Applications of Minimum Spanning Trees

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February 17, 2009

Minimum Spanning Trees

- We motivated MSTs through the problem of finding a low-cost network connecting a set of nodes.
- ▶ MSTs are useful in a number of seemingly disparate applications.
- ▶ We will consider two problems: clustering (Chapter 4.7) and minimum bottleneck graphs (problem 9 in Chapter 4).

Motivation for Clustering

- Given a set of objects and distances between them.
- ▶ Objects can be images, web pages, people, species
- Distance function: increasing distance corresponds to decreasing similarity.
- Goal: group objects into clusters, where each cluster is a set of similar objects.

- ▶ Let U be the set of n objects labelled p_1, p_2, \ldots, p_n .
- ▶ For every pair p_i and p_j , we have a distance $d(p_i, p_j)$.
- ▶ We require $d(p_i, p_i) = 0$, $d(p_i, p_j) > 0$, if $i \neq j$, and $d(p_i, p_j) = d(p_j, p_i)$

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- The spacing of a clustering is the smallest distance between objects in two different subsets:

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CLUSTERING OF MAXIMUM SPACING

INSTANCE: A set U of objects, a distance function $d: U \times U \to \mathbb{R}^+$, and a positive integer K

SOLUTION: A k-clustering of U whose spacing is the largest over all possible k-clusterings.

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- ▶ Process pairs of objects in increasing order of distance.
 - ▶ Let (p,q) be the next pair with $p \in C_p$ and $q \in C_q$.
 - ▶ If $C_p \neq C_q$, add new cluster $C_p \cup C_q$ to C, delete C_p and C_q from C.
- ▶ Stop when there are k clusters in C.

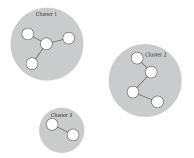


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- ▶ Stop when there are k clusters in C.
- ▶ Same as Kruskal's algorithm but do not add last k-1 edges in MST.

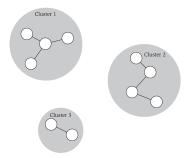


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- ▶ Let C be the clustering produced by the algorithm and let C' be any other clustering.
- ▶ What is spacing(C)? It is the cost of the (k-1)st most expensive edge in the MST. Let this cost be d^* .
- ▶ We will prove that spacing(C') ≤ d^* .

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- ▶ Suppose $p_i \in C'_s$ and $p_j \in C'_t$ in C'.

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- ▶ Suppose $p_i \in C'_s$ and $p_j \in C'_t$ in C'.
- ▶ All edges in the path Q connecting p_i and p_j in the MST have length $\leq d^*$.
- ▶ In particular, there is a point $p \in C'_s$ and a point $p' \notin C'_s$ such that p and p' are adjacent in Q.
- ▶ $d(p, p') \le d*$ \Rightarrow spacing $(C') \le d(p, p') \le d^*$.

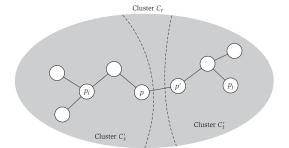


Figure 4.15 An illustration of the proof of (4.26), showing that the spacing of any other clustering can be no larger than that of the clustering found by the single-linkage algorithm.

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MINIMUM BOTTLENECK SPANNING TREE (MBST)

INSTANCE: An undirected graph G(V, E) and a function $c: E \to \mathbb{R}^+$

SOLUTION: A set $T \subseteq E$ of edges such that (V, T) is a spanning tree and there is no spanning tree in G with a cheaper bottleneck edge.

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- 1. Assume edge costs are distinct.
- 2. Is every MBST tree an MST? No. It is easy to create a counterexample.
- 3. Is every MST an MBST? Yes. Use the cycle property.
 - ▶ Let T be the MST and let T' be a spanning tree with a cheaper bottleneck edge. Let e be the bottleneck edge in T.
 - Every edge in T' is cheaper than e.
 - ▶ Adding e to T' creates a cycle consisting only of edges in T' and e.
 - Since e is the costliest edge in this cycle, by the cycle property, e cannot belong to any MST, which contradicts the fact that T is an MST.