

9. Proof of Lemma 1

Let $(\mathbf{x}^*, \boldsymbol{\alpha}^*)$ be a solution to the VI in (17). We want to show that $(\mathbf{x}^*, \boldsymbol{\alpha}^*) \in \mathcal{N}_{\text{opt}}$. First, since the VI holds for all $\mathbf{x}, \boldsymbol{\alpha}$ we can pick $\boldsymbol{\alpha} = \boldsymbol{\alpha}^*$, so for all $\mathbf{x} \in \mathcal{X}$

$$\langle \mathbf{x} - \mathbf{x}^*, F(\mathbf{x}^*, \boldsymbol{\alpha}^*) \rangle \leq 0. \quad (21)$$

Then by Proposition 1.4.2 in (Facchinei & Pang, 2007), \mathbf{x}^* is a NE (note that because $F(\mathbf{x}, \boldsymbol{\alpha}^*)$ is strongly monotone in \mathbf{x} , then $r_n(\mathbf{x}_n, \mathbf{x}_{-n})$ is concave in \mathbf{x}_n for each n and \mathbf{x}_{-n}).

For any k , we can pick $\mathbf{x} = \mathbf{x}^*$ and $\boldsymbol{\alpha} = \boldsymbol{\alpha}_0$ such that $\alpha_0^l = \alpha^{*l}$ for all $l \neq k$ and $\alpha_0^k = \alpha^{*k} + \varepsilon$ for some $\varepsilon > 0$, and get from the VI in (17) that

$$\varepsilon \sum_{n=1}^N (x_n^{*k} - l_k^*) \leq 0 \implies \sum_{n=1}^N x_n^{*k} \leq l_k^*. \quad (22)$$

Now let k be a coordinate for which $\alpha^{*k} > 0$ (if it exists) and pick $\mathbf{x} = \mathbf{x}^*$ and $\boldsymbol{\alpha} = \boldsymbol{\alpha}_0$ such that $\alpha_0^l = \alpha^{*l}$ for all $l \neq k$ and $\alpha_0^k = 0$. Then the VI in (17) gives

$$\alpha^{*k} \sum_{n=1}^N (x_n^{*k} - l_k^*) \geq 0 \quad (23)$$

so from (22) and (23) we conclude that $\sum_{n=1}^N x_n^{*k} = l_k^*$. Hence for every $\boldsymbol{\alpha}^* \in \mathcal{A}^*$ we have that for all k

$$\sum_{n=1}^N x_n^{*k} = l_k^* \text{ or } \left[\sum_{n=1}^N x_n^{*k} < l_k^* \text{ and } \alpha_k = 0 \right] \quad (24)$$

so $(\mathbf{x}^*, \boldsymbol{\alpha}^*) \in \mathcal{N}_{\text{opt}}$.

Now let $(\mathbf{x}^*(\boldsymbol{\alpha}^*), \boldsymbol{\alpha}^*) \in \mathcal{N}_{\text{opt}}$. We want to show that $(\mathbf{x}^*(\boldsymbol{\alpha}^*), \boldsymbol{\alpha}^*)$ solves the VI in (17). Since $\boldsymbol{\alpha}^*$ satisfies (24) for all k then for every $\boldsymbol{\alpha} \in \mathbb{R}_+^K$

$$\left\langle \boldsymbol{\alpha} - \boldsymbol{\alpha}^*, \sum_{n=1}^N \mathbf{x}_n^*(\boldsymbol{\alpha}^*) - \mathbf{l}^* \right\rangle = \left\langle \boldsymbol{\alpha}, \sum_{n=1}^N \mathbf{x}_n^*(\boldsymbol{\alpha}^*) - \mathbf{l}^* \right\rangle \leq 0. \quad (25)$$

Additionally, Since $\mathbf{x}^*(\boldsymbol{\alpha}^*)$ is a NE then by Proposition 1.4.2 in (Facchinei & Pang, 2007) we have for all $\mathbf{x} \in \mathcal{X}$

$$\langle \mathbf{x} - \mathbf{x}^*, F(\mathbf{x}^*(\boldsymbol{\alpha}^*), \boldsymbol{\alpha}^*) \rangle \leq 0. \quad (26)$$

Hence $(\mathbf{x}^*(\boldsymbol{\alpha}^*), \boldsymbol{\alpha}^*)$ is a solution to the VI.

10. Proof of Lemma 2

We start by showing that a large enough $\boldsymbol{\alpha}_0$ leads to a NE where the total loads are below \mathbf{l}^* . Let

$$\mathcal{U}_n = \left\{ \mathbf{x}_n \mid 0 \leq x_n^k < \frac{l_k^*}{N}, \forall k \right\} \quad (27)$$

and let $\mathcal{X}' = \mathcal{X} \setminus \mathcal{U}_1 \times \dots \times \mathcal{U}_N$, which is a closed set as the difference of a closed and an open set. Since $r_n(\mathbf{x})$ is continuous on the compact set \mathcal{X}' then $\max_{n, \mathbf{x} \in \mathcal{X}'} r_n(\mathbf{x}) \leq M$ for some $M > 0$. If we choose $\alpha_0^k \geq 2N \frac{M}{l_k^*}$ for all k , then for some player n and for all $\mathbf{x} \in \mathcal{X}'$

$$u_n(\mathbf{x}) = r_n(\mathbf{x}) - \sum_{k=1}^K \alpha_0^k x_n^k \leq M \left(1 - 2 \sum_{k=1}^K \frac{N}{l_k^*} x_n^k \right) < 0. \quad (28)$$

Hence, no $\mathbf{x} \in \mathcal{X}'$ is a NE since by switching to $\mathbf{x}_n = \mathbf{0}$ player n receives $u_n(\mathbf{x}) = 0$. We conclude that $\mathbf{x}^*(\boldsymbol{\alpha}_0) \in \mathcal{U}_1 \times \dots \times \mathcal{U}_N$, so $\sum_{n=1}^N x_n^{*k}(\boldsymbol{\alpha}_0) \leq l_k^*$ for all k .

