THE CHINESE UNIVERSITY OF HONG KONG Department of Mathematics MATH4230 2024-25 Lecture 23 April 9, 2025 (Wednesday)

Uzawa Algorithm

Recall the optimization problem with only the inequality constraints:

$$P = \inf_{\substack{g_1(x) \le 0 \\ \vdots \\ g_n(x) \le 0}} f(x)$$
(P)

Condition:

- 1. f, g_1, \ldots, g_m are convex functions
- 2. $\operatorname{Hess}(f) \geq \alpha I_n$
- 3. g_1, \ldots, g_m are C-Lipschitz

4.
$$\eta \in \left(0, \frac{2\alpha}{C^2}\right)$$

and its dual problem is given by

$$D = \max_{\lambda \in \mathbb{R}^m_+} d(\lambda),$$

where
$$d(\lambda) = \inf_{x \in \mathbb{R}^n} \left(f(x) + \sum_{i=1}^m \lambda_i g_i(x) \right).$$

Recall that under Slater condition (i.e. $\exists \hat{x} \text{ such that } g_i(\hat{x}) < 0, \forall i = 1, ..., m$), we have

•
$$P = D$$

• there exists $\lambda^* \in \mathbb{R}^m_+$ such that $d(\lambda^*) = f(x^*) + \sum \lambda_i^* g_i(x^*) = D$

and the main algorithm Uzawa Algorithm:

$$\lambda^{k+1} = \prod_{\mathbb{R}^m_+} \left(\lambda^k + \eta \nabla d(\lambda^k) \right) = \prod_{\mathbb{R}^m_+} \left(\lambda^k + \eta g(x^k) \right)$$

where $\Pi_{\mathbb{R}^m_+}$ denotes the projection onto \mathbb{R}^m_+ , and $x^k := \underset{x \in \mathbb{R}^n}{\operatorname{argmin}} \left(f(x) + \sum \lambda_i^k g_i(x) \right)$

Lemma 1. From the Uzawa algorithm, we always have $\lambda^* = \prod_{\mathbb{R}^m_+} (\lambda^* + \eta g(x^*))$

Proof. From $d(\lambda^*) = \min_{\lambda \in \mathbb{R}^m_+} d(\lambda)$, we have

$$0 \stackrel{\text{KKT theorem}}{=} \sum_{i=1}^{m} \lambda_i^* g_i(x^*) \ge \sum_{i=1}^{m} \lambda_i g_i(x^*)$$

This means that

$$\sum_{i=1}^{m} (\lambda_i^* - \lambda_i) g_i(x^*) \ge 0 \iff \langle \lambda^* - \lambda, g(x^*) \rangle \ge 0$$
$$\implies \langle \lambda^* - \lambda, \lambda^* - (\lambda^* + \eta g(x^*)) \rangle \le 0, \quad \forall \lambda \in \mathbb{R}^m_+$$
$$\implies \lambda^* = \Pi_{\mathbb{R}^m_+} (\lambda^* + \eta g(x^*))$$

Theorem 2. Under the above conditions, there exists a unique optimal solution to (P) and $x^k \to x^*$ as $k \to +\infty$.

- *Proof.* 1. To prove there exists there is unique x^* , it is exactly the same as the last theorem since f is coercive.
 - 2. Consider

$$\begin{aligned} \|\lambda^{k+1} - \lambda^*\|^2 &= \|\Pi_{\mathbb{R}^m_+}(\lambda^k + \eta g(x^k)) - \Pi_{\mathbb{R}^m_+}(\lambda^* + \eta g(x^*))\|^2 \\ &\leq \|\lambda^k + \eta g(x^k) - (\lambda^* + \eta g(x^*))\|^2 \\ &\leq \|\lambda^k - \lambda^*\|^2 + 2\eta \left\langle \lambda^k - \lambda^*, g(x^k) - g(x^*) \right\rangle + \eta^2 \|g(x^k) - g(x^*)\|^2 \end{aligned}$$

