

MULTI-LEADER-FOLLOWER POTENTIAL GAMES

SONJA STEFFENSEN¹

Abstract. In this paper, we discuss a particular class of Nash games, where the participants of the game (the players) are divided into two groups (leaders and followers) according to their position or influence on the other players. Moreover, we consider the case, when the leaders' and/or the followers' game can be described as a potential game. This is a subclass of Nash games that has been introduced by Monderer and Shapley in 1996 and has beneficial properties to reformulate the bilevel Nash game. We develop necessary and sufficient conditions for Nash equilibria and present existence and uniqueness results. Furthermore, we discuss some Examples to illustrate our results.

INTRODUCTION

As an important mathematical modeling tool to analytically study the strategic decision making in a competitive environment, game theory has attracted the interest of researchers of various fields such as economics, political sciences, management, computer science and biology.

In classical game theory a *game* describes the situation, where several individuals make a (strategic) choice by taking into account the choices of the others. Moreover, the fundamental elements of a game are the individuals (so-called *players*) of the game, the strategies that are available to each player and the individual payoff functions. Furthermore, games are classified into *cooperative* and *noncooperative* games. Here we will discuss a particular type of noncooperative games, where each individual player is only concerned with his/her own objective, which leads to the well-known concept of a Nash game or equilibrium, in contrast to e.g. Pareto optimality, i.e. the solution of a multi-objective optimization problem which represents a kind of cooperative game.

In the following, we consider so-called multi-leader-follower games (MLFG) that form a particular subclass of noncooperative problems, where the players are divided into two groups, namely leaders and followers, according to their influence (position) in the game. Mathematically, this yields a hierarchical game. However, in contrast to the well-known Stackelberg game, where one single leader is accompanied by one or more followers, MLFGs model the situation where several leaders exist. Games of this structure have recently attained the interest of mathematicians as well as scientists of related fields such as operations research, robotics, and computer science [1, 9, 20]. However, there is still a lack of theoretical results concerning the structure of such games, such as existence and uniqueness theory, characterization of equilibria etc.. Here we consider the following multi-leader-follower Nash game:

Let the leaders' game be given by

$$\min_{x_\nu \in \mathbb{R}^{n_\nu}} \theta_\nu(x_\nu, x_{-\nu}, y) = \varphi_\nu(x_\nu) + \phi(x, y) \quad \text{s.t.} \quad x_\nu \in X_\nu \subseteq \mathbb{R}^{n_\nu}, \quad \text{for} \quad \nu = 1, \dots, N \quad (1)$$

¹ IGPM, RWTH Aachen University, Aachen, Germany, steffensen@igpm.rwth-aachen.de

where x denotes the leaders' joint strategy vector and $x_{-\nu}$ is defined by $x_{-\nu} = (x_1, \dots, x_{\nu-1}, x_{\nu+1}, \dots, x_N)$. Furthermore, the joint strategy vector of the followers $y = (y_j, y_{-j}) \in \mathbb{R}^{mM}$ (for simplicity we chose $m_j = m$) is obtained through the Nash game modeled by the optimization problems

$$\min_{y_j} \tau_j(y, x) \quad \text{s.t.} \quad y_j \in Y_j, \quad \text{for} \quad j = 1, \dots, M, \quad (2)$$

with nonempty, convex, and closed strategy sets $Y_j \subseteq \mathbb{R}^m$. The multi-leader-follower game is then given by

$$\left. \begin{array}{l} \min_{x_\nu \in \mathbb{R}^{n_\nu}} \theta_\nu(x_\nu, x_{-\nu}, y) \\ \text{s.t.} \quad x_\nu \in X_\nu, \\ \min_{y_j \in \mathbb{R}^m} \tau_j(y, x) \quad \text{s.t.} \quad y_j \in Y_j \quad \text{for} \quad j = 1, \dots, M \end{array} \right\} \quad \text{for} \quad \nu = 1, \dots, N. \quad (3)$$

Note, that if the solution of the lower-level problem (the Nash game of the followers) is not single-valued, we consider the *optimistic* case of a bilevel problem (cf. also [4]), since for each leader ν we take the followers response $y \in \mathcal{S}(x)$ that minimizes the leader's objective function value θ_ν .