Next we use this $\boldsymbol{\alpha}_0$ to argue about the set of solutions to our VI in (17). For each $\boldsymbol{\alpha}$ we have

$$\begin{aligned} \langle \mathbf{x} - \mathbf{x}^*(\boldsymbol{\alpha}_0), F(\mathbf{x}, \boldsymbol{\alpha}) \rangle &+ \left\langle \boldsymbol{\alpha} - \boldsymbol{\alpha}_0, \sum_{n=1}^N \mathbf{x}_n - \mathbf{l}^* \right\rangle \stackrel{(a)}{\leq} \\ &\langle \mathbf{x} - \mathbf{x}^*(\boldsymbol{\alpha}_0), F(\mathbf{x}^*(\boldsymbol{\alpha}_0), \boldsymbol{\alpha}) \rangle + \\ &\left\langle \boldsymbol{\alpha} - \boldsymbol{\alpha}_0, \sum_{n=1}^N \mathbf{x}_n - \mathbf{l}^* \right\rangle = \\ &\langle \mathbf{x} - \mathbf{x}^*(\boldsymbol{\alpha}_0), F(\mathbf{x}^*(\boldsymbol{\alpha}_0), \boldsymbol{\alpha}_0) \rangle + \\ &\left\langle \sum_{n=1}^N (\mathbf{x}_n - \mathbf{x}_n^*(\boldsymbol{\alpha}_0)), \boldsymbol{\alpha}_0 - \boldsymbol{\alpha} \right\rangle + \\ &\left\langle \boldsymbol{\alpha} - \boldsymbol{\alpha}_0, \sum_{n=1}^N \mathbf{x}_n - \mathbf{l}^* \right\rangle \stackrel{(b)}{\leq} \\ &\left\langle \boldsymbol{\alpha} - \boldsymbol{\alpha}_0, \sum_{n=1}^N \mathbf{x}_n^*(\boldsymbol{\alpha}_0) - \mathbf{l}^* \right\rangle \end{aligned} \quad (29)$$

where (a) uses the monotonicity of $F(\mathbf{x}, \boldsymbol{\alpha})$ in \mathbf{x} and (b) follows since $\mathbf{x}^*(\boldsymbol{\alpha}_0)$ is a NE (Proposition 1.4.2 in (Facchinei & Pang, 2007)). Hence the set

$$\begin{aligned} L_{\geq} = &\left\{ (\mathbf{x}, \boldsymbol{\alpha}) \in \mathcal{X} \times \mathbb{R}_+^K \mid \langle \mathbf{x} - \mathbf{x}^*(\boldsymbol{\alpha}_0), F(\mathbf{x}, \boldsymbol{\alpha}) \rangle \right. \\ &\left. + \left\langle \boldsymbol{\alpha} - \boldsymbol{\alpha}_0, \sum_{n=1}^N \mathbf{x}_n - \mathbf{l}^* \right\rangle \geq 0 \right\} \end{aligned} \quad (30)$$

is bounded, since (29) shows that

$$L_{\geq} \subseteq \{(\mathbf{x}, \boldsymbol{\alpha}) \mid \mathbf{x} \in \mathcal{X}, \boldsymbol{\alpha} \in \mathcal{C}\} \quad (31)$$

where \mathcal{C} is the following bounded convex polytope

$$\begin{aligned} \mathcal{C} = &\left\{ \boldsymbol{\alpha} \in \mathbb{R}_+^K \mid \left\langle \boldsymbol{\alpha} - \boldsymbol{\alpha}_0, \sum_{n=1}^N \mathbf{x}_n^*(\boldsymbol{\alpha}_0) - \mathbf{l}^* \right\rangle \geq 0 \right\} = \\ &\left\{ \boldsymbol{\alpha} \in \mathbb{R}_+^K \mid \langle \boldsymbol{\alpha}, \mathbf{v} \rangle \leq \underbrace{\langle \boldsymbol{\alpha}_0, \mathbf{v} \rangle}_{\geq 0} \right\} \end{aligned} \quad (32)$$

where $v = \sum_{n=1}^N (l^* - x_n^*(\alpha_0)) \geq 0$. Therefore according to Proposition 2.2.3 in (Facchinei & Pang, 2007) the set of solutions to the VI in (17) is non-empty and compact, which by Lemma 1 is $\mathcal{N}_{\text{opt}} = \mathcal{X}^* \times \mathcal{A}^*$.

11. Proof of Lemma 3

Note that $\mathcal{X} \triangleq \mathcal{X}_1 \times \dots \times \mathcal{X}_N$ is closed and convex since \mathcal{X}_n is closed convex for each n . Also note that $F(x, \alpha)$ is Lipschitz continuous in x since it is continuously differentiable on the closed \mathcal{X} . Then since $F(x, \alpha)$ is strongly monotone on \mathcal{X} (given α), Theorem 2.3.3 in (Facchinei & Pang, 2007) states that for all $x \in \mathcal{X}$, for some $L_0 > 0$

$$\|x - x^*(\alpha)\| \leq L_0 \|x - \Pi_{\mathcal{X}}(x - F(x, \alpha))\|. \quad (33)$$

