

An Investigation of Cross-Project Learning in Online Just-In-Time Software Defect Prediction – Supplementary Material

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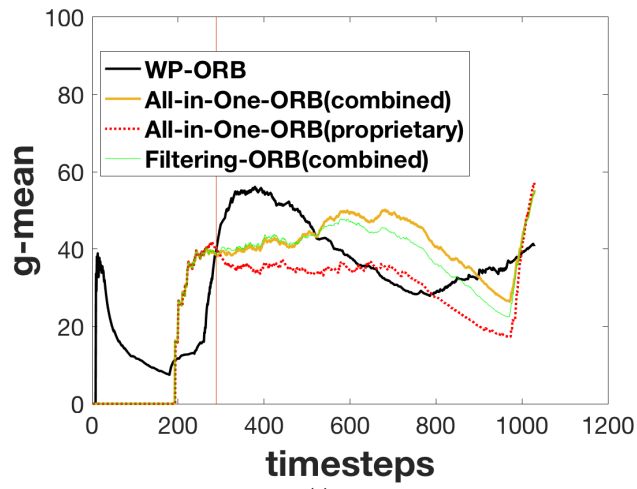
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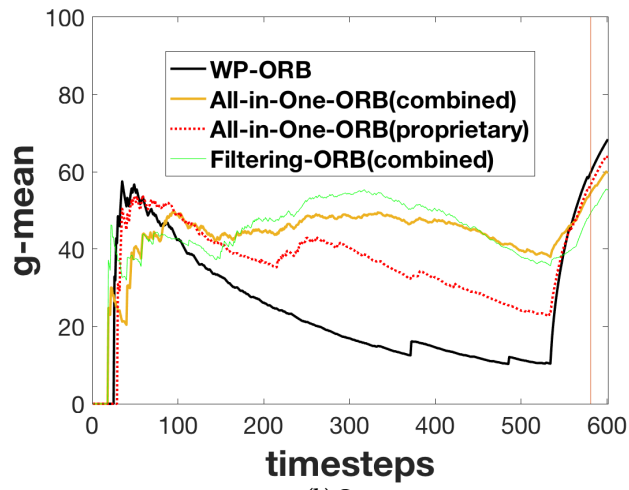
Algorithm 1 Ensemble approach

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1:  $S$  = stream of incoming changes from  $n$  projects,  $b$  = index  
   identifying the test project,  $w$  = waiting period  
2: initialise ensemble model  $M$  consisting of  $n$  models  $\{m_1, m_2, \dots,$   
    $\dots, m_n\}$   
3: for each incoming change  $x_p^t \in S$  do //  $x_p^t$  is a change arriving  
   from project  $p$  at timestamp  $t$   
4:   if  $p = b$  then  
5:      $\hat{Y}\text{-List} \leftarrow$  get prediction  $\hat{y}_i$  from each model in  $M$   
6:      $\hat{y} =$  mean of stored prediction results in  $\hat{Y}\text{-List}$   
7:   end if  
8:   store  $x_p^t$  in a queue  $WFL\text{-}Q_p$  for project  $p$  //  $WFL\text{-}Q_p$  is  
   the queue of incoming changes of project  $p$  waiting to be used  
   for trained  
9:   for each model  $m_p$  in  $M$  do  
10:    for each change  $q^i$  in  $WFL\text{-}Q_p$  do  
11:     if a defect was linked to  $q^i$  at a timestamp  $\leq t$  then  
12:       create defect-inducing training_example for  $q^i$   
13:       train( $m_p$ , training_example)  
14:       remove  $q^i$  from  $WFL\text{-}Q_p$   
15:     else  
16:       create a clean training_example for  $q^i$   
17:       train( $m_p$ , training_example)  
18:       remove  $q^i$  from  $WFL\text{-}Q$   
19:       store training_example in  $CL_p\text{-}H$  //  $CL_p\text{-}H$  is a  
   hash of clean training examples for project  $p$   
20:     end if  
21:   end for  
22: end for  
23: if a defect was linked to a training_example in  $CL_p\text{-}H$  before  
   time  $t$  then  
24:   Swap the label of training_example to defect-inducing  
25:   train( $m_p$ , training_example)  
26:   remove  $h$  from  $CL_p\text{-}H$   
27: end if  
28: end for
```

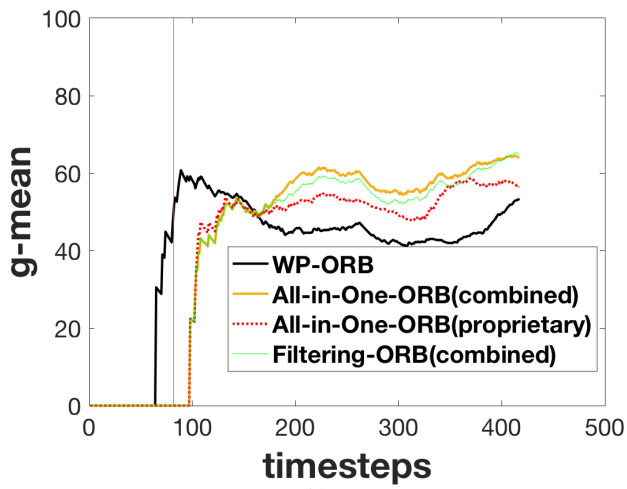
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(a) C1



(b) C2



(c) C3

Figure 1: G-Mean for proprietary datasets through time using ORB. The vertical red bar indicates the last time step of the initial phase of the project.