

**Algorithm 1** On line EM with Optimal Control Learning

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0: Input: Observed data  $\{y_i^{(t)}, x_i^{(t)}\}$ , initial parameters  $\{B^0, \Sigma^0, \theta_i^0, \gamma_{ij}^0, \beta_c^0, \sigma_0^2\}$ , number of clusters
    $N_c, initialclusterpriors\{\pi_c\}_c$ .
0: for each time  $t$  do
0:   E-Step
0:   Compute the cluster assignment probabilities:  $p(c_i^{(t)}|x_i^{(t)}, y_i^{(t)}) = \frac{\pi_{c_i^{(t)}} p(y_i^{(t)}, x_i^{(t)}|c_i^{(t)})}{\sum_c \pi_c p(y_i^{(t)}, x_i^{(t)}|c)}$ 
0:   for each cluster  $c$  do:
0:     for each  $i$  do:
0:       Compute posterior mean:  $\mu_{z|x}^{(i,t)} = B_c^\top (B_c B_c^\top + \Sigma_c)^{-1} x_i^{(t)}$ 
0:       Compute posterior covariance:  $\Sigma_{z|x} = I - B_c^\top (B_c B_c^\top + \Sigma_c)^{-1} B_c$ 
0:     end for
0:   end for
0:   M-Step: Update model parameters
0:   for each cluster  $c$  do
0:     Update  $B$ :  $B_c^{new} = \left( \sum_{i,t} p(c_i^{(t)}|x_i^{(t)}, y_i^{(t)}) x_i^{(t)} \mu_{z|x}^{(i,t)\top} \right) \left( \sum_{i,t} p(c_i^{(t)}|x_i^{(t)}, y_i^{(t)}) (\Sigma_{z|x} + \mu_{z|x}^{(i,t)} \mu_{z|x}^{(i,t)\top}) \right)^{-1}$ 
0:     Update  $\Sigma$ :  $\Sigma_c^{new} = \frac{1}{\sum_{i,t} p(c_i^{(t)}|x_i^{(t)}, y_i^{(t)})} \sum_{i,t} p(c_i^{(t)}|x_i^{(t)}, y_i^{(t)}) \left( x_i^{(t)} x_i^{(t)\top} - B_c^{new} \mu_{z|x}^{(i,t)} x_i^{(t)\top} - x_i^{(t)} \mu_{z|x}^{(i,t)\top} B_c^{new\top} + B_c^{new} \Sigma_c^{new} B_c^{new\top} \right)$ 
0:     for each  $i$  do
0:       Update  $\theta_c$ :
0:        $\theta_c^{new} = \left( \sum_t p(c_i^{(t)}|x_i^{(t)}, y_i^{(t)}) x_i^{(t)} x_i^{(t)\top} \right)^{-1} \left( \sum_t p(c_i^{(t)}|x_i^{(t)}, y_i^{(t)}) x_i^{(t)} y_i^{(t)} \right)$ 
0:     end for
0:     for each  $i, j$  do
0:       Update  $\gamma_{ij}$ :
0:        $\gamma_{ij}^{new} = \left( \sum_t p(c_i^{(t)}|x_i^{(t)}, y_i^{(t)}) x_j^{(t)} x_j^{(t)\top} \right)^{-1} \left( \sum_t p(c_i^{(t)}|x_i^{(t)}, y_i^{(t)}) x_j^{(t)} (y_i^{(t)} - \theta_i^\top x_i^{(t)}) \right)$ 
0:     end for
0:     for each cluster  $c$  do
0:       Update  $\beta_c$ :
0:        $\beta_c^{new} = \frac{\sum_{i, c^*(i)=c} \sum_t p(c_i^{(t)}|x_i^{(t)}, y_i^{(t)}) u_c^{(t)} (y_i^{(t)} - \theta_i^{new\top} x_i^{(t)} - \sum_{j \neq i, c^*(j)=c^*(i)} \gamma_{ij}^{new\top} x_j^{(t)})}{\sum_{i, c^*(i)=c} \sum_t p(c_i^{(t)}|x_i^{(t)}, y_i^{(t)}) u_c^{(t)} u_c^{(t)}}$ 
0:     end for
0:     Update  $\sigma^2$ :
0:      $\sigma_c^{2,new} = \frac{1}{nT} \sum_{i,t} p(c_i^{(t)}|x_i^{(t)}, y_i^{(t)}) \left( y_i^{(t)} - \theta_i^{new\top} x_i^{(t)} - \sum_{j \neq i, c^*(j)=c^*(i)} \gamma_{ij}^{new\top} x_j^{(t)} - \beta_{c^*(i)}^{new} u_{c^*(i)}^{(t)} \right)^2$ 
0:   Final-Step: Optimal Control Learning
0:   for each cluster  $c$  do
0:      $u_c^{(t)} = \frac{\sum_{i, c^*(i)=c} (y_i^{(t)} - \theta_i^{new\top} x_i^{(t)} - \sum_{j \neq i, c^*(j)=c} \gamma_{ij}^{new\top} x_j^{(t)})}{\sum_{i, c^*(i)=c} \beta_{c^*(i)}^{new\top} \beta_{c^*(i)}^{new}}$ 
0:   end for

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