Hence for $x^*(\alpha_2)$, $x^*(\alpha_1)$ we have

$$\begin{aligned} & \|x^*(\alpha_2) - x^*(\alpha_1)\| \leq \\ & L_0 \|x^*(\alpha_2) - \Pi_{\mathcal{X}}(x^*(\alpha_2) - F(x^*(\alpha_2), \alpha_1))\| = \\ & L_0 \left\| x^*(\alpha_2) - \Pi_{\mathcal{X}}(x^*(\alpha_2) - F(x^*(\alpha_2), \alpha_1)) + \right. \\ & \quad \left. \Pi_{\mathcal{X}}(x^*(\alpha_2) - F(x^*(\alpha_2), \alpha_2)) - \right. \\ & \quad \left. \Pi_{\mathcal{X}}(x^*(\alpha_2) - F(x^*(\alpha_2), \alpha_2)) \right\| \stackrel{(a)}{=} \\ & L_0 \left\| \Pi_{\mathcal{X}}(x^*(\alpha_2) - F(x^*(\alpha_2), \alpha_2)) - \right. \\ & \quad \left. \Pi_{\mathcal{X}}(x^*(\alpha_2) - F(x^*(\alpha_2), \alpha_1)) \right\| \leq \\ & L_0 \|F(x^*(\alpha_2), \alpha_1) - F(x^*(\alpha_2), \alpha_2)\| = \\ & \quad \sqrt{N} L_0 \|\alpha_2 - \alpha_1\| \quad (34) \end{aligned}$$

where in (a) we used that

$$x^*(\alpha_2) - \Pi_{\mathcal{X}}(x^*(\alpha_2) - F(x^*(\alpha_2), \alpha_2)) = \mathbf{0} \quad (35)$$

which follows from Proposition 1.5.8 in (Facchinei & Pang, 2007).

12. Proof of Lemma 4

Let $\alpha_1, \alpha_2 \in \mathbb{R}_+^K$. Let $x_1^* = x^*(\alpha_1)$ and $x_2^* = x^*(\alpha_2)$. Since x_1^* is a NE, we have for every $x \in \mathcal{X}$ that (see Proposition 1.4.2 in (Facchinei & Pang, 2007)):

$$\langle x - x_1^*, F(x_1^*, \alpha_1) \rangle \leq 0 \quad (36)$$

so for $x = x_2^*$

$$\langle x_2^* - x_1^*, F(x_1^*, \alpha_1) \rangle \leq 0. \quad (37)$$

Since x_2^* is a NE, we have for $x \in \mathcal{X}$ that

$$\langle x - x_2^*, F(x_2^*, \alpha_2) \rangle \leq 0 \quad (38)$$

so for $x = x_1^*$

$$\langle x_1^* - x_2^*, F(x_2^*, \alpha_2) \rangle \leq 0. \quad (39)$$

By adding (37) and (39) we obtain

$$\langle x_2^* - x_1^*, F(x_2^*, \alpha_2) - F(x_1^*, \alpha_1) \rangle \geq 0. \quad (40)$$

Then

$$\begin{aligned} -\lambda \|x^*(\alpha_2) - x^*(\alpha_1)\|^2 &= -\lambda \|x_2^* - x_1^*\|^2 \stackrel{(a)}{\geq} \\ & \langle x_2^* - x_1^*, F(x_2^*, \alpha_1) - F(x_1^*, \alpha_1) \rangle \stackrel{(b)}{=} \\ & \langle x_2^* - x_1^*, F(x_2^*, \alpha_2) - F(x_1^*, \alpha_1) \rangle + \\ & \left\langle \sum_{n=1}^N (x_n^*(\alpha_2) - x_n^*(\alpha_1)), \alpha_2 - \alpha_1 \right\rangle \stackrel{(c)}{\geq} \\ & \left\langle \sum_{n=1}^N (x_n^*(\alpha_2) - x_n^*(\alpha_1)), \alpha_2 - \alpha_1 \right\rangle \quad (41) \end{aligned}$$

where (a) uses that $F(x, \alpha_1)$ is strongly monotone in x with parameter $\lambda > 0$, (b) uses the linearity of $F(x, \alpha)$ in α and (c) uses (40).

Now let $\alpha_1, \alpha_2 \in \mathcal{A}^*$ and let $x^*(\alpha_1), x^*(\alpha_2)$ be the corresponding NE. For every k ,

- If $\alpha_1^k = \alpha_2^k = 0$ then $(\alpha_2^k - \alpha_1^k) \sum_{n=1}^N (x_n^{*k}(\alpha_2) - x_n^{*k}(\alpha_1)) = 0$.
- If $\alpha_1^k > 0$ and $\alpha_2^k > 0$ then $\sum_{n=1}^N x_n^{*k}(\alpha_1) = \sum_{n=1}^N x_n^{*k}(\alpha_2) = l_k^*$ so $(\alpha_2^k - \alpha_1^k) \sum_{n=1}^N (x_n^{*k}(\alpha_2) - x_n^{*k}(\alpha_1)) = 0$.
- If $\alpha_1^k > 0$ and $\alpha_2^k = 0$ then $\sum_{n=1}^N x_n^{*k}(\alpha_2) < l_k^* = \sum_{n=1}^N x_n^{*k}(\alpha_1)$ so $(\alpha_2^k - \alpha_1^k) \sum_{n=1}^N (x_n^{*k}(\alpha_2) - x_n^{*k}(\alpha_1)) \geq 0$.
- If $\alpha_1^k = 0$ and $\alpha_2^k > 0$ then $\sum_{n=1}^N x_n^{*k}(\alpha_2) = l_k^* > \sum_{n=1}^N x_n^{*k}(\alpha_1)$ so $(\alpha_2^k - \alpha_1^k) \sum_{n=1}^N (x_n^{*k}(\alpha_2) - x_n^{*k}(\alpha_1)) \geq 0$.

We conclude that if $\alpha_1, \alpha_2 \in \mathcal{A}^*$ then

$$\left\langle \sum_{n=1}^N (x_n^*(\alpha_2) - x_n^*(\alpha_1)), \alpha_2 - \alpha_1 \right\rangle \geq 0 \quad (42)$$

which by (41) implies that $x^*(\alpha_2) = x^*(\alpha_1)$.

