

Course Notes for EE227C (Spring 2018): Convex Optimization and Approximation

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26 Submodular functions and Lovász extension

Definition 26.1 (Submodular functions). Let $N = \{1, \dots, n\}$, and 2^N denote the power set of N . Then a function $f : 2^N \rightarrow \mathbb{R}$ is *submodular* if:

$$f(A \cup \{j\}) - f(A) \geq f(B \cup \{j\}) - f(B)$$

for all $A \subseteq B \subseteq N$ and for all $j \in N$.

This is equivalent to:

$$f(A \cap B) + f(A \cup B) \leq f(A) + f(B)$$

for all $\forall A, B \subseteq N$.

Some examples of submodular functions:

- Let A be a subset of nodes of a bipartite graph. Then $f(A) = |\text{neighborhood}(A)|$ is submodular.
- Let $X = \{X_1, \dots, X_n\}$ be a set of random variables, and $A \subseteq X$. Then $f(A) = H(A)$, where H denotes entropy, is submodular.
- Let E, V be the edges and nodes of a graph, and $A \subseteq V$. Then the size of the graph cut, $f(A) = |\{(u, v) \in E : u \in A, v \in V \setminus A\}|$, is submodular.

There is an equivalence between submodular and boolean functions: a submodular function $f : 2^N \rightarrow \mathbb{R}$ can be expressed as a boolean function $f : \{0, 1\}^n \rightarrow \mathbb{R}$.

Definition 26.2 (Lovász extension). Let $f(x)$ be a submodular function. Then its Lovász extension, $\hat{f}(z) : [0, 1]^n \rightarrow \mathbb{R}$ is given by:

$$\hat{f}(z) = \mathbb{E}_{\lambda \sim U(0,1)} \left[f(\{i : z_i \geq \lambda\}) \right]$$

where $U(0, 1)$ denotes the uniform distribution on $[0, 1]$.

Some observations about $\hat{f}(z)$:

- For any $x \in \{0, 1\}^n$, $\hat{f}(x) = f(x)$. That is, f and \hat{f} agree on the domain of f .
- For any $z \in [0, 1]^n$, there exists an $x \in \{0, 1\}^n$, such that $f(x) \leq \hat{f}(z)$. Also, $\min_{x \in \{0, 1\}^n} f(x) = \min_{z \in [0, 1]^n} \hat{f}(z)$.

Theorem 26.3 (Submodularity and convexity). Consider a function $f(x) : 2^N \rightarrow \mathbb{R}$ and its Lovász extension $\hat{f}(z) : 2^N \rightarrow \mathbb{R}$. Then $f(x)$ is submodular iff $\hat{f}(z)$ is convex.

Proof. We only show the forward direction (f is submodular $\Rightarrow \hat{f}$ is convex), the reverse direction is left as an exercise.

Without loss of generality, assume z is order: $z_1 \geq z_2 \geq \dots \geq z_n$.

Also, let $S_i = \{1, 2, \dots, i\}$. We also take $f(\emptyset) = 0$.

Then:

$$\begin{aligned} \hat{f}(z) &= \sum_{i=1}^{n-1} P(z_{i+1} \leq \lambda \leq z_i) \cdot f(S_i) + z_n \cdot f(S_n) \\ &= \sum_{i=1}^{n-1} (z_{i+1} - z_i) \cdot f(S_i) + z_n \cdot f(S_n) \end{aligned}$$

Now, consider the following lemma:

Lemma 26.4. The Lovász extension $\hat{f}(z)$ of submodular $f(x)$ can be expressed as an LP:

$$\hat{f}(z) = \max_x x^T z : x(S) \leq f(S), x(N) = f(N), \forall S \subseteq N$$

where $x(S) = \sum_{i \in S} x_i$. Let F denote the feasible region.

Given [Lemma 26.4](#), the rest of the proof follows easily:

$$\begin{aligned} \hat{f}(\lambda z + (1 - \lambda)z') &= \max_{x \in F} x^T (\lambda z + (1 - \lambda)z') \\ &\leq \lambda \max_{x \in F} x^T z + (1 - \lambda) \max_{x \in F} x^T z' \\ &= \lambda \hat{f}(z) + (1 - \lambda) \hat{f}(z') \end{aligned}$$

Now we just have to prove [Lemma 26.4](#).

