Optimization Theory and Algorithm

October 7, 2021

Homework 3

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HW 1 We consider the following optimization problem:

$$\min_{\mathbf{x}} f(\mathbf{x}) = \frac{1}{m} \sum_{i=1}^{m} f_i(\mathbf{x}).$$

And assume that f is β -smooth and α -strong convex. Using mini-bath SGD with fixed learning rate to solve it as

$$\mathbf{x}^{t+1} = \mathbf{x}^t - \frac{s}{n_b} \sum_{i_t \in D_t} \nabla f_{i_t}(\mathbf{x}^t),$$

where $D_t \subset \{1, 2, ..., m\}$ are drawn randomly and $|D_t| = n_b$ is the size of D_t . We further suppose that

- (1) The index D_t does not depended from the previous $D_0, D_1, \ldots, D_{t-1}$.
- (2) $\mathbb{E}_{i_t \in D_t}[\nabla f_{i_t}(\mathbf{x}^t)] = \nabla f(\mathbf{x}^t)$ (Unbiased Estimation).
- (3) $\mathbb{E}_{i_t \in D_t}[\|\nabla f_{i_t}(\mathbf{x}^t)\|^2] = \sigma^2 + \|\nabla f(\mathbf{x}^t)\|^2$ (control the variance).

Prove

(i)
$$\mathbb{E}_{D_t}[f(\mathbf{x}^{t+1})] \leq f(\mathbf{x}^t) - s\nabla f(\mathbf{x}^{t+1})^{\top} \mathbb{E}_{D_t}[\mathbf{g}^t] + \frac{\beta s^2}{2} \mathbb{E}_{D_t}[\|\mathbf{g}^t\|^2],$$
 where $\mathbf{g}^t = \frac{1}{n_b} \sum_{i \in D_t} \nabla f_i(\mathbf{x}^t).$

(ii)
$$\mathbb{E}_{D_t}[f(\mathbf{x}^{t+1}) - f(\mathbf{x}^t)] \le -(s - \frac{\beta s^2}{2n_b}) \|\nabla f(\mathbf{x}^t)\|^2 + \frac{\beta s^2}{2n_b} \sigma^2.$$

(iii) Let $s \leq \min\{1/\beta, \frac{n_b}{\alpha(2n_b-1)}\}$, then

$$\mathbb{E}[f(\mathbf{x}^{t+1}) - f^*] - \frac{\beta s}{2\alpha(2n_b - 1)}\sigma^2 \le (1 - \alpha s(2 - 1/n_b)) \left[\mathbb{E}[f(\mathbf{x}^t) - f^*] - \frac{\beta s}{2\alpha(2n_b - 1)}\sigma^2 \right].$$

HW 2 Derive BCD algorithm for LASSO problem.

HW 3 Derive ADMM algorithm for Fused LASSO problem.

HW 4 Consider the following Basis Pursuit problem as:

$$\min_{\mathbf{x}} \|\mathbf{x}\|_1$$

$$s.t. \ A\mathbf{x} = \mathbf{b}.$$

Derive ADMM algorithm for it.

Hit: using the indicator function of $\Omega = \{\mathbf{x} | A\mathbf{x} = \mathbf{b}\}.$

References