13. Proof of Lemma 5

First we bound the distance between \mathbf{x}_t and the new NE $\mathbf{x}^*(\alpha_t)$. With probability 1, for some constant $C_0 > 0$,

$$\begin{aligned} \|\mathbf{x}_t - \mathbf{x}^*(\alpha_t)\|^2 &= \\ &\|\mathbf{x}_t - \mathbf{x}^*(\alpha_{t-1}) + \mathbf{x}^*(\alpha_{t-1}) - \mathbf{x}^*(\alpha_t)\|^2 \stackrel{(a)}{\leq} \\ &\|\mathbf{x}_t - \mathbf{x}^*(\alpha_{t-1})\|^2 + \|\mathbf{x}^*(\alpha_t) - \mathbf{x}^*(\alpha_{t-1})\|^2 + \\ &2\|\mathbf{x}_t - \mathbf{x}^*(\alpha_{t-1})\| \|\mathbf{x}^*(\alpha_t) - \mathbf{x}^*(\alpha_{t-1})\| \stackrel{(b)}{\leq} \\ &\|\mathbf{x}_t - \mathbf{x}^*(\alpha_{t-1})\|^2 + \varepsilon_{t-1}^2 L^2 \left\| \sum_{n=1}^N \mathbf{x}_{n,t} - \mathbf{l}^* \right\|^2 \\ &+ 2\varepsilon_{t-1} L \left\| \sum_{n=1}^N \mathbf{x}_{n,t} - \mathbf{l}^* \right\| \|\mathbf{x}_t - \mathbf{x}^*(\alpha_{t-1})\| \stackrel{(c)}{\leq} \\ &\|\mathbf{x}_t - \mathbf{x}^*(\alpha_{t-1})\|^2 + C_0 \varepsilon_t \quad (43) \end{aligned}$$

where (a) is Cauchy-Schwarz, (b) follows from Lemma 3 and (c) uses that $\mathbf{x}_{n,t}$, \mathbf{x}_t and $\mathbf{x}^*(\alpha_{t-1})$ are bounded and that $\frac{\varepsilon_{t+1}}{\varepsilon_t} \rightarrow 1$ as $t \rightarrow \infty$ (condition 4 of Theorem 1).

Note that $\mathbf{1}_N \otimes \alpha_t$ concatenates α_t N times. Next we bound the norm of the stochastic gradient vector. With probability 1, we have that for some constants $B_0, B_1, B_2 > 0$,

$$\begin{aligned} \eta_t \|\mathbf{g}_t - \mathbf{1}_N \otimes \alpha_t\| &\leq \eta_t \left(\sqrt{N} \|\alpha_t\| + \|\mathbf{g}_t\| \right) \stackrel{(a)}{\leq} \\ \eta_t B_0 \left(\|\alpha_0\| + \sum_{\tau=0}^{t-1} \varepsilon_\tau \left\| \sum_{n=1}^N \mathbf{x}_{n,\tau+1} - \mathbf{l}^* \right\| + \|\mathbf{g}_t\| \right) &\leq \\ \eta_t B_1 \left(\sum_{\tau=0}^{t-1} \varepsilon_\tau + \|\mathbf{g}_t\| \right) &\stackrel{(b)}{\leq} B_2 (\sqrt{\varepsilon_t} + \eta_t \|\mathbf{g}_t\|) \quad (44) \end{aligned}$$

where (a) iterates over

$$\begin{aligned} \|\alpha_{t+1}\| &= \left\| \left[\alpha_t + \varepsilon_t \left(\sum_{n=1}^N \mathbf{x}_{n,t+1} - \mathbf{l}^* \right) \right]^+ \right\| \leq \\ &\|\alpha_t\| + \varepsilon_t \left\| \sum_{n=1}^N \mathbf{x}_{n,t+1} - \mathbf{l}^* \right\| \quad (45) \end{aligned}$$

and (b) uses condition 3 of Theorem 1. To see that, let $\rho > 0$. Then pick a large enough T_0 such that for all $t > T_0$ we have $B_1 \frac{\eta_t \sum_{\tau=0}^{t-1} \varepsilon_\tau}{\sqrt{\varepsilon_t}} < \rho$. Hence we can use $B_2 = \max \left\{ B_1 \max_{0 \leq t \leq T_0} \frac{\eta_t \sum_{\tau=0}^{t-1} \varepsilon_\tau}{\sqrt{\varepsilon_t}}, \rho, B_1 \right\}$.

Now we can analyze the gradient behavior. Recall the definition of $F(\mathbf{x}, \alpha)$ in (16). Then, with probability 1, for

some constants $C_1, C_2, C_3 > 0$,

$$\begin{aligned} \eta_t \mathbb{E} \{ \langle \mathbf{x}_t - \mathbf{x}^*(\alpha_t), \mathbf{g}_t - \mathbf{1}_N \otimes \alpha_t \rangle \mid \mathcal{F}_t \} &= \\ \eta_t \mathbb{E} \{ \langle \mathbf{x}_t - \mathbf{x}^*(\alpha_{t-1}), \mathbf{g}_t - \mathbf{1}_N \otimes \alpha_t \rangle \mid \mathcal{F}_t \} &+ \\ + \eta_t \mathbb{E} \{ \langle \mathbf{x}^*(\alpha_{t-1}) - \mathbf{x}^*(\alpha_t), \mathbf{g}_t - \mathbf{1}_N \otimes \alpha_t \rangle \mid \mathcal{F}_t \} &\leq \\ \eta_t \langle \mathbf{x}_t - \mathbf{x}^*(\alpha_{t-1}), \mathbb{E} \{ \mathbf{g}_t - \mathbf{1}_N \otimes \alpha_t \mid \mathcal{F}_t \} - F(\mathbf{x}_t, \alpha_{t-1}) \rangle &+ \\ + \eta_t \langle \mathbf{x}_t - \mathbf{x}^*(\alpha_{t-1}), F(\mathbf{x}_t, \alpha_{t-1}) \rangle &+ \\ + \eta_t \mathbb{E} \{ \|\mathbf{x}^*(\alpha_t) - \mathbf{x}^*(\alpha_{t-1})\| \|\mathbf{g}_t - \mathbf{1}_N \otimes \alpha_t\| \mid \mathcal{F}_t \} &\stackrel{(a)}{\leq} \\ \sqrt{N} \eta_t \|\mathbf{x}_t - \mathbf{x}^*(\alpha_{t-1})\| \|\alpha_t - \alpha_{t-1}\| &+ \\ + \eta_t \|\mathbf{x}_t - \mathbf{x}^*(\alpha_{t-1})\| \|\mathbb{E} \{ \mathbf{g}_t \mid \mathcal{F}_t \} - F(\mathbf{x}_t) \| &- \\ - \lambda \eta_t \|\mathbf{x}_t - \mathbf{x}^*(\alpha_{t-1})\|^2 &+ \\ + C_1 \varepsilon_t \eta_t \mathbb{E} \{ \|\mathbf{g}_t - \mathbf{1}_N \otimes \alpha_t\| \mid \mathcal{F}_t \} &\stackrel{(b)}{\leq} \\ \eta_t \delta_t \|\mathbf{x}_t - \mathbf{x}^*(\alpha_{t-1})\| - \lambda \eta_t \|\mathbf{x}_t - \mathbf{x}^*(\alpha_{t-1})\|^2 &+ \\ + C_2 \varepsilon_t^{3/2} + C_3 \eta_t \varepsilon_t \quad (46) \end{aligned}$$

