

IMP Project

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1. Preliminaries

Influence maximization(IMP) is the problem of finding a small subset of nodes(seed nodes) in a social network that could maximize the spread of influence. The influence spread is the expected number of nodes that are influenced by the nodes in the seed set in a cascade manner. And I will deeply study this problem into two parts. The first part is to find an influence value on specific seeds into a network graph, the second part is to find the given number seed set that best spread seed in given network graph.

1.1. Software

This project is implemented in Python using IDE Py-Charm. The libraries being used include Numpy, multiprocessing, time, sys, getopt and Random.

1.2. Algorithm

The Algorithm being used in this project including climbing the IC model, LT model and the ISE for calculate the influence value in part 1. Then the part 2 is the Greedy Algorithm that CELF for finding small seed number and get a heuristic algorithm for find big seed number (larger than 10). [1]

2. Methodology

For ISE part is write for IC and LT model that do the calculate the influence value on specific seeds into a network graph. Then is the second part to find an influence value on specific seeds into a network graph, the second part is to find the given number seed set that best spread seed in given network graph. IMP problem can find best result by greedy algorithm, but its time is too long for large graph. For CELF it can do for fast enough for a limit number seeds (less than 10) but if seed number larger the data need refresh is more and more [1]. So for seeds number larger than 10 use another heuristic algorithm to do calculate for a better cost performance.

2.1. Representation

Firstly, global variables defined ahead of the file

- **vertices**: the number of vertices.
- **edges**: the number of edges.
- **seeds**: the seeds given in ISE.
- **seeds_num**: the number of seeds in IMP.

Then the class Graph

- **Graph**: the class for main data need use.
 - **node**: a set for all node attend.
 - **edge**: a list to remember all edge for the graph.
 - **in_edge**: a list to remember all in edge for the graph.
 - **weight**: a dict to remember all edge distance for the graph.

Other use data in the algorithm

- **file_name**: the file name get from command line.
- **time_limit**: the time limit get from command line.
- **model**: the model need to use in this calculate.
- **seed_path**: the seeds path name in ISE.
- **seed_num**: the seeds number in IMP.

2.2. Architecture

Here are the important functions in Python file ISE with pseudocode format.

- **IC_model**: the calculate model by the weight for edges only.
- **LT_model**: the calculate model by the weight for edges and the weight that has active points.
- **spread_check**: use the prescribed model to calculate the end result.

Then is other important functions in Python file IMP with pseudocode format.

- **CELF**: the Greedy algorithm to find good seeds in small number.
- **fast_find**: the Heuristic algorithm to find good seeds in big number.

2.3. Detail of Algorithm

Here show the psudocode main use that IC_model , LT_model , spread_check ,CELF and fast_find .

Algorithm 1 IC_model:the calculate modle by the weight for edges only .

Input: *Graph* graph ,*set* seed

Output: *int* result

```

1: function IC_MODEL(graph, seed)
2:   work  $\leftarrow$  seed
3:   all  $\leftarrow$  seed
4:   result  $\leftarrow$  len(seed)
5:   while len(work) > 0 do
6:     new_work  $\leftarrow$  set()
7:     for node in work do
8:       for near_node in graph.edge[node] do
9:         if random() larger then edge_weight then
10:          if near_node not in all then
11:            all.add(near_node)
12:            new_work.add(near_node)
13:          end if
14:        end if
15:      result += len(new_work)
16:      work = new_work
17:    end for
18:  end for
19:  end while
20:  return result
21: end function

```

Algorithm 2 spread_check:use the prescribed model to calculate the end result.

Input: *Graph* graph ,*set* seed,*String*model

Output: *float* result

```

1: function SPREAD_CHECK(graph, seed, model)
2:   result  $\leftarrow$  0.0
3:   if model == "IC" then
4:     for i in range(10000) do
5:       sum += IC_model(graph, seed)
6:     end for
7:   end if
8:   if model == "LT" then
9:     for i in range(10000) do
10:      sum += LT_model(graph, seed)
11:    end for
12:   end if
13:   return result/10000
14: end function

```

Algorithm 3 LT_model:the calculate modle by the weight for edges and the weight that has active points .

Input: *Graph* graph ,*set* seed

Output: *int* result

```

1: function LT_MODEL(graph, seed)
2:   work  $\leftarrow$  seed
3:   all  $\leftarrow$  seed
4:   result  $\leftarrow$  len(seed)
5:   ran  $\leftarrow$  dict()
6:   while len(work) > 0 do
7:     new_work  $\leftarrow$  set()
8:     for node in work do
9:       for near_node in graph.edge[node] do
10:        if near_node not in all then
11:          tol_weight take all connect edge has
          added weight
12:          if near_node not in ran then
13:            ran[near_node]  $\leftarrow$  random()
14:          end if
15:          if tol_weight > ran[near_node] then
16:            all.add(near_node)
17:            new_work.add(near_node)
18:          end if
19:        end if
20:      end for
21:      result += len(new_work)
22:      work = new_work
23:    end for
24:  return result
25:

```

Algorithm 4 CELF:the Greedy algorithm to find good seeds in small number .

