

Data Mining Homework - 5

K-way Graph Partitioning Using JaBeJa

Group – 7

Xing Zeng <xingzeng@kth.se> Sevkett Melih Zenciroglu <smzen@kth.se>

1. Task1

In task1, we implement the JabeJa algorithm in paper[1]. In method SampleAndSwap, we use hybrid heuristic for node selection and swap the node if the partner returned by method FindPartner is not null. Method FindPartner shows how the partner is selected.

To avoid becoming stuck in a local optimum, in task1, we use simulated Annealing (SA) technique] which introduces a temperature (T) and decreases it over time linearly as shown saCoolDown() method.

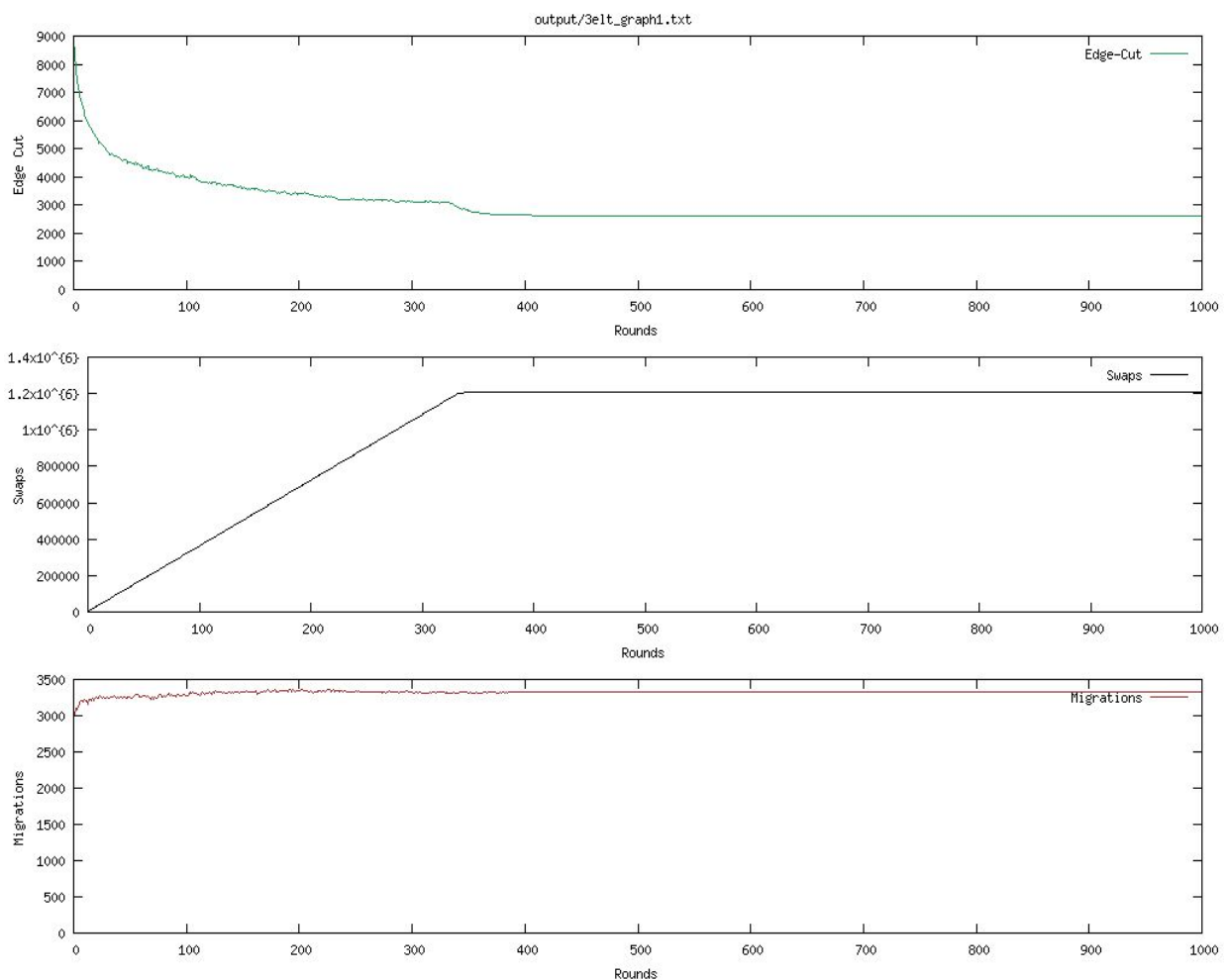
```
97 public Node findPartner(int nodeId, Integer[] nodes){
98
99     Node nodep = entireGraph.get(nodeId); //HashMap<Integer, Node>
100
101     Node bestPartner = null;
102     double highestBenefit = 0;
103     double alpha = config.getAlpha();
104
105     // TODO
106     for(Integer nodeqId : nodes){
107         Node nodeq = entireGraph.get(nodeqId);
108         int d_pp = getDegree(nodep, nodep.getColor());
109         int d_qq = getDegree(nodeq, nodeq.getColor());
110         double old_ = Math.pow(d_pp, alpha) + Math.pow(d_qq, alpha);
111         int d_pq = getDegree(nodep, nodeq.getColor());
112         int d_qp = getDegree(nodeq, nodep.getColor());
113         double new_ = Math.pow(d_pq, alpha) + Math.pow(d_qp, alpha);
114         if(new_ * T > old_ && new_ > highestBenefit){
115             bestPartner = nodeq;
116             highestBenefit = new_;
117         }
118     }
119     return bestPartner;
120 }
```

```

49      * Simulated analealing cooling function
50      */
51      private void saCoolDown(){
52          // TODO for second task
53          if (T > 1)
54              T -= config.getDelta();
55          if (T < 1)
56              T = 1;
57      }