where (a) uses that $\|\mathbf{x}^*(\alpha_t) - \mathbf{x}^*(\alpha_{t-1})\| \leq C_1 \varepsilon_t$ (Lemma 3 and $\frac{\varepsilon_{t+1}}{\varepsilon_t} \rightarrow 1$), and also that since $F(\mathbf{x}, \alpha_{t-1})$ is strongly monotone in \mathbf{x} with parameter $\lambda > 0$, then

$$\begin{aligned} \langle \mathbf{x}_t - \mathbf{x}^*(\alpha_{t-1}), F(\mathbf{x}_t, \alpha_{t-1}) \rangle &\stackrel{(a)}{\leq} \\ \langle \mathbf{x}_t - \mathbf{x}^*(\alpha_{t-1}), F(\mathbf{x}_t, \alpha_{t-1}) - F(\mathbf{x}^*(\alpha_{t-1}), \alpha_{t-1}) \rangle &+ \\ \leq -\lambda \|\mathbf{x}_t - \mathbf{x}^*(\alpha_{t-1})\|^2 \quad (47) \end{aligned}$$

where (a) follows since $\langle \mathbf{x} - \mathbf{x}^*(\alpha), F(\mathbf{x}^*(\alpha), \alpha) \rangle \leq 0$ for all $\alpha \in \mathbb{R}_+^K$ and $\mathbf{x} \in \mathcal{X}$, since $\mathbf{x}^*(\alpha)$ is a NE (see Proposition 1.4.2 in (Facchinei & Pang, 2007)). Inequality (b) in (46) follows from (44) and the assumption in Definition 3.

Now we can bound how the distance from NE evolves. Then, with probability 1, for some constants $C_4, C_5, C_6 > 0$,

$$\begin{aligned} \mathbb{E} \left\{ \|\mathbf{x}_{t+1} - \mathbf{x}^*(\alpha_t)\|^2 \mid \mathcal{F}_t \right\} &\stackrel{(a)}{\leq} \\ \mathbb{E} \left\{ \|\mathbf{x}_t + \eta_t (\mathbf{g}_t - \mathbf{1}_N \otimes \alpha_t) - \mathbf{x}^*(\alpha_t)\|^2 \mid \mathcal{F}_t \right\} &= \\ \mathbb{E} \left\{ \|\mathbf{x}_t - \mathbf{x}^*(\alpha_t)\|^2 + \eta_t^2 \|\mathbf{g}_t - \mathbf{1}_N \otimes \alpha_t\|^2 \mid \mathcal{F}_t \right\} &+ \\ + 2\eta_t \mathbb{E} \{ \langle \mathbf{x}_t - \mathbf{x}^*(\alpha_t), \mathbf{g}_t - \mathbf{1}_N \otimes \alpha_t \rangle \mid \mathcal{F}_t \} &\stackrel{(b)}{\leq} \\ \|\mathbf{x}_t - \mathbf{x}^*(\alpha_{t-1})\|^2 + C_0 \varepsilon_t + C_4 \left(\varepsilon_t + \eta_t^2 \mathbb{E} \left\{ \|\mathbf{g}_t\|^2 \mid \mathcal{F}_t \right\} \right) &+ \\ + 2\eta_t \delta_t \|\mathbf{x}_t - \mathbf{x}^*(\alpha_{t-1})\| - 2\lambda \eta_t \|\mathbf{x}_t - \mathbf{x}^*(\alpha_{t-1})\|^2 &+ \\ + 2C_2 \varepsilon_t^{3/2} + 2C_3 \eta_t \varepsilon_t &\stackrel{(c)}{\leq} \\ (1 - 2\eta_t (\lambda - \delta_t)) \|\mathbf{x}_t - \mathbf{x}^*(\alpha_{t-1})\|^2 + 2\eta_t \delta_t + C_5 \varepsilon_t + C_6 \eta_t^2 & \quad (48) \end{aligned}$$

where (a) uses $\|\Pi_{\mathcal{X}}\mathbf{y} - \mathbf{x}\| \leq \|\mathbf{y} - \mathbf{x}\|$ for any $\mathbf{x} \in \mathcal{X}$ since \mathcal{X} is convex. Inequality (b) uses (43), (44), and (46) and inequality (c) uses Definition 3 and $x \leq x^2 + 1$.

