

## Text S1. The algorithm of the two-phase search (TPS) method

The two-phase search (TPS) method is presented to systematically search all plausible parameter vectors satisfying a target or desired property without any biases in a mathematical model (Maeda and Kurata, IPSJ Trans Bioinfo, 2009). TPS consists of a random search (RS) and genetic algorithms (GAs). First, the RS explores a large parameter space without any biases to find a coarse solution showing a good fitness value. The resultant coarse solution is employed to generate the initial populations for the subsequent GAs. Second, after the initial population is created around the coarse solution vectors, GAs intensively search all plausible solutions. Let  $\mathbf{p} = (p_1, p_2, \dots, p_n)$  be the parameter vector and  $\text{Fitness}(\mathbf{p})$  be the fitness function that evaluates how well the parameters reproduce the experimental behaviors. A low value of  $\text{Fitness}(\mathbf{p})$  means high consistency between the simulated time course and experimental data. The end condition of the first phase is provided by:

$$\text{Fitness}(\mathbf{p}) \leq AEC, \dots\dots\dots (S1)$$

where  $AEC$  is the allowable error for the coarse solutions. The initial population for the GA is randomly generated around a coarse solution (**Figure S4**). The lower and upper bound vectors for the initial population are derived from the coarse solutions as follows:

$$\mathbf{L} = (L_1, L_2, \dots, L_n) \text{ and } \mathbf{U} = (U_1, U_2, \dots, U_n), \dots\dots\dots (S2)$$

where  $L_i$  and  $U_i$  are determined by using a coarse solution  $\mathbf{p}^c = (p_1^c, p_2^c, \dots, p_n^c)$ :

$$\begin{aligned} L_i &= p_i^c \cdot r^{-1} \\ U_i &= p_i^c \cdot r \end{aligned}, \dots\dots\dots (S3)$$

where  $r$  is the factor that expands the region. The end condition of the second phase for a final solution is given by:

$$\text{Fitness}(\mathbf{p}) \leq AE, \dots\dots\dots (S4)$$

where  $AE$  is the allowable error of plausible solution ( $AE \leq AEC$ ). In this study,  $AE$  and  $r$  are set to zero and  $10^{0.5}$ , respectively. The values of  $AEC$  are set to a value from 0.1 to 0.6, depending on models. We confirmed that the settings of  $AEC$  values do not affect our results in this study.

The pseudocode for TPS is shown in the below box, where  $n_p$  and  $n_g$  are the population size and the maximum generation for GAs, respectively.

$N \leftarrow 0$

**While**  $N < \text{Required number of solutions}$ :

**# First Phase**

  Randomly generate parameter vector  $\mathbf{p}$

**If**  $\text{Fitness}(\mathbf{p}) \leq AE$ :

    Store  $\mathbf{p}$  as a solution

$N \leftarrow N + 1$

**Else if**  $\text{Fitness}(\mathbf{p}) \leq AEC$ :

**# Second Phase**

    As an initial population, randomly create  $n_p - 1$  individuals between  $\mathbf{L}$  and  $\mathbf{U}$

    Add  $\mathbf{p}$  to the initial population

$\text{Generation} \leftarrow 1$

**While**  $\text{Generation} \leq n_g$  :

**If** the minimum value of fitness in population  $\leq AE$ :

        Store the individual that gives the minimum fitness as a solution

$N \leftarrow N + 1$

        Break

**End**

      Execute selection, crossover, and mutation

$\text{Generation} \leftarrow \text{Generation} + 1$

**End**

**End**

**End**