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Topo2vec: A novel node embedding generation based on the topology of network for the link prediction (supplementary file)

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1 ALGORITHMS

21: end for=0

In the proposed method section (Section 3 of the original manuscript), we have described two algorithms that are used in our proposed method, Topo2vec. Here, for references to the used algorithms, we have a more elaborate explanations of the same.

Algorithm 1 To explore the initial context subgraph of each individual nodes in Graph G, $G = \{V, E\}$,

```
1: The input: Adjlist (Adjacency list data structure of
    the Graph.) Initialize context\_nodes \leftarrow
                                                          NULL,
    node\_number = ||V||.
 2: for CurrNode \leftarrow (0 : node\_number - 1) do
      visited[0:node\_number-1] \leftarrow 0
      visited[CurrNode] = 1
 4:
      templist = Adjlist[CurrNode];
 5:
      if templist==NULL then
 7:
         continue;
      else
 8:
         size \leftarrow length(templist), Score[0:size-1] \leftarrow 0
 9:
         for i \leftarrow (0 : \text{size-1}) do
10:
           Score[i] = NA_{CurrNode}^{templist(i)}
11:
12:
         [Score, ind] = Sort(Score);
13:
         templist \leftarrow templist[ind];
14:
         templist(i) = [\exists i, Score[i] > \tau];
15:
16:
17:
      for k \leftarrow 0: Average_degree do
         templist = SEARCH(templist, visited, CurrNode)
18:
19:
      context_nodes=[context_nodes; templist]
```

Algorithm 2 To expand the context sub-graph of a source node in a network.

```
0: procedure SEARCH(templist, visited, CurrNode)
 1: Tempauglist \leftarrow NULL
 2: Utemplist \leftarrow Uniquelist(templist)
 3: for i \leftarrow (0 : length(Utemplist) - 1) do
      if (visited[Utemplist[i]]==0) then
         visited[Utemplist[i]] \leftarrow 1
 5:
         neighborl = Adjlist[Utemplist[i]]
 6:
         for j \leftarrow (0 : |neighbor l| -1) do
 7:
            Score[i] \leftarrow NA_{CurrNode}^{neighborl(j)}
 8:
         end for
 9:
         [Score, indx] = Sort(Score)
10:
         neighborl = neighborl[indx]
11:
12:
         neighborl(i) = [\exists i, Score[i] > \tau]
         Auglist \leftarrow neighborl
13:
         for k \leftarrow (0: |(Auglist)|-1) do
14:
            Score[k] \leftarrow SA_{CurrNode}^{Auglist(k)}
15:
         end for
16:
         [Score, indx] = Sort(Score)
17:
         Auglist = Auglist[indx]
18:
         Auglist(k) = [\exists k, Score[k] \ge \tau]
19:
20:
         Tempauglist \leftarrow [Tempauglist, Auglist]
21:
       else
         Continue:
22:
       end if
23:
24: end for
\textbf{25: } templist \leftarrow [templist, Tempauglist]
26: return templist
26: end procedure=0
```

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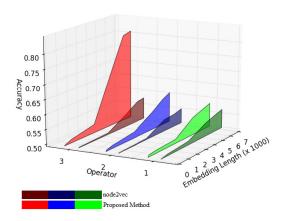


Fig. 1: Performance comparison of different pairwise feature representation operators defined in Table 1 in the manuscript, through clustering. Results are shown using PPI data (4 fold linked information) with the proposed method and node2vec. Index in the operator-axis represents (1) Hadamard, (2) SimKron_Average and (3) SimKron_Hadamard operators respectively.