The last step of the proof is to use (48) to show by induction that for every $t \geq 1$

$$\mathbb{E} \left\{ \|\mathbf{x}_t - \mathbf{x}^*(\alpha_{t-1})\|^2 \right\} \leq A \frac{\varepsilon_t}{\eta_t} \quad (49)$$

for some $A > 0$. First we define T_0 to be large enough such that $\delta_t \leq \frac{\lambda}{4}$, $\max\{\eta_t^2, \eta_t \delta_t\} \leq C_7 \varepsilon_t$ for some $C_7 > 0$ and also that $\frac{\varepsilon_t - \varepsilon_{t+1}}{\eta_t} \leq \lambda \varepsilon_t$ for all $t > T_0$ (using conditions 1,2,4 of Theorem 1). Then we pick $A = \max \left\{ \max_{1 \leq t \leq T_0} \frac{\eta_t}{\varepsilon_t} \mathbb{E} \left\{ \|\mathbf{x}_t - \mathbf{x}^*(\alpha_{t-1})\|^2 \right\}, A_0 \right\}$ for some A_0 that is specified below, which is a constant with respect to t . Hence for all $1 \leq t \leq T_0$ (49) holds. For $t > T_0$ we take the expectation on both sides of (48) to obtain

$$\begin{aligned} & \mathbb{E} \left\{ \|\mathbf{x}_{t+1} - \mathbf{x}^*(\alpha_t)\|^2 \right\} \leq \\ & (1 - 2\eta_t(\lambda - \delta_t)) \mathbb{E} \left\{ \|\mathbf{x}_t - \mathbf{x}^*(\alpha_{t-1})\|^2 \right\} \\ & + 2\eta_t \delta_t + C_5 \varepsilon_t + C_6 \eta_t^2 \stackrel{(a)}{\leq} \left(1 - \frac{3}{2}\eta_t \lambda\right) A \frac{\varepsilon_t}{\eta_t} + D_0 \varepsilon_t = \\ & A \frac{\varepsilon_t}{\eta_t} + \left(D_0 - \frac{3}{2}\lambda A\right) \varepsilon_t \stackrel{(b)}{\leq} A \left(\frac{\varepsilon_t}{\eta_t} - \lambda \varepsilon_t\right) \stackrel{(c)}{\leq} A \frac{\varepsilon_{t+1}}{\eta_{t+1}} \end{aligned} \quad (50)$$

where (a) follows for some constant $D_0 > 0$ since $\delta_t \leq \frac{\lambda}{4}$ and $\max\{\eta_t^2, \eta_t \delta_t\} \leq C_7 \varepsilon_t$ for $t > T_0$. In (b) we used $A \geq \frac{2D_0}{\lambda}$ so we set $A_0 = \frac{2D_0}{\lambda}$ and in (c) we used that $\frac{\varepsilon_t - \varepsilon_{t+1}}{\eta_t} \leq \lambda \varepsilon_t$ so $\frac{\varepsilon_{t+1}}{\eta_{t+1}} \geq \frac{\varepsilon_t}{\eta_t} - \lambda \varepsilon_t$.

14. Proof of Theorem 1

We have that with probability 1

$$\begin{aligned} & \min_{\alpha^* \in \mathcal{A}^*} \|\alpha_{t+1} - \alpha^*\|^2 = \\ & \min_{\alpha^* \in \mathcal{A}^*} \left\| \left[\alpha_t + \varepsilon_t \left(\sum_{n=1}^N \mathbf{x}_{n,t+1} - \mathbf{l}^* \right) \right]^+ - \alpha^* \right\|^2 \stackrel{(a)}{\leq} \\ & \min_{\alpha^* \in \mathcal{A}^*} \left\| \alpha_t + \varepsilon_t \left(\sum_{n=1}^N \mathbf{x}_{n,t+1} - \mathbf{l}^* \right) - \alpha^* \right\|^2 = \\ & \min_{\alpha^* \in \mathcal{A}^*} \left[\|\alpha_t - \alpha^*\|^2 + 2\varepsilon_t \left\langle \sum_{n=1}^N \mathbf{x}_{n,t+1} - \mathbf{l}^*, \alpha_t - \alpha^* \right\rangle \right] \\ & \quad + \varepsilon_t^2 \left\| \sum_{n=1}^N \mathbf{x}_{n,t+1} - \mathbf{l}^* \right\|^2 \stackrel{(b)}{\leq} \\ & \min_{\alpha^* \in \mathcal{A}^*} \left[\|\alpha_t - \alpha^*\|^2 + 2\varepsilon_t \left\langle \sum_{n=1}^N (\mathbf{x}_{n,t+1} - \mathbf{x}_n^*(\alpha_t)), \alpha_t - \alpha^* \right\rangle \right] \\ & \quad + 2\varepsilon_t \left\langle \sum_{n=1}^N \mathbf{x}_n^*(\alpha_t) - \mathbf{x}_n^*(\alpha^*), \alpha_t - \alpha^* \right\rangle \\ & \quad + 2\varepsilon_t \left\langle \sum_{n=1}^N \mathbf{x}_n^*(\alpha^*) - \mathbf{l}^*, \alpha_t - \alpha^* \right\rangle + \varepsilon_t^2 D_1 \stackrel{(c)}{\leq} \\ & \min_{\alpha^* \in \mathcal{A}^*} \left[\|\alpha_t - \alpha^*\|^2 + 2\varepsilon_t \left\| \sum_{n=1}^N (\mathbf{x}_{n,t+1} - \mathbf{x}_n^*(\alpha_t)) \right\| \|\alpha_t - \alpha^*\| \right] \\ & \quad + 2\varepsilon_t \left\langle \sum_{n=1}^N (\mathbf{x}_n^*(\alpha_t) - \mathbf{x}_n^*(\alpha^*)), \alpha_t - \alpha^* \right\rangle \\ & \quad + 2\varepsilon_t \left\langle \sum_{n=1}^N \mathbf{x}_n^*(\alpha^*) - \mathbf{l}^*, \alpha_t - \alpha^* \right\rangle + \varepsilon_t^2 D_1 \stackrel{(d)}{\leq} \\ & \left(1 + 2\varepsilon_t \left\| \sum_{n=1}^N (\mathbf{x}_{n,t+1} - \mathbf{x}_n^*(\alpha_t)) \right\| \right) \min_{\alpha^* \in \mathcal{A}^*} \|\alpha_t - \alpha^*\|^2 \\ & \quad + 2\varepsilon_t \max_{\alpha^* \in \mathcal{A}^*} \left[\left\langle \sum_{n=1}^N (\mathbf{x}_n^*(\alpha_t) - \mathbf{x}_n^*(\alpha^*)), \alpha_t - \alpha^* \right\rangle \right. \\ & \quad \left. + \left\langle \sum_{n=1}^N \mathbf{x}_n^*(\alpha^*) - \mathbf{l}^*, \alpha_t - \alpha^* \right\rangle \right] \\ & \quad + 2\varepsilon_t \left\| \sum_{n=1}^N (\mathbf{x}_{n,t+1} - \mathbf{x}_n^*(\alpha_t)) \right\| + \varepsilon_t^2 D_1 \quad (51) \end{aligned}$$