Since we assume g_1, \ldots, g_m are Lipschitz, so the last term we have

$$||g(x^k) - g(x^*)|| \le C||x^k - x^*||$$

Moreover, by the Euler's first order condition, we have

$$\begin{cases} \nabla f(x^*) + \langle \lambda^*, \nabla g(x^*) \rangle = \mathbf{0} \\ \nabla f(x^k) + \langle \lambda^k, \nabla g(x^k) \rangle = \mathbf{0} \end{cases}$$

Taking differencing of the above two equations gives

$$\nabla f(x^*) - \nabla f(x^k) + \langle \lambda^*, \nabla g(x^*) \rangle - \langle \lambda^k, \nabla g(x^k) \rangle = \mathbf{0}$$

Moreover, recall $\langle \nabla f(x^*) - \nabla f(x^k), x^* - x^k \rangle \ge \alpha ||x^* - x^k||^2$ as $\operatorname{Hess}(f) \ge \alpha I_n$. Combining the above equality and the inequality, we have

$$\sum_{i=1}^{m} \lambda_{i}^{*} \underbrace{\left\langle \nabla g_{i}(x^{*}), x^{k} - x^{*} \right\rangle}_{\leq g_{i}(x^{k}) - g_{i}(x^{*})} + \sum_{i=1}^{m} \lambda_{i}^{k} \underbrace{\left\langle \nabla g_{i}(x^{k}), x^{*} - x^{k} \right\rangle}_{\leq g_{i}(x^{*}) - g_{i}(x^{k})} \geq \alpha \|x^{*} - x^{k}\|^{2}$$

and the above bounds are followed by the convexity of g_i . Therefore, we have

$$\sum_{i=1}^{m} \lambda_{i}^{*}(g_{i}(x^{k}) - g_{i}(x^{*})) + \sum_{i=1}^{m} \lambda_{i}^{k}(g_{i}(x^{k}) - g_{i}(x^{k})) \ge \alpha ||x^{*} - x^{k}||^{2}$$
$$\sum_{i=1}^{m} (\lambda_{i}^{*} - \lambda_{i}^{k}) \left(g_{i}(x^{k}) - g_{i}(x^{*})\right) \ge \alpha ||x^{*} - x^{k}||^{2}$$
$$\left\langle \lambda^{*} - \lambda^{k}, g(x^{k}) - g(x^{*})\right\rangle \ge \alpha ||x^{*} - x^{k}||^{2}$$

Prepared by Max Shung

Putting all together back to the top inequality, we have

$$\begin{aligned} \|\lambda^{k+1} - \lambda^*\|^2 &\leq \|\lambda^k - \lambda^*\|^2 + 2\eta \left\langle \lambda^k - \lambda^*, g(x^k) - g(x^*) \right\rangle + \eta^2 \|g(x^k) - g(x^*)\|^2 \\ &\leq \|\lambda^k - \lambda^*\|^2 + 2\eta (-2\eta\alpha \|x^* - x^k\|^2) + \eta^2 C^2 \|x^k - x^*\|^2 \\ &= \|\lambda^k - \lambda^*\|^2 + \left(\eta^2 C^2 - 2\eta\alpha\right) \|x^k - x^*\|^2 \end{aligned}$$

So, we can deduce that

 $\begin{array}{l} \textcircled{1} \quad k \mapsto \|\lambda^k - \lambda^*\| \geq 0 \text{ is decreasing, hence converges.} \\ \text{So by Cauchy criterion, we can deduce that } \|\lambda^{k+1} - \lambda^*\| - \|\lambda^k - \lambda^*\| \to 0 \text{ as } k \to +\infty \end{array}$

(2)
$$||x^k - x^*||^2 \le \frac{1}{\eta^2 C^2 - 2\eta\alpha} \left(||\lambda^{k+1} - \lambda^*||^2 - ||\lambda^k - \lambda^*||^2 \right) \to 0$$

— End of Lecture 23 —