

## APPENDIX B BASELINE APPROACHES

We compare iFit with the following eight baseline approaches in the literature [14]:

- **h-index** [9]: It calculates the  $h$ -index of a class in the software network as its importance. Generally, if a class has  $N_p$  neighbors that are ranked in descending order according to their received degrees, then the class has  $h$ -index  $h$  if each top- $h$  neighbor receives a degree  $\geq h$ , and the other  $(N_p - h)$  neighbors receive a degree  $< h$  each. The top- $h$  neighbors form the Hirsch core.
- **a-index** [9]: It calculates the  $a$ -index of a class in the software network as its importance. Formally,  $a = \frac{1}{h} \sum_{j=1}^h d_j$ , where  $a$  is the  $a$ -index value,  $h$  is the  $h$ -index value,  $d_j$  is the degree of class  $j$  in the Hirsch core (cf.  $h$ -index), and the  $h$  classes in the Hirsch core are ranked in descending order according to their received degrees.
- **k-core** [10]: It applies a  $k$ -core decomposition to calculate the *coreness* of a class in the software network as its importance. The  $k$ -core of a network is defined as the largest subgraph, where each node has a degree  $\geq k$ . Then a node has coreness  $k$  if it belongs to the  $k$ -core but not to the  $(k + 1)$ -core.
- **PageRank** [19]: It applies a PageRank algorithm to calculate the PageRank value of a class in the software network as its importance. The PageRank value of class  $u$  (i.e.,  $PR(u)$ ) is computed by  $PR(u) = \frac{1-d}{m} + d \times \sum_{v \in IN(u)} \frac{PR(v) \times w(v,u)}{wOutDeg(v)}$ , where  $IN(u)$  is the in-neighbor of class  $u$ ,  $w(v,u)$  is the weight on the link from  $v$  to  $u$ ,  $wOutDeg(v)$  is the weighted out-degree of class  $v$ , and  $m$  is the number of classes in the software network.  $d$  is the damping factor which is fixed to 0.85.
- **PageRank\_BR** [19]: It improves the PageRank by considering the back recommendation. Specifically, if there is a link from classes **A** to **B**, the authors treated it as a forward recommendation of **A** on **B**. For each forward recommendation, they added a back recommendation (i.e., a link from **B** to **A**) with the weight on the link being a fraction  $F$  of the weight on the corresponding forward recommendation. In this work, we used the best value of  $F$  reported in [19], i.e.,  $F=0.5$ .
- **ICOOK** [11]: It calculates the *generalized coreness* of a class in the software network as its importance. The *generalized coreness* is computed on weighted directed networks and based on generalized degrees.
- **ElementRank** [12]: It takes a similar way as PageRank to compute the PageRank value of classes. The only difference is that  $\frac{1-d}{m}$  is replaced by  $\frac{(1-d) \times InLinks(u)}{\sum_{k=1}^N InLinks(k)}$ , where  $InLinks(u)$  is the sum of weights over the incoming links of node  $u$ .
- **Pride** [14]: It takes a similar way as PageRank to com-

pute the PageRank value of classes. The only difference is that  $\frac{1-d}{m}$  is replaced by  $\frac{(1-d) \times wDeg(u) \times Deg(u)}{wDegSum \times DegSum}$ , where  $wDeg(v)$  is the weighted degree of class  $v$ ,  $Deg(v)$  is the degree of class  $v$ ,  $wDegSum$  is the sum of the weighted degree of all classes, and  $DegSum$  is the sum of the degree of all classes.