```

The result of task1 is as follows,



2. Task2

Task2, we tweak different JaBeJa to analyze how changing the simulated annealing parameters and the acceptance probability function affects the performance of Ja-Be-Ja.

- Change parameters

We set $T = 1$, and try different delta 0.003 and 0.9. And we use acceptance probability functions $e^{\frac{\text{new}-\text{old}}{T}}$ in [2].

```

49  /**
50   * Simulated analealing cooling function
51   */
52  private void saCoolDown(){
53      // TODO for second task
54      // if (T > 1)
55      //     T -= config.getDelta();
56      // if (T < 1)
57      //     T = 1;
58      // T *= 0.9;
59      T *= config.getDelta();
60  }

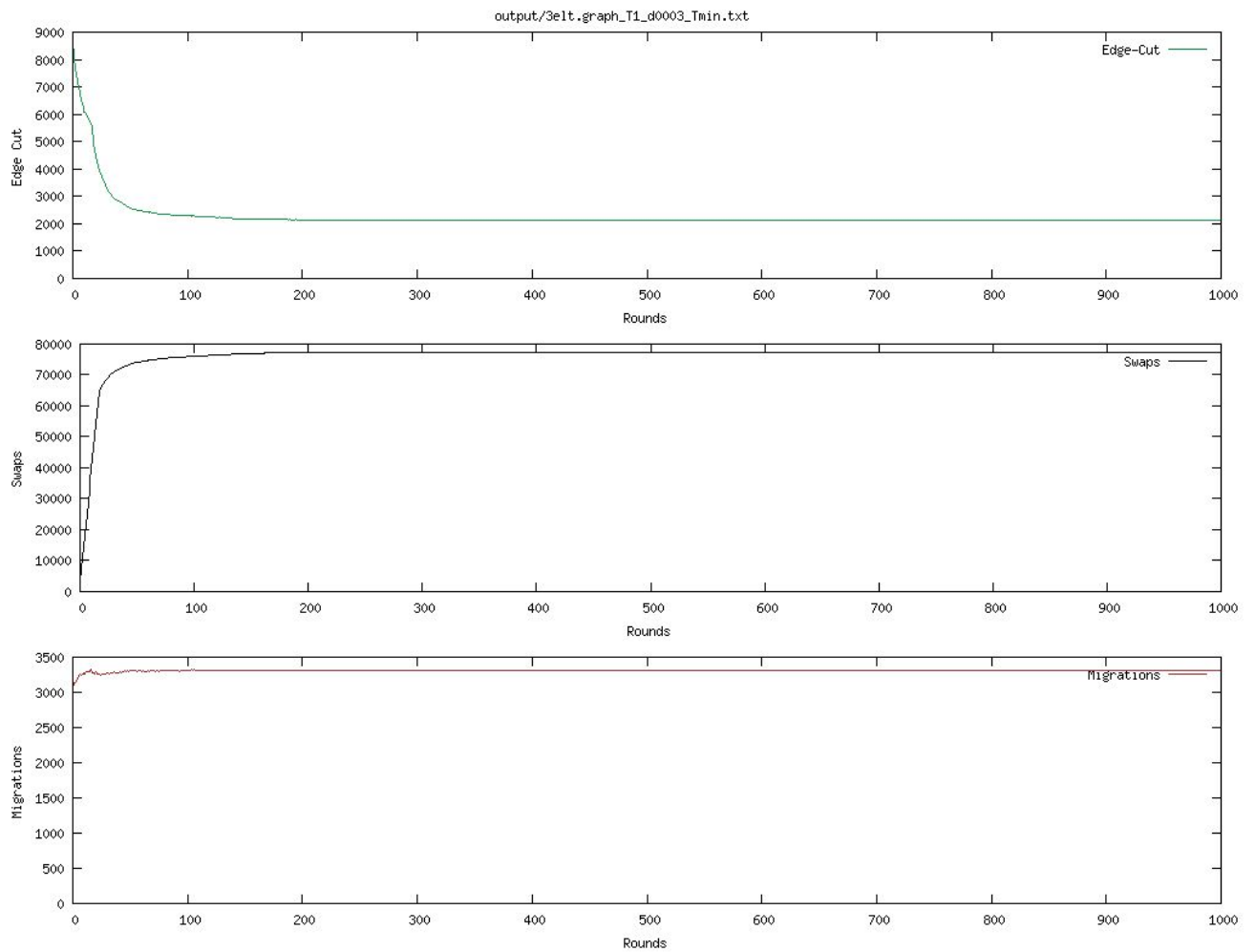
```

```

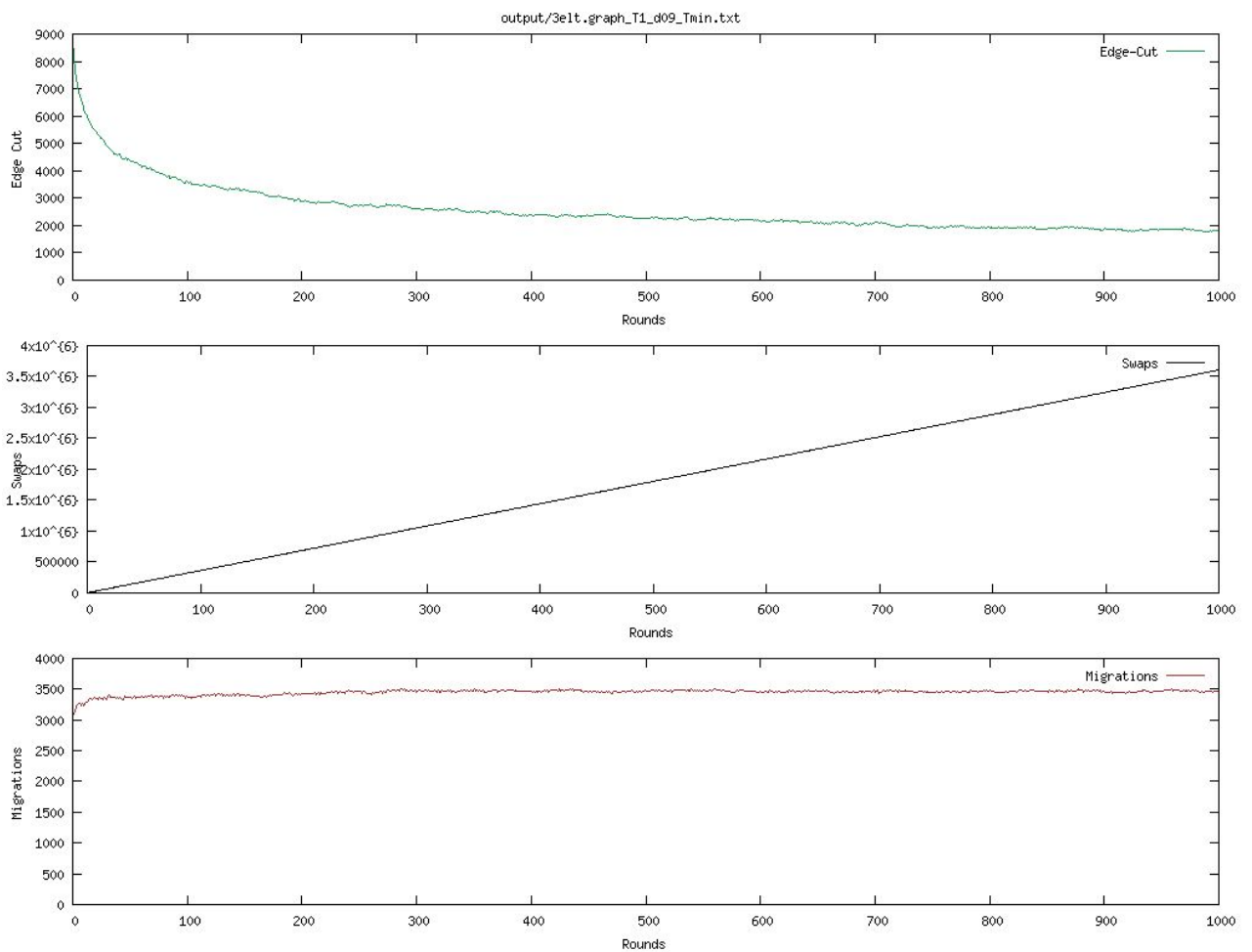
106 // TODO
107 for(Integer nodeqId : nodes){
108     Node nodeq = entireGraph.get(nodeqId);
109     int d_pp = getDegree(nodep, nodep.getColor());
110     int d_qq = getDegree(nodeq, nodeq.getColor());
111     double old_ = Math.pow(d_pp, alpha) + Math.pow(d_qq, alpha);
112     int d_pq = getDegree(nodep, nodeq.getColor());
113     int d_qp = getDegree(nodeq, nodep.getColor());
114     double new_ = Math.pow(d_pq, alpha) + Math.pow(d_qp, alpha);
115     double loss = new_ - old_;
116     double ap = 0;
117     // if new_solution is better than old one
118     if(loss > highestBenefit){
119         ap = 1;
120     }
121     else{
122         //ap = Math.exp((newBenefit- highestBenefit)/T);
123         //acceptance probability in paper
124         ap = Math.pow(Math.E, (loss-highestBenefit)/T);
125
126         // self design
127         // ap = 1/Math.pow((loss-highestBenefit)/T,2);
128     }
129     if(ap > Math.random()){
130         bestPartner = nodeq;
131         highestBenefit = loss;
132     }
133 }
134 return bestPartner;
135 }
136

