

# Approximation Algorithms Lecture 24

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## 1 Recap

Completion of primal dual algorithms for steiner forest.

## 2 Content

### 2.1 Maximum cut in a graph

#### Definition 1

**Cut:** Given an undirected and unweighted graph  $G = (V, E)$ , a *cut* is a partition of the vertex set into two sets  $S$  and  $V \setminus S$ . The edges of the cut are the edges whose endpoints are in different parts of this partition.

The size/weight of the cut is the number (or total weight) of edges in the cut.

The maximum cut problem is to find a cut of the maximum size.

In a bipartite graph, the number of edges is the actual max cut. For a general graph, the max cut is at most the number of edges in the graph.

#### Local search algorithm for max cut

1. Take an arbitrary partition of the vertex set  $S, V \setminus S$ .
  - If there exists a vertex such that moving it to the other partition increases the size of the cut, then we make the change.
  - Repeat this until no such change is possible.
2. Return the final cut  $(S, V \setminus S)$ .

The operation is also called a *local move*, and the final solution is called a locally optimal solution.

#### Claim 0.1

This algorithm returns a cut of size at least half the number of edges in the graph which is at least  $\frac{\text{OPT}}{2}$ .

*Proof.* Consider a locally optimum solution  $(S^*, V \setminus S^*)$ . Consider any vertex  $v \in S^*$ , then we have  $\deg_{S^*}(v) \leq \deg_{V \setminus S^*}(v)$ . This is true for all vertices in  $S^*$ . Summing for all vertices in  $S^*$ , we get the LHS as 2 times the number of edges in  $S^*$ , and the RHS as number of edges in the cut  $(S^*, V \setminus S^*)$ . By summing the corresponding inequality for all vertices in  $V \setminus S^*$ , we get that the number of edges in the cut is at least twice the number of edges in  $V \setminus S^*$ . Now adding twice the number of edges in  $V \setminus S^*$  to the inequality found by adding these two, we get  $2|E| \leq 4$  times the number of edges in the cut, and we are done.  $\square$

This can be implemented in  $O(m \cdot (n + \max \deg(v)))$ .

## 2.2 Balanced cuts

A cut which has equal number of vertices in each set of the partition.

A relaxed version is where no part has number of vertices more than twice that of another part. For cuts, this implies that each part has at least  $1/3$ rd and at most  $2/3$ rd the vertices of the graph.

Suppose we want to find a balanced max-cut (exactly half the vertices on each side). How to modify the previous algorithm?

1. Pick an arbitrary balanced cut  $(S, V \setminus S)$ .
  - If  $\exists u \in S, v \in V \setminus S$  such that swapping the pair increases the size of the cut, then we do so.
  - Repeat while such a pair exists.
2. Return the locally optimal solution found.

Let's analyse this algorithm.

Let green edges be edges across sets starting from either of  $u, v$  but not both, and let blue ones be the one within the sets starting from either of  $u, v$ . When will we make the swap? We make the swap if the number of blue edges is greater than the number of green edges.

Suppose  $(S^* = S, V \setminus S^*)$  is a locally optimal solution. We pair vertices in  $S^*$  with the vertices in  $V \setminus S^* = T$  in an arbitrary manner.

Then it holds that  $\deg_T(u_i) + \deg_S(v_i) \geq \deg_S(u_i) + \deg_T(v_i)$  where we also include the red edges in the LHS.

Summing this inequality, we get  $2 \cdot (\text{number of edges in both sets}) \leq 2 \cdot (\text{number of edges in the cut})$ .

Hence we get the same conclusion as before.

So we get a  $\frac{1}{2}$  approximation algorithm for balanced max-cut.