Consider the dual problem to the above LP:

$$\begin{aligned} \min_y \quad & \sum_{S \subseteq N} y_S \cdot f(S) : \sum_{S \subseteq N} y_S e_S = z, y_S \geq 0 \\ \text{where } (e_S)_i = & \begin{cases} 1 & \text{if } i \in S \\ 0 & \text{else} \end{cases} \end{aligned}$$

Our goal is to find a primal-dual feasible solution x^*, y^* that satisfies:

$$\hat{f}(z) = c^T x^* = \sum_{S \subseteq N} y_S^* \cdot f(S)$$

Consider the following x^*, y^* :

$$\begin{aligned} x_i^* &= f(S_i) - f(S_{i-1}) \\ y_S^* &= \begin{cases} z_i - z_{i+1} & \text{if } S = S_i \\ z_n & \text{if } S = N \\ 0 & \text{else} \end{cases} \end{aligned}$$

First, we show that x^* is primal feasible.

For the constraint $x(N) = f(N)$, we see that:

$$\begin{aligned} x^*(N) &= \sum_{i=1}^n x_i^* = \sum_{i=1}^n f(S_i) - f(S_{i-1}) \\ &= f(N) - f(\emptyset) = f(N) \end{aligned}$$

For the other constraint, $x(S) \leq f(S)$, we proceed by induction on $|S|$ (the size of S).

As a base case, we take $S = \emptyset$.

Suppose that the constraint is satisfied for all sets of size $i - 1$.

Then consider an arbitrary S , whose largest element is i :

$$\begin{aligned} f(S) + f(S_{i-1}) &\geq f(S \cup S_{i-1}) + f(S \cup S_{i-1}) \\ &= f(S_i) + f(S \setminus \{i\}) \\ &\geq f(S_i) + x^*(S \setminus \{i\}) \\ \Rightarrow f(S) &\geq f(S_i) - f(S_{i-1}) + x^*(S \setminus \{i\}) \\ &= x_i^* + x^*(S \setminus \{i\}) \\ &= x^*(S) \end{aligned}$$

where we used fact that f is submodular and that i is the largest element in S .

It follows that $x^*(S) \leq f(S)$, for all $S \subseteq N$, and thus x^* is primal feasible.

Next, we show that y^* is dual feasible.

$$\begin{aligned}
\sum_{S \subseteq N} y_S^* \cdot f(S) &= \sum_{j=1}^{n-1} (z_j - z_{j+1}) e_{S_j} + z_n e_n \\
\left(\sum_{S \subseteq N} y_S^* \cdot f(S) \right)_i &= \left(\sum_{j=1}^{n-1} (z_j - z_{j+1}) e_{S_j} + z_n e_n \right)_i \quad (\text{consider it elementwise}) \\
&= \sum_{j=i}^n (z_j - z_{j+1}) + z_n \\
&= z_i
\end{aligned}$$

Since this equality holds elementwise for each i , we see that $\sum_{S \subseteq N} y_S^* \cdot f(S) = z$.

We also observe that $y_S^* \geq 0$ due to the nonincreasing ordering of z_1, \dots, z_n .

Thus, it follows that y_S^* is dual feasible.

Finally, we show that x^*, y^* are optimal, achieving a duality gap of zero:

$$\begin{aligned}
z^T x^* &= \sum_{i=1}^n z_i (f(S_i) - f(S_{i-1})) \\
&= \sum_{i=1}^n (z_i - z_{i+1}) f(S_i) + z_n f(S_n) \\
&= \hat{f}(z) \\
\hat{f}(z) &= \sum_{i=1}^n (z_i - z_{i+1}) f(S_i) + z_n f(S_n) \\
&= \sum_{S \subseteq N} y_S^* f(S)
\end{aligned}$$

This concludes the proof of [Lemma 26.4](#), and that of [Theorem 26.3](#).

References