```

The result of $T=1$, $\Delta = 0.003$ is as follows,



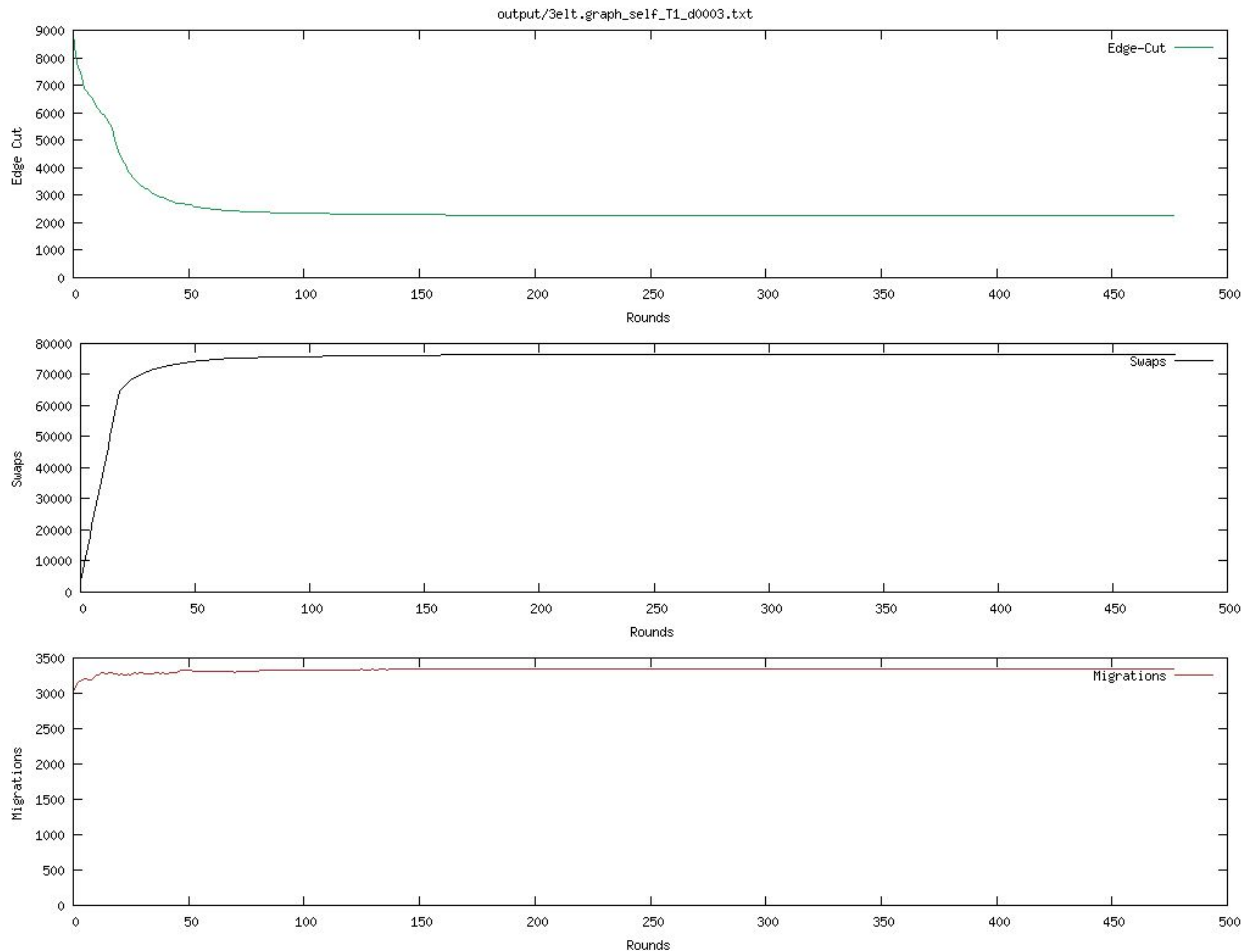
The result of $T=1$, $\Delta = 0.9$



- Change the acceptance probability

As shown in annotation in above screenshot (line 127), we define another acceptance probability as $\frac{1}{e^{\frac{(new-old)^2}{T^2}}}$

The result of T=1, Delta = 0.003 is as follows,



3. Analysis

From the above result screenshots, we can notice that

- 1) The performance of simulated annealing approach in Task2 is better than Task1 since it converges more quickly and the cutting edge is 2141 less than 2604 in task1.
- 2) From task2, once no more bad swaps are allowed, then Ja-Be-Ja converges to an edge cut rapidly and the edge cut does not change over time. And when T is set initially equals 1, the smaller the delta is, the less the swapping is and the quicker the converge rate is. While if the delta is larger, the cutting edge is less.

4. Reference

- [1] F. Rahimian, A. H. Payberah, S. Girdzijauskas, M. Jelasity and S. Haridi, JA-BE-JA: A Distributed Algorithm for Balanced Graph Partitioning Preview the document, SASO2013, pp. 51-60.
- [2] <http://katrinaeg.com/simulated-annealing.html>