1 SAMGEP E-step Algorithm

Our data, dat, is a data frame consisting of T_i timepoints for each of $i = \{1, ..., N\}$ patients. Features include:

- 1. ID: Patient IDs
- 2. T: Timepoints
- 3. Tlog: $\log T + 1$
- 4. H: Measure of a patient's healthcare utilization
- 5. Hlog: H+1
- 6. $\{X_1, X_2, \dots X_p\}$: p features

Other variables include:

- 1. likeModel: List consisting of
 - (a) β : $(p \times 8)$ -dimensional matrix of estimated GLS coefficients
 - (b) α : p-dimensional vector of estimated variance exponents
 - (c) σ : p-dimensional vector of estimated model standard deviations
 - (d) std.errors: $(8 \times 8 \times p)$ -dimensional array of GLS coefficient standard errors
 - (e) corrMat: $(p \times p)$ -dimensional feature correlation matrix

The E-step algorithm goes as follows:

Algorithm 1: SAMGEP E-step

```
1 for i = 1 to N patients do
          Compute \mu \mid \mathbf{Y} = \mathbf{0}, H_i, \mathbf{T}_i
          Compute \mu \mid \mathbf{Y} = \mathbf{1}, H_i, \mathbf{T}_i

Compute \sigma^2_{like} = \sigma^2 \times H^{2\alpha}

Compute \sigma^2_{prior}
 3
 4
 5
          Compute \sigma = \sqrt{\sigma_{prior}^2 + \sigma_{like}^2}
 6
          Compute \Sigma = [corrMat]\sigma^T \sigma for t = 1 to T_i timepoints do
 7
               if length(T_i) = 1 then
 8
                      Compute log prior of Y_{i,t}
 9
                      Compute log likelihood of \mathbf{X}_{i,t} \mid Y_{i,t}
10
                     Compute posterior of Y_{i,t} = 1 \mid \mathbf{X}_{i,t}
11
                else if t = 1 then
12
                      Compute log prior of \{Y_{i,t}, Y_{i,t+1}\}
13
                      Compute log likelihood of \mathbf{X}_{i,t}, \mathbf{X}_{i,t+1} \mid \{Y_{i,t}, Y_{i,t+1}\}
14
                     Compute posterior of Y_{i,t} = 1 \mid \mathbf{X}_{i,t}, \mathbf{X}_{i,t+1}
15
                else if t = T_i then
16
                      Compute log prior of \{Y_{i,t-1}, Y_{i,t}\}
17
                      Compute log likelihood of \mathbf{X}_{i,t-1}, \mathbf{X}_{i,t} \mid \{Y_{i,t-1}, Y_{i,t}\}
18
                     Compute posterior of Y_{i,t} = 1 \mid \mathbf{X}_{i,t-1}, \mathbf{X}_{i,t}
19
                else
20
                      Compute log prior of \{Y_{i,t-1}, Y_{i,t}, Y_{i,t+1}\}
21
                      Compute log likelihood of
22
                       \mathbf{X}_{i,t-1}, \mathbf{X}_{i,t}, \mathbf{X}_{i,t+1} \mid \{Y_{i,t-1}, Y_{i,t}, Y_{i,t+1}\}
                      Compute posterior of Y_{i,t} = 1 \mid \mathbf{X}_{i,t-1}, \mathbf{X}_{i,t}, \mathbf{X}_{i,t+1}
23
                Append estimated posterior and transition probabilities to output
\mathbf{24}
                 vector
```