where (a) follows since $[\mathbf{x}]^+$ can only decrease the distance of \mathbf{x} to the set \mathcal{A}^* since $\alpha^* \geq 0$ for all $\alpha^* \in \mathcal{A}^*$. Inequality (b) uses that $\left\| \sum_{n=1}^N \mathbf{x}_{n,t+1} - \mathbf{l}^* \right\|^2 \leq D_1$ for some $D_1 > 0$ since \mathcal{X} is bounded, (c) is Cauchy-Schwarz and (d) uses that $\min_{\mathbf{x}} (f(\mathbf{x}) + g(\mathbf{x})) \leq \min_{\mathbf{x}} f(\mathbf{x}) + \max_{\mathbf{x}} g(\mathbf{x})$ for any functions $f(\mathbf{x}), g(\mathbf{x})$ and then uses $\|\alpha_t - \alpha^*\| \leq$

$\|\alpha_t - \alpha^*\|^2 + 1$. Define

$$f(\alpha) = -\max_{\alpha^* \in \mathcal{A}^*} \left[\left\langle \sum_{n=1}^N (\mathbf{x}_n^*(\alpha) - \mathbf{x}_n^*(\alpha^*)), \alpha - \alpha^* \right\rangle + \left\langle \sum_{n=1}^N \mathbf{x}_n^*(\alpha^*) - \mathbf{l}^*, \alpha - \alpha^* \right\rangle \right]. \quad (52)$$

Then $\left\langle \sum_{n=1}^N (\mathbf{x}_n^*(\alpha) - \mathbf{x}_n^*(\alpha^*)), \alpha - \alpha^* \right\rangle \leq 0$ by Lemma 4 and $\left\langle \sum_{n=1}^N \mathbf{x}_n^*(\alpha^*) - \mathbf{l}^*, \alpha - \alpha^* \right\rangle \leq 0$ by Lemma 1, for any $\alpha^* \in \mathcal{A}^*$, so $f(\alpha) \geq 0$ for all $\alpha \in \mathbb{R}_+^K$.

Define the filtration $\mathcal{G}_t \triangleq \mathcal{F}_{t+1}$. By taking the conditional expectation on both sides of (51):

$$\begin{aligned} & \mathbb{E} \left\{ \min_{\alpha^* \in \mathcal{A}^*} \|\alpha_{t+1} - \alpha^*\|^2 \mid \mathcal{G}_t \right\} \leq \\ & \mathbb{E} \left\{ (1 + 2N\varepsilon_t \|\mathbf{x}^*(\alpha_t) - \mathbf{x}_{t+1}\|) \min_{\alpha^* \in \mathcal{A}^*} \|\alpha_t - \alpha^*\|^2 \mid \mathcal{G}_t \right\} \\ & - 2\varepsilon_t f(\alpha_t) + 2N\varepsilon_t \mathbb{E} \left\{ \|\mathbf{x}^*(\alpha_t) - \mathbf{x}_{t+1}\| \mid \mathcal{G}_t \right\} + \varepsilon_t^2 D_1 = \\ & (1 + 2N\varepsilon_t \|\mathbf{x}^*(\alpha_t) - \mathbf{x}_{t+1}\|) \min_{\alpha^* \in \mathcal{A}^*} \|\alpha_t - \alpha^*\|^2 \\ & - 2\varepsilon_t f(\alpha_t) + 2N\varepsilon_t \|\mathbf{x}^*(\alpha_t) - \mathbf{x}_{t+1}\| + \varepsilon_t^2 D_1. \quad (53) \end{aligned}$$

Next we want to apply Robbins-Siegmund ‘‘almost’’ super-martingale Theorem (Robbins & Siegmund, 1971) based on (53). Using their notation, we have $z_t = \min_{\alpha^* \in \mathcal{A}^*} \|\alpha_t - \alpha^*\|^2$, $\beta_t = 2N\varepsilon_t \|\mathbf{x}^*(\alpha_t) - \mathbf{x}_{t+1}\|$, $\xi_t = 2N\varepsilon_t \|\mathbf{x}^*(\alpha_t) - \mathbf{x}_{t+1}\| + \varepsilon_t^2 D_1$ and $\zeta_t = 2\varepsilon_t f(\alpha_t) \geq 0$.

From Lemma 5 we obtain for some $A > 0$ that

$$\begin{aligned} & \mathbb{E} \left\{ \sum_{t=0}^{\infty} \varepsilon_t \|\mathbf{x}^*(\alpha_t) - \mathbf{x}_{t+1}\| \right\} \stackrel{(a)}{\leq} \\ & \sum_{t=0}^{\infty} \varepsilon_t \mathbb{E} \left\{ \|\mathbf{x}^*(\alpha_t) - \mathbf{x}_{t+1}\| \right\} \leq A \sum_{t=0}^{\infty} \varepsilon_t \sqrt{\frac{\varepsilon_t}{\eta_t}} \stackrel{(b)}{<} \infty \end{aligned} \quad (54)$$

where (a) is Fatou’s Lemma (Billingsley, 2008) and (b) is condition 3 of the Theorem. Hence, $\sum_{t=0}^{\infty} \varepsilon_t \|\mathbf{x}^*(\alpha_t) - \mathbf{x}_{t+1}\| < \infty$ with probability 1 so $\sum_{t=0}^{\infty} \beta_t < \infty$ and $\sum_{t=0}^{\infty} \xi_t < \infty$, also using that $\sum_{t=0}^{\infty} \varepsilon_t^2 < \infty$. Then (Robbins & Siegmund, 1971) states that with probability 1

