## Age Zircon model

## Main model parameters

Let us define a dataset of N dated zircons from S samples. We indicate with  $\mathbf{z} = \{z_1, \dots, z_N\}$  the (unknown) ages of the zircons and with  $\mathbf{x} = \{x_1, \dots, x_S\}$  the ages of the samples. The samples are ordered according on their depth and we assume their age strictly follows that order such that  $x_i > x_{i-1}$ .

The exact age of each zircon  $(z_i)$  is assumed to be unknown but linked to its measured age, which is expressed as a mean  $\mu_i$  and a standard deviation  $\sigma_i$ . The mean ages  $\mu$  and a standard deviations  $\sigma$  of all zircons represent the input data of the model. Since the age of zircons can be measured based on different methods (e.g.  $m \in \{1, ..., M\}$ , e.g. Zr-U-Pb, ZFT, Ar-Ar), the true age of a zircon is assumed to be further affected by the how it was measured. We compute the likelihood of the age of a zircon i based on dating method m as:

$$P(\mu_i|z_i,\sigma_i,\eta_m,\epsilon_m) \sim \mathcal{N}(z_i+\eta_m,\sigma_i+\epsilon_m)$$
 (1)

where  $\eta_m \in \mathbb{R}$  is the age bias associated with the measurement method m and  $\epsilon_m \in \mathbb{R}^+$  is the biased in the estimated error of the measurement method m. The parameters  $z_i, \eta_m$ 

and  $\epsilon_m$  are considered as unknown and estimated from the data using a Bayesian algorithm. The prior probability of  $z_{i,j}$  is modeled by a Cauchy distribution with mode equal to the age of the sample  $x_j$  and a sample-specific scale  $s_j$ :

$$P(z_{i,j}|x_j,s_j) \sim \mathcal{C}(x_j,s_j). \tag{2}$$

The value of  $x_j$ , the sample age, is determined by two latent variables and constrained by sampled values of  $\mathbf{z}$  such that  $x_i > x_{i-1}$  for  $i \in \{2, ..., S\}$ . Specifically we define as

$$\zeta_j = \min(z_n), \text{ for } n \in \{j, \dots, N\},\tag{3}$$

the minimum age across all zircons included in sample j and in all older samples. Thus,  $\zeta_j$  represents the maximum boundary for the age of sample j. Under this notation we define the age of a sample as:

$$x_{j} = \begin{cases} r_{j}(\zeta_{j}), & \text{if } j = 1\\ x_{j-1} + r_{j}(\zeta_{j} - x_{j-1}), & \text{if } j > 1 \end{cases}$$

$$(4)$$

where  $r_j \in (0,1)$  is a sample-specific latent parameter determining how close  $x_j$  is to its lower boundary  $x_{j-1}$  (or 0 if j=1). We also account for the fact that a zircon can be younger then the sample it is assigned to, due to dating error or a later recrystallization of the zircon. To do that, we additionally sample a vector of identifiers  $\mathbf{I} = \{I_1, \dots, I_N\}$  that define which zircons are accounted for (I=1) or excluded (I=0) in determining the maximum boundary for the age of sample j:

$$\zeta_j = \min(z_n), \text{ for } n \in \{j, \dots, N\}, \text{ if } z_j = 1.$$
 (5)

## Priors and hyper-priors

We use a gamma prior for the vector of scale parameters of the Cauchy distributions  $\{s_1, \ldots, s_S\} \sim \Gamma(5, 5)$ , a normal prior centered on 0 on the vector of bias parameters  $\{\eta_1, \ldots, \eta_M\} \sim \mathcal{N}(0, 5)$  and a gamma prior on the vector of parameters  $\{\epsilon_1, \ldots, \epsilon_M\} \sim \mathcal{N}(0, 5)$ . We use a beta distribution as prior on the vector of parameters  $\{r_1, \ldots, r_S\} \sim \mathcal{B}(a, b)$  and consider the shape parameters a and b themselves as unknown parameters and assign them gamma distributed hyper-priors,  $a, b \sim \Gamma(1, 0.1)$ . Finally we use a Bernoulli distribution as prior on the indicators  $\{I_1, \ldots, I_N\}$  with parameter p = 0.99, thus assigning a 0.01 prior probability for a zircon to be younger than the sample it is assigned to.

## Parameter estimation

The model includes 2N + 2M + 2S + 2 parameters (z and I,  $\eta$  and  $\epsilon$ , s and r, a and b, respectively). All parameters are estimated through Metropolis-Hastings Markov chain Monte Carlo (MCMC). We use a sliding window proposal with reflection at the boundaries for r, normal kernel proposals for z, binomial proposals for I and sliding window for  $\eta$ . We use multiplier proposals on all other parameters since they only span the positive range.