 0.01128 |
| All | M _{1->2} | 74.000 | 85.333 | 107.667 | 108.000 | 108.000 | 159.000 | 157.942 |
| All | $M_{2->3}$ | 18.667 | 28.667 | 37.000 | 44.667 | 63.333 | 39.667 | 39.941 |
| | | | | | | | | |

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 | Raw thermodynamic score(1a) | Bezier approximation score(1b) | Harmonic mean(2) |
|-------|-----------------------------|--------------------------------|------------------|
| 1 | -2786.78 | -2367.22 | -2295.30 |
| 2 | -2320.86 | -2010.66 | -1958.44 |
| 3 | -2496.88 | -2163.85 | -2115.76 |
| 4 | -2844.37 | -2324.81 | -2239.87 |
| 5 | -2622.60 | -2292.59 | -2181.90 |
| 6 | -2862.65 | -2526.67 | -2374.74 |
| 7 | -2678.95 | -2265.09 | -2201.38 |
| 8 | -2808.78 | -2319.77 | -2229.52 |
| 9 | -2762.79 | -2341.81 | -2275.50 |
| 10 | -2572.29 | -2213.13 | -2152.15 |
| All | -26716.72 | -22785.37 | -21984.34 |

(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 [Scaling factor = 40.230092]

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 | 227084/500435 | 0.45377 |
| Θ_2 | 269735/500584 | 0.53884 |
| Θ_3^2 | 238606/499322 | 0.47786 |
| $M_{1\rightarrow 2}$ | 236063/498334 | 0.47370 |
| $M_{2->3}$ | 261122/500345 | 0.52188 |
| Genealogies | 370637/2500980 | 0.14820 |

MCMC-Autocorrelation and Effective MCMC Sample Size

| Parameter | Autocorrelation | Effective Sampe Size |
|--|-----------------|----------------------|
| Θ_1 | 0.52454 | 25267.70 |
| Θ_2 | 0.56819 | 22884.81 |
| $\begin{bmatrix} \Theta_3^2 \\ M_{-1} > 2 \end{bmatrix}$ | 0.34132 | 35923.26 |
| $M_{1\rightarrow 2}$ | 0.40080 | 33138.37 |
| M _{2->3} | 0.45100 | 28173.07 |
| Genealogies | 0.72194 | 12046.30 |

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 reducing number of parameters may help in such situations.

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, Param 6 (Locus 6): Upper prior boundary seems too low!