$$\lim_{t \rightarrow \infty} \min_{\alpha^* \in \mathcal{A}^*} \|\alpha_t - \alpha^*\|^2 = \Lambda \quad (55)$$

for some random variable Λ , and that with probability 1

$$\sum_{t=0}^{\infty} \varepsilon_t f(\alpha_t) < \infty. \quad (56)$$

Now we show that for every $\alpha \notin \mathcal{A}^*$ we have $f(\alpha) > 0$. If $f(\tilde{\alpha}) = 0$ then both the non-negative terms in (52) are zero, so Lemma 4 implies that if $f(\tilde{\alpha}) = 0$ then there must exist an $\alpha^* \in \mathcal{A}^*$ such that $\mathbf{x}^*(\tilde{\alpha}) = \mathbf{x}^*(\alpha^*)$ and

$$\begin{aligned} 0 &= \left\langle \sum_{n=1}^N \mathbf{x}_n^*(\alpha^*) - \mathbf{l}^*, \tilde{\alpha} - \alpha^* \right\rangle \stackrel{(a)}{=} \\ & \left\langle \sum_{n=1}^N \mathbf{x}_n^*(\alpha^*) - \mathbf{l}^*, \tilde{\alpha} \right\rangle = \left\langle \sum_{n=1}^N \mathbf{x}_n^*(\tilde{\alpha}) - \mathbf{l}^*, \tilde{\alpha} \right\rangle \end{aligned} \quad (57)$$

where (a) uses $\left\langle \sum_{n=1}^N \mathbf{x}_n^*(\alpha^*) - \mathbf{l}^*, \alpha^* \right\rangle = 0$ (Lemma 1). Hence, by invoking Lemma 1 again, (57) and $\sum_{n=1}^N \mathbf{x}_n^*(\tilde{\alpha}) = \sum_{n=1}^N \mathbf{x}_n^*(\alpha^*) \leq \mathbf{l}^*$ imply that $(\mathbf{x}^*(\tilde{\alpha}), \tilde{\alpha})$ solves the VI in (17), so $\tilde{\alpha} \in \mathcal{A}^*$.

Let $\omega \in \Omega$ such that $\Lambda(\omega) > 0$, if exists. Then there exist $T_0(\omega)$ and $a(\omega), b(\omega) > 0$ such that for all $t \geq T_0(\omega)$

$$\alpha_t \in \mathcal{R}(\omega) = \left\{ \alpha \mid a(\omega) \leq \min_{\alpha^* \in \mathcal{A}^*} \|\alpha - \alpha^*\|^2 \leq b(\omega) \right\}. \quad (58)$$

Since \mathcal{A}^* is compact (Lemma 2 shows that $\mathcal{X}^* \times \mathcal{A}^*$ is compact) then $g(\alpha) = \min_{\alpha^* \in \mathcal{A}^*} \|\alpha - \alpha^*\|^2$ is continuous, which makes $\mathcal{R}(\omega)$ compact as well.

Since $\mathbf{x}^*(\alpha)$ is continuous (Lemma 3), then $f(\alpha)$ is continuous as a maximum of continuous functions over the compact \mathcal{A}^* (Lemma 2). Hence, $\min_{\alpha \in \mathcal{R}(\omega)} f(\alpha) = c$ for

some $c > 0$. But then $\alpha_t \in \mathcal{R}(\omega)$ for all $t \geq T_0(\omega)$ implies that $\sum_{t=0}^{\infty} \varepsilon_t f(\alpha_t) = \infty$, which by (56) cannot occur for more than a measure zero set of $\omega \in \Omega$. We conclude that $\min_{\alpha^* \in \mathcal{A}^*} \|\alpha_t - \alpha^*\|^2 \rightarrow 0$ as $t \rightarrow \infty$ with probability 1.

Let $\varepsilon > 0$. The above implies that for almost all $\omega \in \Omega$ there exists $T(\omega)$ such that for all $t > T(\omega)$ there is a sequence $\alpha_t^*(\omega) \in \mathcal{A}^*$ such that

$$\|\mathbf{x}^*(\alpha_t) - \mathbf{x}^*(\alpha_t^*(\omega))\| \stackrel{(a)}{\leq} L \|\alpha_t - \alpha_t^*(\omega)\| \leq \varepsilon \quad (59)$$

where (a) follows from the Lipschitz continuity of $\mathbf{x}^*(\alpha)$ (Lemma 3). However, by Lemma 4 we know that $\mathbf{x}^*(\alpha) = \mathbf{x}^*$ for all $\alpha \in \mathcal{A}^*$ for some NE \mathbf{x}^* such that for all k

$$\sum_{n=1}^N x_n^{*k} = l_k^* \text{ or } \left[\sum_{n=1}^N x_n^{*k} < l_k^* \text{ and } \alpha_k = 0 \right]. \quad (60)$$

Finally, we conclude that

$$\begin{aligned} \lim_{t \rightarrow \infty} \mathbb{E} \left\{ \|\mathbf{x}_t - \mathbf{x}^*\|^2 \right\} &\leq 2 \lim_{t \rightarrow \infty} \mathbb{E} \left\{ \|\mathbf{x}_t - \mathbf{x}^*(\alpha_{t-1})\|^2 \right\} \\ &+ 2 \lim_{t \rightarrow \infty} \mathbb{E} \left\{ \|\mathbf{x}^* - \mathbf{x}^*(\alpha_{t-1})\|^2 \right\} \stackrel{(a)}{=} 0 \end{aligned} \quad (61)$$

where (a) uses Lemma 5 for the first term (conditions 1, 3 and 4 imply that $\frac{\varepsilon_t}{\eta_t} \rightarrow 0$) and (59) for the second term.