In [12] A. Kulkarni and U. Shanbhag present an example of a similar type in the context of congestions control in communication networks. Here, the users of the network are the leaders that decide about the flow rates x on the network. Moreover, a single follower, namely the network manager solves an optimization problem of the form

$$\min_y \tau(y, x) \quad \text{s.t.} \quad y \in Y(x),$$

where his/her strategy y represent the decisions that have to be made on the network, such as network capacities, flow specifications etc.. Note, that the feasibility regions for the decisions y also depend on the users' decision x . Furthermore, the users' problems are given by

$$\min_{x_\nu \in \mathbb{R}^{n_\nu}} U_\nu(x_\nu) - c(y) \quad \text{s.t.} \quad x_\nu \in X_\nu,$$

where $U_\nu(x_\nu)$ denotes the utility function of player ν and $c(y)$ denote the congestion cost associated to the network manager's decision y , which is imposed on every user of the network. Other research on multi-leader-follower games include their paper [11], where they consider the setting for multi-leader-follower games that can be reformulated as a Nash game with shared constraints.

Further theoretical work on leader-follower games include the early work by Sherali [16] that generalizes Stackelberg games in the setting of Cournot competition and the more recent paper [7], where Hu and Fukushima discuss existence and uniqueness results of robust Nash equilibria for a class of quadratic MLFGs. Numerical schemes for multi-leader-follower games have e.g. been presented in [8, 13, 17, 19].

This paper is organized as follows. In Section 1 we discuss the followers' problem, which then gives rise to a reformulation of the leaders' game, i.e. the full multi-leader-follower game. The reformulation and the subsequent analysis thereof is then presented in Section 2.

1. THE FOLLOWERS' GAME

In this section, we are concerned with the *lower-level* problem, i.e. the followers' game (2). In case that $M = 1$, we have a single follower and the game reduces to a *lower-level* minimization problem. In case that M is larger than one, a common solution concept for the resulting Nash game defined by (2) is the *Nash equilibrium* that was introduced by J.F. Nash in the 1950th [15].

Definition 1.1 (Nash Equilibrium). Consider the game (2) and let $\bar{x} \in X$ be fixed. Then, a joint strategy vector $y^* \in Y$ is a *Nash equilibrium*, if the following condition holds for all $j = 1, \dots, M$

$$NEP(x) : \quad \tau_j(y_j^*, y_{-j}^*, \bar{x}) \leq \tau_j(y_j, y_{-j}^*, \bar{x}) \quad \text{for all} \quad y_j \in Y_j \quad (4)$$

Hence, a Nash equilibrium corresponds to a multistrategy vector of all players, where no player of the game has the incentive to change his/her chosen strategy unilaterally.

If the followers' objective functions τ_j are convex and continuously differentiable and the feasible sets Y_j are also convex, each follower's minimization is a convex problem, i.e. a necessary and sufficient condition for (4) to hold for each follower j is given by the inequality [2]

$$\nabla_{y_j} \tau_j(y_j^*, y_{-j}^*, \bar{x})^T (z_j - y_j^*) \geq 0 \quad \forall z_j \in Y_j. \tag{5}$$

Define the $f(y, x) = (\nabla_{y_1} \tau_1(y, x)^T, \dots, \nabla_{y_M} \tau_M(y, x)^T)^T$ and obtain the following characterization of the set of Nash equilibria of the parameterized Nash game (2).

Theorem 1.2. *Assume that Y_j is nonempty, closed and convex for any $j = 1, \dots, M$. Let for all $j = 1, \dots, M$ the objective τ_j be continuously differentiable in x and y on an open set $\Omega \supseteq X \times Y$ and convex in y_j for any feasible y_{-j} and any feasible, fixed $\bar{x} \in X$. Then $y^*(\bar{x})$ is a Nash equilibrium of $NEP(\bar{x})$, iff $y^*(\bar{x})$ solves the variational inequality*

$$f(y, \bar{x})^T (z - y) \geq 0 \quad \forall z \in Y = \prod_{j=1}^M Y_j. \tag{6}$$

The proof of the theorem is a standard result and can e.g. be found in [5] (Prop 1.4.2).

