

Weekly Report 2

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What I did

I did the rest of the assigned reading: Lectures 10 and 11 of ORIE 6334 from Fall 2019, and Trevisan 2012 from the beginning to section 4.

My Thoughts

This was a much harder topic to understand than last week's (semidefinite programming). I had to learn some topics mentioned in the paper on my own (such as spectral partitioning, incidence matrix of a graph, induced subgraphs, etc.).

From my understanding after reading the materials several times, this is my pseudocode of the algorithm from the beginning of section 4 (page 9 of the PDF), and thus the meat of the paper:

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function Recursive-Spectral-Cut(V,E, $\delta$ ){

  y = 2-Thresholds Spectral Cut with  $\delta$ ;

  M = number of weighted number of edges (i,j) such that at
      least one of y_i and y_j is not 0 (i.e. weighted number of
      edges incident to  $S \subseteq V$ );

  C = the weighted number of cut edges (i,j) such that y_i and
      y_j are both not 0 and have opposite signs (i.e. one
      vertex is in L and the other is in R);

  X = weighted number of cross edges (i,j) such that exactly one
      of y_i and y_j is 0 (i.e. one vertex is in S and the other
      is not)

  if (C+X/2 <= M/2){
    use a greedy algorithm to find a partition of V that cuts at
    least |E|/2 edges (coin flipping could work?), and return it
  }
  else {
    L = {i such that y_i == -1};
    R = {i such that y_i == +1};
    V_prime = {i such that y_i == 0};

    G_prime(V_prime, E_prime) = induced subgraph of V_prime;
    (V_1, V_2) = Recursive-Spectral-Cut(V_prime, E_prime,  $\delta$ );

    return bigger_cut((V_1  $\cup$  L, V_2  $\cup$  R), (V_1  $\cup$  R, V_2  $\cup$  L));
  }
}

```

When the paper said, "Run the algorithm of Theorem 1 with accuracy parameter δ " (section 4), I'm assuming it's the algorithm given 2-Thresholds Spectral Cut (abbreviated 2TSC) given on pages 1775-1776 (pages 7-8 in the PDF). 2TSC seems mostly straightforward, but I'm having trouble understanding how to get the x and where the δ comes into play. To the best of my understanding, δ is not an input to any program, but is rather a means of measuring if a solution y satisfies – when plugged into (3) – an upper bound. I took note of "the optimization problem in (4) is the problem of computing the smallest eigenvalue of $D^{-\frac{1}{2}}AD^{-\frac{1}{2}}$ " (1775). So, does this mean x is the eigenvector corresponding to the smallest eigenvalue of the matrix $D^{-\frac{1}{2}}AD^{-\frac{1}{2}}$? I really can't tell how to get x used in 2TSC.

What I will do next

I will start learning Julia and hopefully start implementing the algorithms, starting with naïve random algorithms and simple, greedy algorithms, then semidefinite programming, then Trevisan's (though I have to understand his algorithm completely first).