

1) Simulation data are generated, for multiple models of population histories which have known parameters ( $\theta$ ) and known Summary Statistics (S) :

Model 1: [  $\theta_1, \theta_2 \dots \theta_n, S_1, S_2 \dots S_n$  ]  
 Model 2: [  $\theta_1, \theta_2 \dots \theta_n, S_1, S_2 \dots S_n$  ]  
 Model 3: [  $\theta_1, \theta_2 \dots \theta_n, S_1, S_2 \dots S_n$  ]  
 ⋮  
 Model m: [  $\theta_1, \theta_2 \dots \theta_n, S_1, S_2 \dots S_n$  ]

2) A single simulation is selected, which has known parameters ( $\theta_{sim}$ ) and known Summary Statistics ( $S_{sim}$ ) :

[  $\theta_{sim}, \theta_{sim} \dots \theta_n, S_{1(sim)}, S_{2(sim)} \dots S_{n(sim)}$  ]

3) Standardized Sum of Squares is used to calculate the Euclidean distance between summary statistics of the selected simulation and that of every other simulations:

$$\sqrt{\sum \left[ \frac{S_n - S_n(obs)}{SD S_n} \right]^2}$$

4) Simulations are ranked based on Euclidean distance. Epsilon ( $\epsilon$ ) is determined as the closest fraction (0.0005%) of models:

Model x: [  $\theta_1, \theta_2 \dots \theta_n, S_1, S_2 \dots S_n$  ]  
 Model y: [  $\theta_1, \theta_2 \dots \theta_n, S_1, S_2 \dots S_n$  ]  
 $\epsilon$  -----  
 Model z: [  $\theta_1, \theta_2 \dots \theta_n, S_1, S_2 \dots S_n$  ]  
 ⋮  
 Model m: [  $\theta_1, \theta_2 \dots \theta_n, S_1, S_2 \dots S_n$  ]

5) The parameters ( $\theta_m$ ) of the selected models are scored against the known parameters of the single selected simulation ( $\theta_{sim}$ ) and reported as a percentage of all simulations which reported the exact same parameter value, as a measure of statistical support.

[  $\theta_{sim}, \theta_{sim} \dots \theta_n$  ]  
 Model 1: [  $\theta_1, \theta_2 \dots \theta_n$  ]  
 Model 2: [  $\theta_1, \theta_2 \dots \theta_n$  ]  
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 Score of represented parameters