Weekly Exercise 8

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Branch and Bound

(2 points)

Exercise 1 (Lower bound, 2 Points). For a finite set Ω , consider the following segmentation energy function $E: \mathbb{B}^n \to \mathbb{R}$:

$$E(\mathbf{y}) = \min_{\omega \in \Omega} C(\omega) + \sum_{i=1}^{n} f_i(\omega) y_i + \sum_{i=1}^{n} b_i(\omega) (1 - y_i) + \sum_{i=1}^{n} \sum_{j \in \mathcal{N}(i)} w_{ij}(\omega) |y_i - y_j|, \quad (1)$$

with $C: \Omega \to \mathbb{R}$, $f_i: \Omega \to \mathbb{R}$, $b_i: \Omega \to \mathbb{R}$, $w_{ij}: \Omega \to \mathbb{R}$. Prove the following lower bound:

$$E(\mathbf{y}) \ge \left(\min_{\omega \in \Omega} C(\omega)\right) + \sum_{i=1}^{n} \left(\min_{\omega \in \Omega} f_i(\omega)\right) y_i + \sum_{i=1}^{n} \left(\min_{\omega \in \Omega} b_i(\omega)\right) (1 - y_i)$$

$$+ \sum_{i=1}^{n} \sum_{j \in \mathcal{N}(i)} \left(\min_{\omega \in \Omega} w_{ij}(\omega)\right) |y_i - y_j| =: \ell(\mathbf{y}, \Omega).$$
(2)

Remark: This shows that $E^* = \min_{\mathbf{y}} E(\mathbf{y}) \ge \min_{\mathbf{y}} \ell(\mathbf{y}, \Omega) = L(\Omega)$ and $L(\Omega)$ is a lower bound for the global optimum. Note that the lower bound $L(\Omega)$ fulfills three important properties which make it applicable for branch and bound optimization methods:

- 1. Monotonicity: $\Omega_1 \subset \Omega_2 \Rightarrow L(\Omega_1) \geq L(\Omega_2)$.
- 2. **Computability:** Evaluating $L(\Omega)$ for some given Ω corresponds to minimizing a submodular quadratic pseudo-Boolean function.
- 3. **Tightness:** For $|\Omega| = 1$, i.e. $\Omega = {\omega}$ we have $L({\omega}) = \min_{\mathbf{y}} E(\mathbf{y})$.

Solution. One can prove the inequality from the mathematical deduction,

$$E(\mathbf{y}) = \min_{\omega \in \Omega} C(\omega) + \sum_{i=1}^{n} f_{i}(\omega)y_{i} + \sum_{i=1}^{n} b_{i}(\omega)(1 - y_{i}) + \sum_{i=1}^{n} \sum_{j \in \mathcal{N}(i)} w_{ij}(\omega)|y_{i} - y_{j}|$$

$$\geq \min_{\omega \in \Omega} C(\omega) + \min_{\omega \in \Omega} \sum_{i=1}^{n} f_{i}(\omega)y_{i} + \min_{\omega \in \Omega} \sum_{i=1}^{n} b_{i}(\omega)(1 - y_{i})$$

$$+ \min_{\omega \in \Omega} \sum_{i=1}^{n} \sum_{j \in \mathcal{N}(i)} w_{ij}(\omega)|y_{i} - y_{j}|$$

$$= \min_{\omega \in \Omega} C(\omega) + \sum_{i=1}^{n} \min_{\omega \in \Omega} f_{i}(\omega)y_{i} + \sum_{i=1}^{n} \min_{\omega \in \Omega} b_{i}(\omega)(1 - y_{i})$$

$$+ \sum_{i=1}^{n} \sum_{j \in \mathcal{N}(i)} \min_{\omega \in \Omega} w_{ij}(\omega)|y_{i} - y_{j}|$$

$$= \left(\min_{\omega \in \Omega} C(\omega)\right) + \sum_{i=1}^{n} \left(\min_{\omega \in \Omega} f_{i}(\omega)\right)y_{i} + \sum_{i=1}^{n} \left(\min_{\omega \in \Omega} b_{i}(\omega)\right)(1 - y_{i})$$

$$+ \sum_{i=1}^{n} \sum_{j \in \mathcal{N}(i)} \left(\min_{\omega \in \Omega} w_{ij}(\omega)\right)|y_{i} - y_{j}| =: \ell(\mathbf{y}, \Omega).$$
(3)

Or, since y_i , $1 - y_i$ and $|y_i - y_j|$ are all positive for $\mathbf{y} \in \mathbb{B}^n$ and we have by definition from the minimum,

$$C(\omega') \ge \min_{\omega \in \Omega} C(\omega), \ \forall \omega' \in \Omega,$$
 (4)

the inequality holds.

Programming (7 points)

Exercise 2 (Branch-and-Mincut¹, 7 Points). In this exercise we apply the branch and bound method from the lecture to find a *global minimizer* of a discrete version of the celebrated Chan-Vese² segmentation energy function:

$$E(\mathbf{y}, \{c_f, c_b\}) = \mu \sum_{i=1}^n \sum_{j \in \mathcal{N}(i)} |y_i - y_j| + \sum_{i=1}^n \left(\nu + \lambda_1 (I_i - c_f)^2\right) y_i + \sum_{i=1}^n \lambda_2 (I_i - c_b)^2 (1 - y_i).$$
(5)

Here I denotes a gray-scale input image with n pixels, i.e. at every pixel $1 \le i \le n$ we have $I_i \in [0,255]$. The variable $\omega = (c_f,c_b) \in \Omega = [0,255]^2$ denotes the mean intensity of foreground respectively the background of the segmentation $x \in \mathbb{B}^n$.

¹V. Lempitsky, A. Blake, C. Rother, Image Segmentation by Branch-and-Mincut, ECCV 2008

²T. Chan, L. Vese: Active contours without edges. Trans. Image Process., 10(2), 2001.

Compute a global minimizer of (5) using the branch and bound best-first tree search. The search space Ω is the rectangle $[0,255]^2$. In your implementation, you can keep a sorted queue of rectangles Ω_i , and every iteration remove the rectangle with the smallest lower bound and split it into two smaller rectangles along the longest edge. As a lower bound on (5) use the bound (2) dervied in the theoretical exercise. You can use chanvese_global.cpp as a starting point.





Figure 1: The figure shows the input image and a global minimizer of (5) for parameters $\lambda_1 = \lambda_2 = 0.0001$, $\mu = 1$, $\nu = 0.1$. The optimal foreground and background colors were found as $c_f^* = 81$ and $c_b^* = 167$.