Proposition 1.3. *Assume, that Y_j is nonempty, closed and convex for any $j = 1, \dots, M$. Let $\bar{x} \in X$ be fixed and assume that for all $j = 1, \dots, M$ the objective $\tau_j(\cdot, \bar{x})$ is twice continuously differentiable on an open set $\Omega \supseteq Y$ and convex in y_j for any feasible y_{-j} . Furthermore, assume that the Jacobian $D_y f(y, \bar{x})$ is uniformly positive definite for all $y \in Y$. Then, the Nash game $NEP(\bar{x})$ admits a unique Nash equilibrium.*

Proof. Since the assumptions of Theorem 1.2 are satisfied, for any feasible and fixed $\bar{x} \in X$, the set of Nash equilibria of (2) coincides with the solution set of the variational inequality (6). Moreover, f is strongly monotone on Y , since $D_y f(y, \bar{x})$ is uniformly positive definite (cf. [5], Prop. 2.3.2). Hence, we can apply Theorem 2.3.3 in [5] to obtain the result. \square

The assumptions of this existence and uniqueness result guides us to consider the case, where each individual follower's objective τ_j can be replaced by a joint objective function. The particular structure leads to the subclass of Nash games that was first introduced by Monderer and Shapley in [14].

Definition 1.4 (Potential Game/ Potential). Consider the game (2) and let $\bar{x} \in X$ be fixed. If there exists a function $\pi(y, x)$ such that the following condition holds for all $j = 1, \dots, M$

$$\tau_j(y_j, y_{-j}, \bar{x}) - \tau_j(z_j, y_{-j}, \bar{x}) = \pi(y_j, y_{-j}, \bar{x}) - \pi(z_j, y_{-j}, \bar{x}) \quad \text{for all } z_j \in Y_j, \tag{7}$$

the game (2) is called a *potential game* and π is called an *exact potential function* (*potential*) of (2).

Now, if the followers' Nash game (2) is a potential game, i.e. there exists a potential function π that satisfies (7), τ_j can be replaced in (2) by π for all $j = 1, \dots, M$. Hence, the symmetry of $D_y f(y, \bar{x}) = D_y^2 \pi(y, \bar{x})$ is directly given and moreover, if π is uniformly convex in y , $D_y f(y, \bar{x})$ is uniformly positive definite on Y and therefore f is strongly monotone on Y . Thus we have the following corollary.

Corollary 1.5. *Assume, that Y_j is nonempty, closed and convex for any $j = 1, \dots, M$. Let (2) be a potential game with potential $\pi(y, x)$. Moreover, assume that π is twice continuously differentiable in x and y on an open set $\Omega \supseteq X \times Y$ and uniformly convex in y for any feasible, fixed $\bar{x} \in X$. Then, the Nash game $NEP(\bar{x})$ admits a unique Nash equilibrium $y^*(\bar{x})$.*

Proof. First we can apply Lemma 2.1 in [14]. Then, replacing τ_j in (5) by π , (6) has to hold for $f = D_y \pi$. Thus, $D_y f(y, \bar{x}) = D_y^2 \pi(y, \bar{x})$. Next, since the assumptions of Proposition 1.3 are all satisfied, we obtain the result. \square

The result of the corollary also leads to another advantage of the particular structure of potential games. In fact, the set of Nash equilibria of (2) corresponds to the set of global optima of the convex potential function on the joint feasible set Y .

Proposition 1.6. *Let the assumptions of Corollary 1.5 be satisfied. Then $y^*(\bar{x})$ is a Nash equilibrium of (2) if and only if $y^*(\bar{x})$ is a global optimum of the convex minimization problem*

$$\min_{y \in Y} \pi(y, \bar{x}). \quad (8)$$

Proof. Since the conditions of Corollary 1.5 are satisfied, the variational inequality (6) has to hold for $f = D\pi$. Thus, the necessary and sufficient conditions for a global optimum of the convex minimization problem (8) hold. \square

Remark 1.7. If the convex sets Y_j can be described by a finite number of affine linear equalities $h_j(x) = 0$ and inequalities $g_j(y_j) \leq 0$ with continuously differentiable, convex functions $g_{j,k}$ and some regularity conditions such as the Slater constraint qualification are satisfied, one could also make use of the KKT-conditions associated with (6) or the minimization problem (8), respectively.

Moreover, an equivalent condition for the variational inequality (6) to hold, which is independent of the assumptions on π , is given as follows.

Proposition 1.8. *Let the feasible set Y be nonempty, closed and convex. Then $y^*(\bar{x}) \in Y$ solves the variational inequality (6) if and only if it solves the equation*

$$y - \text{Proj}_Y(y - f(y, \bar{x})) = 0. \quad (9)$$

Next, let us consider some special cases with regard to the followers' strategy sets Y_j and their objective functions τ_j .

