# **Combinatorial Optimization**

# Fall 2013

# Assignment Sheet 6

Exercises marked with a  $\star$  can be handed in for bonus points. Due date is December 17.

In this exercise sheet, we will mostly deal with statements left unproved in the last two lectures.  $f: 2^U \to \mathbb{R}_+$  is always submodular. Recall that, for a set  $S \subseteq U$  and for an element  $u \in U$ , we have  $f_u(S) := f(S \cup u) - f(S)$  (i.e. the *marginal contribution* of u to S).

#### Exercise 1

Let  $H \subseteq U$ . Prove that  $g(A) = f(A \cup H)$  for  $A \subseteq U$  is a submodular function.

## Greedy

**Input**: a monotone, nonnegative submodular function  $f: 2^U \to \mathbb{R}_+$ , an integer k **Output**: a set  $S \subseteq U$  of cardinality k.

- 1. Let  $S_0 = \emptyset$ .
- 2. **For** i = 1, ..., k: let  $u_i \in U$  such that  $f_{u_i}(S_{i-1})$  is maximum. Set  $S_i = S_{i-1} \cup \{u_i\}$ .
- 3. Output *S*.

#### **Exercise 2**

Let  $S^*$  be the subset of U of maximum cost among those of cardinality k.

- (*i*) show that wrt the greedy algorithm above,  $f(S^*) f(S_i) \le (1 1/k)^i f(S^*)$  for each *i*. (Hint: find a lower bound on  $f_{u_i}(S_{i-1})$ , then prove the statement by induction on *i*.)
- (*ii*) apply (i) to deduce that  $f(S) \ge (1 1/e)f(S^*)$  (again wrt the greedy algorithm).

#### **Exercise 3**

Recall the following result (\*) proved in class, and prove its generalization (\*\*).

- (\*) For  $A \subseteq U$ , denote by A(p) a random subset of A where each element appears with probability p. Then  $E(f(A(p))) \ge (1-p)f(\emptyset) + pf(A)$ .
- (\*\*) Let  $A, B \subseteq U$ , and let A(p), B(q) be their independently sampled subsets, where each element of A appears in A(p) with probability p and each element of B appears in B(q) with probability q. Then  $E(f(A(p) \cup B(q))) \ge (1-p)(1-q)f(\emptyset) + p(1-q)f(A) + q(1-p)f(B) + pqf(A \cup B)$ . (Hint: first use (\*) (conditioning on the outcome of A(p)) and Exercise 1 to show

 $E(f(A(p) \cup B(q))) \ge E((1-q)f(A(p)) + qf(A(p) \cup B))$ . Then apply similar arguments to conclude.)

#### Exercise $4 (\star)$

Suppose f is nonnegative. For  $A \subseteq U$ , denote by A(p) a random subset of A where each element appears with probability  $at \ most \ p$  (nb: different elements may have different probabilities). Show that  $E(f(A(p))) \ge (1-p)f(\emptyset)$ .

#### **Exercise 5**

Recall the following algorithm, where we are assuming (wlog) that  $2k \le n$ , and that there is a subsets of U of 2k dummy elements whose marginal contribution to any set is 0.

## **Randomized Greedy**

**Input**: a nonnegative submodular function  $f: 2^U \to \mathbb{R}_+$ , an integer k **Output**: a set  $S \subseteq U$  of cardinality k.

- 1. Let  $S_0 = \emptyset$ .
- 2. **For** i = 1, ..., k,
  - a) Let  $M_i \subseteq U \setminus S_i$  be a set of size k maximizing  $\sum_{u \in M_i} f_u(S_i)$ .
  - b) Pick an element u of  $M_i$  uniformly at random, and set  $S_i = S_{i-1} \cup \{u\}$ .
- 3. Output  $S_k$ .

In class we saw that it provides a 1-1/e approximation for the problem  $\max\{f(S): |S| \le k\}$  when f is monotone. We are now going to show it gives a 1/e approximation for f non-monotone. Let  $S^*$  be the optimum solution. First, recall that in class we argued (for the monotone case, but the proof still holds) that:  $E(f_{u_i}(S_{i-1})) \ge \frac{1}{k} E(f(S_{i-1} \cup S^*) - f(S_{i-1}))$ .

- (i) Prove that  $f(S_{i-1} \cup S^*) \ge (1 1/k)^i f(S^*)$ . (Hint: use Exercises 1 and 4)
- (ii) Prove that  $E(f(S_i)) \ge (1/k)(1 1/k)^{i-1} f(S^*)$ , and conclude  $E(f(S_k)) \ge f(S^*)/e$ .

#### **Exercise 6**

Prove the following statement. Let  $(M, \mathcal{I})$  be a matroid, and A, B be two basis. Then there exists  $g: A \leftrightarrow B$  such that, for each  $u \in A$ ,  $(A \setminus \{u\}) \cup g(u)) \in \mathcal{I}$ .

#### Exercise 7 (\*)

In class we saw that, if we pick a random set where each element appears with probability 1/2, we obtain an expected 1/4 approximation to the problem of computing the set A that maximizes f(A). Call the output of this algorithm S, and the optimum solution  $S^*$ .

- (i) Is 1/4 the right answer? Give an example where the ratio  $E(f(S))/f(S^*)$  is as small as you can.
- (ii) Prove that, when f is the cut function of an undirected graph, S is a 1/2 approximation in expectation.