Algorithm 1: Selection2Sieve(S2S)

Input: 噪声数据 $\tilde{\mathcal{D}} = \{(\mathbf{x_i}, \tilde{\mathbf{y_i}})\}_{i=1}^n$ 可学习的特征提取网络 $\mathcal{V}(\cdot; \Theta_1)$ 、分类网络 $f(\cdot; \Theta_2)$ 、OOD 检测网络 $F^j(\cdot; \Theta_3^j) \, \forall \, j \in \mathcal{Y}$,总体网络参数为 θ . 权重 $\lambda_{con}, \lambda_e, \lambda_{id}$. 训练周期数 E. 预热网络周期数 selection_epochs,筛选开 始周期数 sieve_epochs. 学习率 η . 每个周期迭代次数 iters. 阈值 τ_{clean} ,

Output: 网络参数 $\theta = (\Theta_1, \Theta_2, \Theta_3)$.

```
1 for e from 1 to E do
             step = 0
             // 预热网络
            if e<selection epochs then
  3
                   while step < iters do
  4
                           采样 \mathcal{B} \subseteq \tilde{\mathcal{D}}
  5
                          计算 \mathcal{L}(\mathcal{B};\theta) = \mathcal{L}_{CE}(\mathcal{B};\Theta_1,\Theta_2)
  6
                          \theta = \theta - \eta \cdot \nabla_{\theta} \mathcal{L}(\mathcal{B}; \theta)
  7
                          step = step + 1
  8
                   \quad \text{end} \quad
  9
             else
10
                   // 划分干净样本集和无标记样本集 \mathcal{D}_c, \mathcal{D}_u
                   if e=selection_epochs then
11
                           划分 \mathcal{D}_c \subseteq \tilde{\mathcal{D}}; \ \mathcal{D}_u = \tilde{\mathcal{D}} - \mathcal{D}_c
12
                   end
13
                   // 筛选 ID 样本
                   if e>sieve epochs then
14
                           筛选 \mathcal{D}_{id} \subseteq \mathcal{D}_u; \mathcal{D}_{ood} = \mathcal{D}_u - \mathcal{D}_{id}
15
                   end
16
                    while step < iters do
17
                           采样 \mathcal{B} \subseteq \mathcal{D}_c; \mathcal{U} \subseteq \mathcal{D}_u
18
                          \mathcal{L}(\mathcal{B}, \mathcal{U}; \theta) = \mathcal{L}_{sup} + \lambda_{con} \mathcal{L}_{con} + \lambda_{e} \mathcal{L}_{e}
19
                          if e>sieve epochs then
20
                                 采样 \mathcal{B}_{id} \subseteq \mathcal{D}_{id}
21
                                 \mathcal{L}+=\lambda_{id}\mathcal{L}_{id}(\mathcal{B}_{id};\theta)
22
                          end
23
                          \theta = \theta - \eta \cdot \nabla_{\theta} \mathcal{L}(\mathcal{B}, \mathcal{B}_{id}, \mathcal{U}; \theta)
\mathbf{24}
                          step = step + 1
25
                   end
26
             end
27
28 end
```