Example 1.9.

First, assume that the sets Y_j are polyhedrons, i.e. they can be defined by a finite number of affine linear mappings

$$Y_j = \{y_j \in \mathbb{R}^{m_j} \mid A_j y_j = b_j\} \quad \text{for all } j = 1, \dots, M$$

Next, let $\hat{y}_j \in Y_j$ and define a matrix T_j such that the range of T_j recovers the nullspace of A_j for all $j = 1, \dots, M$, i.e. any $y_j \in Y_j$ satisfies $y_j = T_j z_j + \hat{y}_j$ for some z_j . If T_j is defined such that it has full rank, then $z_j \in \mathbb{R}^{m_j}$ with $m_j = m - \text{rk}(A_j)$. This variable transformation then yields a Nash game in reduced form

$$\min_{z \in \mathbb{R}^K} \tilde{\pi}(z, \bar{x}) = \pi(Tz + \hat{y}, \bar{x}), \quad (10)$$

for a suitable chosen matrix $T = \text{diag}(T_j)_{j=1}^M$, a vector $\hat{y} = (\hat{y}_j)_{j=1}^M$ and $K = \sum_{j=1}^M m_j$. Since the differentiability and convexity properties of π transfer to $\tilde{\pi}$ the necessary and sufficient optimality conditions

$$\nabla_z \tilde{\pi}(z, \bar{x}) = 0 \quad (11)$$

guarantee $z(\bar{x})$ to be the possibly unique minimizer of $\tilde{\pi}(\cdot, \bar{x})$. Hence, since this yields an implicit or even explicit expression for the continuous (smooth) path $y^*(\bar{x}) = Tz^*(\bar{x}) + \hat{y}$, the followers Nash game can be substituted by equation (11) or y can directly be replaced by $y^*(\bar{x})$ in the leaders' objective function $\theta(x, y)$.

Example 1.10. Next, let each strategy set Y_j ($j = 1, \dots, M$) be the nonnegative orthant, i.e. $Y_j = \mathbb{R}_+^m$. Then, under the conditions of Proposition 1.6, the Nash equilibrium $y^*(\bar{x})$ of (2) is given by the solution of the Lipschitz-continuous, but nonsmooth equation (cf. Proposition 1.8):

$$y - \text{Proj}_{\mathbb{R}_+^m}(y - f(y, \bar{x})) = 0 \quad \Leftrightarrow \quad y - \max(0, y - f(y, \bar{x})) = 0$$

This can be reformulated as the complementarity condition: $\min(y, f(y, \bar{x})) = 0 \quad \Leftrightarrow \quad 0 \leq y \perp f(y, \bar{x}) \geq 0$.

If we assume in addition that each follower’s objective function is given as a quadratic function of the form

$$\tau_j(y_j, y_{-j}, \bar{x}) = \frac{1}{2} y_j^T Q_j y_j - b(\bar{x})^T y_j,$$

with a diagonal positive definite matrix Q_j , then we obtain $y_j^*(\bar{x}) = \max(0, Q_j^{-1} b(\bar{x}))$ for all $j = 1, \dots, M$ (see also [6, 18]). Moreover, the exact potential function π in this case is given by

$$\pi(y, x) = \frac{1}{2} \sum_{j=1}^M (y_j^T Q_j y_j - b(x)^T y_j)$$

Thus, the associated Nash game is in fact a potential game.

Example 1.11. In the more general case where τ_j is given by $\tau_j = \alpha_j(y_j) + \beta(y, x)$ for some continuously differentiable functions α_j and β , we have

$$\pi(y, x) = \sum_{j=1}^M \alpha_j(y_j) + \beta(y, x)$$

since for all $j = 1, \dots, M$ it holds

$$\pi(y_j, y_{-j}, \bar{x}) - \pi(z_j, y_{-j}, \bar{x}) = \tau_j(y_j, y_{-j}, \bar{x}) - \tau_j(z_j, y_{-j}, \bar{x}) \quad \text{for all } z_j \in Y_j.$$

Moreover, if each Y_j ($j = 1, \dots, M$) is defined by $Y_j = \{y_j \in \mathbb{R}^m \mid y_j \geq l_j(x)\}$, the Nash equilibrium is described by the complementarity condition

$$\min(y - l_j(x), f(y, \bar{x})) = 0, \quad \text{where } f(y, \bar{x}) = \nabla_y \pi(y, \bar{x}) = (\nabla_{y_j} \alpha_j(y_j))_{j=1}^M + \nabla_y \beta(y, \bar{x}).$$

2. THE LEADERS’ POTENTIAL GAME

In the previous section, we discussed under which conditions, the Nash game of the followers admits a Nash equilibrium, when it is unique and how it can be characterized. In this section we now use this information and focus on the Nash game played by the leaders.

