

# Applications of nested sampling in systems biology

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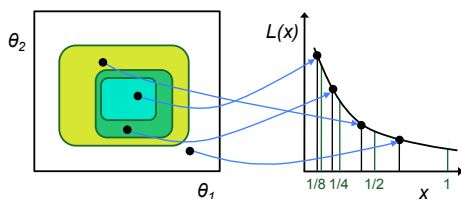
**Stochastic models** are commonly used in systems biology to represent the interaction of small numbers of molecules, and the discrete states that a molecule might adopt. The optimisation of complex stochastic models is challenging as, typically, they cannot be solved analytically.

**Nested sampling** is an effective method for sampling the posterior distributions of model parameters. The samples are obtained as a by-product of calculating the *Bayesian evidence*. Nested sampling requires a likelihood function, and, in the context of systems biology, the extent to which the data is explained by a given set of model parameters can be computed by an approximate log likelihood function derived from a number of Gillespie simulations. This optimisation strategy is therefore generic, and applicable to kinetic data and steady-state distributions.

We have demonstrated that this approach performs well as an optimiser for a number of systems biology models, including models of circadian rhythms. The method can also be used for model comparison – which will be the topic of future work.

## Nested sampling

Nested sampling explores the Bayesian evidence, transforming the multi-dimensional integral for the evidence into a one-dimensional integral over the prior mass (Skilling, 2006). The sorted likelihood function  $L(x)$  is used as an evolving constraint in the generation of a set of objects  $\{x_p, x_s\}$ , each object is an array of values randomly sampled from the prior range of a model parameter.



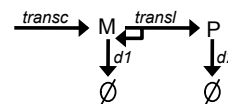
Let  $x(L)$  be the prior mass enclosed within the contour  $L(\theta) = L$ , § and  $L(x)$  be the contour value such that the volume enclosed is  $x$ .  
 $Z = \int dx L(x)$   
Samples  $\bullet$  are drawn uniformly from  $L(x) > L^*$

Given a set of  $n$  active objects, the worst is replaced by a new object  $x_{new}$ , subject to the constraint  $L(x_{new}) > L^*$ . The new object is discovered by an exploration method that takes one of the remaining  $n-1$  objects as a starting point. The constraint  $L^*$  is then set equal to the log likelihood of the worst object, and the process repeats. The objects eliminated from the active set are added to the set of *posterior* samples, and can be analysed further. The new object returned by the exploration method must be an unbiased sample of the prior subject to the constraint on the likelihood  $L^*$ : unbiased sampling is necessary for correctly computing  $Z$ , and can be achieved by the correct choice of exploration function (Murray, 2007).

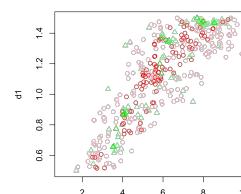
Nested sampling requires a likelihood function to be defined in order to compute the probability of the data given the model. An approximate likelihood can be derived from a number of realisations of a model obtained by stochastic simulation.

§ following MacKay <http://www.inference.phy.cam.ac.uk/bayesys/>

## A simple model of transcription and translation

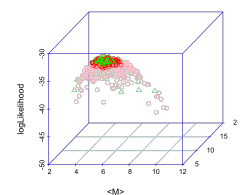


A stochastic model of the transcription of mRNA (M) and the translation of protein (P) from mRNA was fitted to a synthetic data set of samples from the distributions of M and P. Nested sampling reveals the obvious trade-off between *transc* and *d1* in establishing the mean  $\langle M \rangle$ .



Circles are (*transc*, *d1*) values in objects obtained when  $n=100$ ; triangles are values obtained when  $n=20$ . Lighter colours indicate posterior samples, darker colours indicate active objects.

The log likelihood is maximised when  $\langle M \rangle = \text{transc}/d1 = 5.14$  and  $\langle P \rangle = \text{transc} \times \text{transl}/d1 \times d2 = 10.28$ . The posterior samples generated by nested sampling converge on these ratios as shown below.



**Future work** will consider more complex systems biology models, and the use of Bayesian evidence for model selection.

## References

- Murray, I. (2007) *Advances in Markov chain Monte Carlo methods*. PhD Thesis University College London.
- Skilling, J. (2006) Nested sampling for Bayesian computations. *Proc. ISBA 2006*.
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