

GOP Report

October 29, 2017

1 Problem Formulation

$$\begin{aligned} & \min_{\theta, \bar{b}} \\ & \sum_h \sum_{X_h} \theta_h(X_h) (M * b(X_h) - \bar{b}(X_h)) + \sum_i \sum_{X_i} \sum_{X_h} \theta_h(X_i, X_h) (M * b(X_i, X_h) - \bar{b}(X_i, X_h)) + \\ & \sum_m \sum_{X_h} b_h^{(m)} \log(b_h^{(m)}) - \frac{\lambda}{2} \|\theta\|^2 \\ & \text{S.t:} \\ & 1- \bar{b}(X_h) = \sum_{j=1}^m b^{(m)}(X_h) \\ & 2- \bar{b}(X_i, X_h) = \sum_{j=1}^m b^{(m)}(X_i, X_h) \\ & 3- b^{(m)}(X_h) = \sum_{X_i} b^{(m)}(X_i, X_h) \end{aligned}$$

2 Primal Formulation

$$\begin{aligned} & \min_{\theta, \bar{b}} \\ & \sum_h \sum_{X_h} \theta_h(X_h) (M * b(X_h) - \bar{b}(X_h)) + \sum_i \sum_{X_i} \sum_{X_h} \theta_h(X_i, X_h) (M * b(X_i, X_h) - \bar{b}(X_i, X_h)) + \\ & \sum_m \sum_{X_h} b_h^{(m)} \log(b_h^{(m)}) - \frac{\lambda}{2} \|\theta\|^2 + \mu_1 \left(\sum_{j=1}^m b^{(m)}(X_h) - \bar{b}(X_h) \right) + \mu_2 \left(\sum_{j=1}^m b^{(m)}(X_i, X_h) - \right. \\ & \left. \bar{b}(X_i, X_h) \right) + \sum_{j=1}^m \gamma_j \left(\sum_{X_i} b^{(m)}(X_i, X_h) - b^{(m)}(X_h) \right) \end{aligned}$$

3 Create upper bound by solving the Primal problem for fixed θ

$$\begin{aligned} & P^k(\theta^k) = \min_{\bar{b}} \\ & \sum_h \sum_{X_h} \theta_h^k(X_h) (M * b(X_h) - \bar{b}(X_h)) + \sum_i \sum_{X_i} \sum_{X_h} \theta_h^k(X_i, X_h) (M * b(X_i, X_h) - \bar{b}(X_i, X_h)) + \\ & \sum_m \sum_{X_h} b_h^{(m)} \log(b_h^{(m)}) + \mu_1 \left(\sum_{j=1}^m b^{(m)}(X_h) - \bar{b}(X_h) \right) + \mu_2 \left(\sum_{j=1}^m b^{(m)}(X_i, X_h) - \bar{b}(X_i, X_h) \right) \end{aligned}$$

$$) + \sum_{j=1}^m \gamma_j (\sum_{X_i} b^{(m)}(X_i, X_h) - b^{(m)}(X_h))$$

we will take the derivatives to solve for both μ_1, μ_2

$$\frac{\partial P^k(\theta^k)}{\partial b(X_h)} = -\theta_h^k(X_h) - \mu_1 = 0 \implies -\theta_h^k(X_h) = \mu_1$$

$$\frac{\partial P^k(\theta^k)}{\partial b(X_i, X_h)} = -\theta_h^k(X_i, X_h) - \mu_1 = 0 \implies -\theta_h^k(X_i, X_h) = \mu_2$$

Note: I didn't calculate γ because it's irrelevant in the Lower bound construction

4 Relaxed Dual Problem

First let's define L^k as follow :

$$L^k(\bar{b}^{B_i}, \theta, \mu^k) =$$

$$\sum_h \sum_{X_h} \theta_h(X_h)(M * b(X_h)) + \sum_i \sum_{X_i} \sum_{X_h} \theta_h(X_i, X_h)(M * b(X_i, X_h)) + \frac{\lambda}{2} \|\theta\|^2 + \sum_h \sum_{X_h} \bar{b}(X_h) * (-1 * (\theta_h(X_h) + \theta_h(X_h)^k)) + \sum_h \sum_{X_h} \bar{b}(X_i, X_h) * (-1 * (\theta_h(X_i, X_h) + \theta_h(X_i, X_h)^k)) =$$

$$L^k(\theta, \mu^k) + \sum_h \sum_{X_h} \bar{b}(X_h) * g(X_h) + \sum_h \sum_{X_h} \bar{b}(X_i, X_h) * g(X_i, X_h)$$

Where,

$$L^k(\theta, \mu^k) = \sum_h \sum_{X_h} \theta_h(X_h)(M * b(X_h)) + \sum_i \sum_{X_i} \sum_{X_h} \theta_h(X_i, X_h)(M * b(X_i, X_h)) + \frac{\lambda}{2} \|\theta\|^2,$$

$$g(X_h) = (-1 * (\theta_h(X_h) + \theta_h(X_h)^k))$$

$$g(X_i, X_h) = (-1 * (\theta_h(X_i, X_h) + \theta_h(X_i, X_h)^k))$$

With out loss of generality let's assume that θ is a positive vector (we can force this constraint by shifting all singular values of θ by constant scalar)

Given this assumption we can argue that:

- 1- $g(X_h)$ is always negative for any selection of $\bar{b}(X_h)$
- 2- $g(X_i, X_h)$ is always negative for any selection of $\bar{b}(X_i, X_h)$

from 1, 2 only one relaxed dual problem needs to be solved, with a valid under

estimator to $L^k(\bar{b}^{B_l}, \theta, \mu^k)$, where $\bar{b}(X_h)$, and $\bar{b}(X_i, X_h)$ are fixed to their upper bound.