

Drift Diffusion Modeling

Megan Kairiss

April 12, 2019

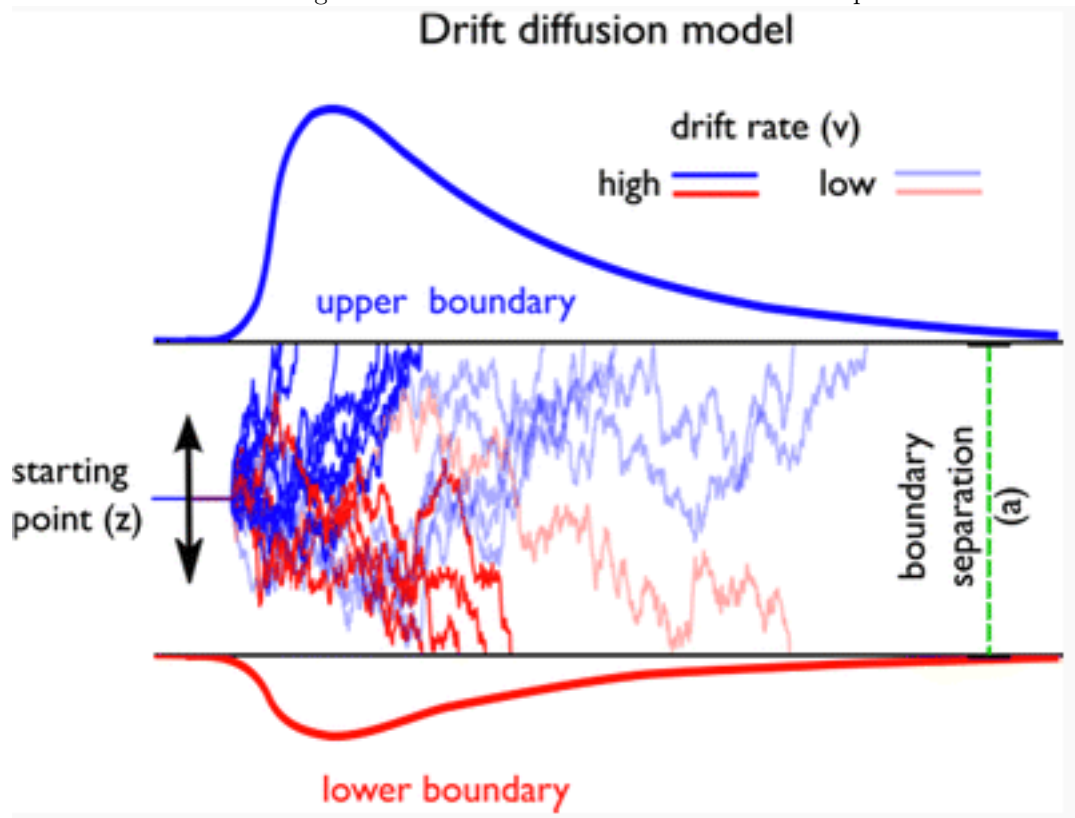
Introduction and Background

Attempts of using diffusion processes to explain the noisy process of decision making has been around for more than 40 years, yet using these diffusion models on the choice reaction time of animal behaviour is still yet to be mainstream. We present an application of Drift Diffusion modeling, alternatively referred to as a two-alternative forced choice, fitted to mouse subjects under an ambiguous choice response task. Recent advances in computational methods have enabled us to compute a full hierarchical bayesian posterior distribution of Drift Diffusion model parameters, with the aide of MCMC sampling (Markov Chain Monte Carlo) to estimate the joint distribution of free parameters; with the assumption that computing parameter estimates across a population can strengthen the recovery of parameters of the individual.

Theoretical Formalization

Drift Diffusion models allow for an experimental design with only two alternative options. As first proposed by Ratcliff (1978), subjects noisely fluctuate between two options until a certain decision threshold has been reached, akin to the vibrating frequencies of a tuning fork. This “fluctuation” can be represented by a drift rate; parameterized in the model as v . This slope must have a starting point, which we parameterize as t_0 . The subject samples between these two alternative options, each having independent decision thresholds, hereonout referred to as A^- and A^+ .

Figure 1: Visualization of Drift Diffusion model parameters



Experimental Approach

Experimental Design

Data Collection

Data Processing

Model Application

HDDM

Fast-DM

Discussion

References