Hence, assume that for any feasible leader multi strategy vector x , there exists at least one equilibrium of the followers’ Nash game, i.e. the set of equilibrium points $\mathcal{S}(x)$ of (2) is nonempty. Then the multi-leader-follower game (3) can be reformulated as

$$\left. \begin{array}{l} \min_{x_\nu \in \mathbb{R}^{\nu}} \theta_\nu(x_\nu, x_{-\nu}, y) \\ \text{s.t. } x_\nu \in X_\nu, \\ y \in \mathcal{S}(x) \end{array} \right\} \quad \text{for } \nu = 1, \dots, N \quad (12)$$

Moreover, if e.g. the conditions of Proposition 1.3 or Corollary 1.5, respectively, are satisfied, then $\mathcal{S}(x)$ is single-valued and we might replace the variable y in (12) by the path $y(x)$. This yields the single-level Nash

game

$$\left. \begin{array}{l} \min_{x_\nu \in \mathbb{R}^{n_\nu}} \quad \tilde{\theta}_\nu(x_\nu, x_{-\nu}) = \theta_\nu(x_\nu, x_{-\nu}, y(x)) \\ \text{s.t.} \quad x_\nu \in X_\nu, \end{array} \right\} \quad \text{for} \quad \nu = 1, \dots, N. \quad (13)$$

Remark 2.1. If $S(x)$ is not single-valued, a worst-case scenario might be considered which leads to the application of robust optimization tools such as e.g. presented in [7].

Next, assume that this Nash game forms a potential game, i.e. it admits a potential function $P(x)$. In particular, we consider the case, where the objective functions θ_ν can be decomposed into two parts: an individual part $\varphi_\nu(x_\nu)$ and a common part $\phi(x, y)$: for all $\nu = 1, \dots, N$ it holds

$$\theta_\nu(x_\nu, x_{-\nu}) = \varphi_\nu(x_\nu) + \phi(x, y)$$

and thus

$$\tilde{\theta}_\nu(x_\nu, x_{-\nu}) = \varphi_\nu(x_\nu) + \tilde{\phi}(x_\nu, x_{-\nu}),$$

where $\tilde{\phi}(x) = \phi(x, y(x))$. Then, the potential function $P(x)$ is given by

$$P(x_\nu, x_{-\nu}) = \sum_{\nu=1}^N \varphi_\nu(x_\nu) + \tilde{\phi}(x_\nu, x_{-\nu}), \quad (14)$$

as for all $\nu = 1, \dots, N$

$$\begin{aligned} P(x_\nu, x_{-\nu}) - P(s_\nu, x_{-\nu}) &= \varphi_\nu(x_\nu) + \tilde{\phi}(x_\nu, x_{-\nu}) - \varphi_\nu(s_\nu) - \tilde{\phi}(s_\nu, x_{-\nu}) \\ &= \tilde{\theta}_\nu(x_\nu, x_{-\nu}) - \tilde{\theta}_\nu(s_\nu, x_{-\nu}) \end{aligned}$$

holds for any $s_\nu \in X_\nu$. Due to this defining property of the potential function, we have the following result.

Theorem 2.2. *Let x^* be a global minimizer of the problem*

$$\min_x P(x) \quad \text{s.t.} \quad x \in X = \prod_{\nu=1}^N X_\nu. \quad (15)$$

Then x^ is a Nash-equilibrium of the MLFG (12).*

Proof. Assume that x^* is not a Nash equilibrium of (12). Then there exist a player ν that can benefit from unilaterally changing his/her strategy, i.e. there exist some $s_\nu \in X_\nu$ such that

$$0 < \tilde{\theta}_\nu(x_\nu, x_{-\nu}) - \tilde{\theta}_\nu(s_\nu, x_{-\nu}) = P(x_\nu, x_{-\nu}) - P(s_\nu, x_{-\nu})$$

which contradicts the global optimality of x^* for (15). \square

Next, having related the global optima of (15) and the Nash equilibria of (12), we consider the question, under which assumptions a global minimizer of (15) exists. The main assumption that we make here concerns the fact, that the solution of the lower-level is given by a continuous (but not necessarily smooth) solution mapping $y(x)$. In addition to this main assumption, the first result is based merely on the compactness assumption on the feasible set X , whereas the second one is based on the coercivity of P .