Input: *Graph* graph ,*int* seed_num,*String*model

Output: *set* ans

```

1: function CELF(graph, seed_num, model)
2:   ans  $\leftarrow$  set()
3:   ans_list  $\leftarrow$  dict()
4:   for node in graph.node do
5:     ans.add(node)
6:     ans_list[node] = spread_check(graph, ans, model)
7:     ans.remove(node)
8:   end for
9:   new_add = max(ans_list, key=ans_list.get)
10:  ans.add(new_add)
11:  while len(ans) < seed_num do
12:    new_add = max(ans_list, key=ans_list.get)
13:    ans_list[new_add] = spread_check(graph, new_add — ans, model)
14:    new_add_r = max(ans_list, key=ans_list.get)
15:    if new_add == new_add_r then
16:      ans.add(new_add)
17:      ans_list.pop(new_add)
18:    end if
19:  end while
20:  return ans
21: end function = 0

```

Algorithm 5 fast_find:the Heuristic algorithm to find good seeds in big number .

Input: *Graph* graph ,*int* seed_num,*String*model

Output: *set* ans

```

1: function FAST_FIND(graph, seed_num, model)
2:   point  $\leftarrow$  0
3:   ans  $\leftarrow$  set()
4:   ans_list  $\leftarrow$  dict()
5:   has_add  $\leftarrow$  set()
6:   for node in graph.node do
7:     ans.add(node)
8:     ans_list[node]=spread_check(graph,ans,model)
9:     ans.remove(node)
10:  end for
11:  ans_list=sorted(ans_list)
12:  while len(ans)<seed_num do
13:    new_add= ans_list[point]
14:    if new_add not in has_add then
15:      ans.add(new_add)
16:      has_add.add(new_add)
17:
18:      do the model IC or LT use ans
19:      when make one node active
20:      has_add.add(node)
21:
22:      point+=1
23:    end if
24:    if new_add in has_add then
25:      point+=1
26:    end if
27:  end while
28:  return ans
29: end function

```

For the ISE just use the one special model and the spread_check is enough

For the IMP use the one special model , the spread_chec and choice method by the need seed_num if seed_num smaller then 10 use CELF is enough. If larger then 10 use the fast_find method that say in this article.

3. Empirical Verification

In Empirical Verification,I used two data sets to test my ISE and IMP algorithms separately. In IMP, I will use the basic CELF and the fast algorithm (design by myself) to compare the results and time, and discuss the use domain by the result and time use.

3.1. Design

For ISE is use the special model to calculate for enough times .

The IMP problem is a NP hard problem and under the IC model and LT model its computation is P hard. Because we are only given a little network data so the design of the

testing plan become much more harder. We should consider the difference of model we use and the key factor should be the model, strategy, size of data.

3.2. Data and data structure

In this project, the dataset I use is network.txt and NetHEPT.txt. These two datasets are one small dataset with 62 vertices and 159 edges, and the other is a large dataset with 15233 Vertices and 58891 edges.

3.3. Performance and result

ISE

data set	run time	result
Network-seeds-IC	1.166s	5.02
Network-seeds-LT	1.355s	5.0226
Network-seeds2-IC	2.110s	30.4695
Network-seeds2-LT	3.480s	36.9932
NetHEPT-50seeds-IC	25.654s	1002.5355
NetHEPT-50seeds-LT	57.781s	1268.179
NetHEPT-5seeds-IC	8.569s	276.2911
NetHEPT-5seeds-LT	35.646s	337.4792

For ISE the time relate the graph size and seed number the result has shown in the table

IMP

data set	run time	result
Network-5-IC	1.306s	30.72
Network-5-LT	1.289s	37.54
NetHEPT-5-IC	9.036s	322.89
NetHEPT-5-LT	18.316s	392.975
NetHEPT-50-IC-bonus	55.618s	1161.6776
NetHEPT-50-LT-bonus	56.879s	1522.8746

Use CELF for small seed number can get a good result in limit time but if do for 50 seeds in NetHEPT will take for about 400s but for fast_find one time just about 3 s and the result has shown in the table.

3.4. Hype-parameters

In ISE, my Hype-parameter is the average calculation parameter, and I set it to 10000 in order to get a stable solution.

In IMP, I set the average calculation parameter K=1000 in the basic greedy algorithm and CELF. However, when the value of seed_size is greater than 10 or the fixed point and number of edges of the graph are too large, I set my average calculation parameter time = 100 in fast_find. The purpose is to be able to draw results within a limited time .

3.5. Analysis

For the Influence maximization(IMP) project is like the carp problem and not like carp. The IMP is NP hard problem is large then CARP that both can not check all result in a limited time . But for CARP use get initial solver and do for better result. But IMP is for get result in add result add node one by one . So for fast_find result the problem is to take the right heuristic .For IMP has many good algorithm like Greedy, CELF_Greedy, Lv_CELF_Greedy, CELF++_Greedy and Lv_CELF++_Greedy [2]. For many codes try and understand then write by self thought is the best project do in my understand .

Acknowledgments

I would like to thank my friends Peijun Ruan who together discuss the IMP's frame . Read and discuss the paper which the hard to understand part . Last I would like to thank SA who will check my codes and reports.

References

- [1] Li, J. Fan, Y. Wang and K. Tan, "Influence Maximization on Social Graphs: A Survey", IEEE Transactions on Knowledge and Data Engineering, vol. 30, no. 10, pp. 1852-1872, 2018.
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