

NOTES ON OPTIMISATION WITH COMMUNICATION CONSTRAINTS

Consider the optimisation problem P :

$$(1) \quad \min_{x \in D} f(x)$$

$$(2) \quad s.t. \quad g_j(x) \leq 0, \quad j = 1, \dots, m$$

where $x \in \mathcal{R}^n$, $f : \mathcal{R}^n \rightarrow \mathcal{R}$ and $g : \mathcal{R}^n \rightarrow \mathcal{R}$ are both convex. We assume that $f(\cdot)$ and $g(\cdot)$ are Lipschitz on domain D (this is quite a mild assumption for convex functions) and have bounded subgradients on D . Let D^* denote set of optima solving problem P .

Define

$$(3) \quad F(x, \lambda) := f(x) + \lambda \sum_{j=1}^m \max\{0, g_j(x)\}$$

and consider the iterative update

$$(4) \quad x(k+1) = x(k) - \alpha(k) \partial_x F(x(k), \lambda(k))$$

$$(5) \quad \lambda(k+1) = \lambda(k) + \alpha(k)$$

with $\alpha(k) > 0$. For $x^* \in D^*$, following the usual Bertsekas argument we have that

$$(6) \quad \|x(k+1) - x^*\|^2 = \|x(k) - \alpha(k) \partial_x F(x(k), \lambda(k)) - x^*\|^2$$

$$(7) \quad = \|x(k) - x^*\|^2 - 2\alpha(k) \partial_x F^T(x(k), \lambda(k))(x(k) - x^*) + \alpha^2(k) \|\partial_x F(x(k), \lambda(k))\|^2$$

$$(8) \quad \stackrel{(a)}{\leq} \|x(k) - x^*\|^2 - 2\alpha(k)(F(x(k), \lambda(k)) - F(x^*, \lambda(k))) + \alpha^2(k) \|\partial_x F(x(k), \lambda(k))\|^2$$

where (a) follows from the convexity of $F(x, \lambda)$ and the definition of the subgradient. It now follows that

$$(9) \quad \|x(k+1) - x^*\|^2 \leq \|x(1) - x^*\|^2 - 2 \sum_{i=1}^k \alpha(i)(F(x(i), \lambda(i)) - F(x^*, \lambda(i))) + 2 \sum_{i=1}^k \alpha^2(i) \|\partial_x F(x(i), \lambda(i))\|^2$$

By assumption, there exists a constant \bar{v} such that $\|\partial f(x)\| \leq \bar{v}$ and $\|\partial g(x)\| \leq \bar{v}$. Hence, $\|\partial_x F(x(i), \lambda(i))\| \leq (1 + m\lambda(i))\bar{v}$ and

$$(10) \quad \|x(k+1) - x^*\|^2 \leq \|x(1) - x^*\|^2 - 2 \sum_{i=1}^k \alpha(i)(F(x(i), \lambda(i)) - F(x^*, \lambda(i))) + 2\bar{v}^2 \sum_{i=1}^k \alpha^2(i)(1 + m\lambda(i))^2$$

That is,

$$(11) \quad \sum_{i=1}^k \alpha(k)(F(x(i), \lambda(i)) - F(x^*, \lambda(i))) \leq \frac{1}{2} \|x(1) - x^*\|^2 + \bar{v}^2 \sum_{i=1}^k \alpha^2(i)(1 + m\lambda(i))^2$$

Suppose $\lim_{k \rightarrow \infty} \sum_{i=1}^k \alpha^2(i)(1 + m\lambda(i))^2 = 0$ and $\sum_{i=1}^k \alpha(i) \rightarrow \infty$, $\lambda(k) \rightarrow \infty$ as $k \rightarrow \infty$. Then

$$(12) \quad \lim_{k \rightarrow \infty} \sum_{i=1}^k \alpha(k)(F(x(i), \lambda(i)) - F(x^*, \lambda(i))) \leq \frac{1}{2} \|x(1) - x^*\|^2$$

and so

$$(13) \quad \lim_{k \rightarrow \infty} \left(\min_{i \in \{1, \dots, k\}} F(x(i), \lambda(i)) - F(x^*, \lambda(i)) \right) \sum_{i=1}^k \alpha(k) \leq \frac{1}{2} \|x(1) - x^*\|^2$$

Since the RHS is finite and $\sum_{i=1}^k \alpha(i) \rightarrow \infty$ as $k \rightarrow \infty$, it follows that $\min_{i \in \{1, \dots, k\}} (F(x(i), \lambda(i)) - F(x^*, \lambda(i))) = 0$ as $k \rightarrow \infty$ *****need to double check this argument carefully; note also that it doesn't prove convergence – can we do better ?**.

Recall that $F(x(k), \lambda(k)) - F(x^*, \lambda(k)) = f(x(k)) - f(x^*) + \lambda(k) \max\{0, g(x(k))\}$. Now $f(x(k)) - f(x^*) \geq 0$ with equality only when $x(k) \in D^*$, and $\lambda(k) \max\{0, g(x(k))\} \geq 0$ with equality only when $g(x(k)) \leq 0$ (since $\lambda(k) > 0$). Since $\min_{i \in \{1, \dots, k\}} (F(x(i), \lambda(i)) - F(x^*, \lambda(i))) = 0$ as $k \rightarrow \infty$, it follows that the minimising $x(i)$ must lie in set D^* .