Corollary 2.3. *Assume that X is compact and that the solution path of the lower-level $y(x)$ is continuous on an open set $\Omega \supseteq X$. Then there exists at least one Nash equilibrium of the reformulated Nash game (12).*

Proof. Since by the assumptions the objective function P is continuous, the minimization problem (15) admits at least one global minimizer. Moreover, by Theorem 2.2, the Nash game (12) has at least one Nash equilibrium. \square

Corollary 2.4. *Let X be nonempty and closed, $y(x)$ being continuous on an open set $\Omega \supseteq X$ and P be coercive on X , i.e. for any sequence $(x^k) \subseteq X$ with $\lim_k \|x^k\| = +\infty$ it holds $\lim_k P(x^k) = +\infty$. Then, the set of Nash equilibria is nonempty and compact. In particular, there exists at least one Nash equilibrium of the reformulated Nash game (12).*

Proof. By assumption the set of global minimizers of (15) is nonempty and compact. Therefore, by Theorem 2.2, the Nash game (12) has at least one Nash equilibrium. \square

Next, we first discuss assumptions on the defining functions that guarantee the convexity of the objective function P , such that if X is convex (15) is a convex optimization problem.

Lemma 2.5. *Assume that for all $\nu = 1, \dots, N$ the functions φ_ν are convex. Let $\phi(x, y)$ be convex in x and y and nondecreasing in y . Moreover, suppose that $y(x)$ is a continuous, convex function. Then P is a continuous, convex function.*

Proof. The continuity of P is directly given by the continuity of the defining functions. Next, since $y(x)$ is supposed to be a convex function, we have $y(\lambda x_1 + (1 - \lambda)x_2) \leq \lambda y(x_1) + (1 - \lambda)y(x_2)$. Next, because $\phi(x, y)$ is convex in x and y and nondecreasing in y , we obtain with $x_\lambda = \lambda x_1 + (1 - \lambda)x_2$

$$\begin{aligned} \tilde{\phi}(x_\lambda) = \phi(x_\lambda, y(x_\lambda)) &\leq \phi(x_\lambda, \lambda y(x_1) + (1 - \lambda)y(x_2)) \\ &\leq \lambda \phi(x_1, y(x_1)) + (1 - \lambda) \phi(x_2, y(x_2)) \leq \lambda \tilde{\phi}(x_1) + (1 - \lambda) \tilde{\phi}(x_2) \end{aligned}$$

Therefore, as a sum of convex functions, P is convex. \square

Note, that we obtain strict convexity of P , if one of the inequalities in the proof of Lemma 2.5 holds strictly. In this case we can furthermore apply a standard uniqueness result.

Theorem 2.6. *Assume that X is nonempty, closed and convex and P is continuous and strictly convex on an open set $\Omega \supseteq X$. Then (12) admits a unique Nash equilibrium.*

Having discussed the existence of Nash equilibria of (12), we are now interested in the necessary and sufficient conditions that characterize these solutions. As for the existence, we will derive such conditions in terms of conditions for (15). Note, that we assume again that $y(x)$ denotes a continuous (single-valued) solution mapping for the followers problem. However, we do not claim, that $y(x)$ is a smooth function of x . We therefore apply convex, nonsmooth analysis results as in [3, 10] to the (nonsmooth) problem (15).

Theorem 2.7. *Let the assumptions of Lemma 2.5 be satisfied and let $\mathcal{N}_X(x^*)$ be the normal cone of X at x^* and $\partial P(x)$ be the convex subdifferential of P at x . Then x^* is a global minimizer of (15) iff*

$$0 \in \partial P(x^*) + \mathcal{N}_X(x^*)$$

with $\partial P(x)$ be given by

$$\partial P(x) = \sum_{\nu=1}^N \nabla \varphi_\nu(x_\nu) + \nabla \phi_1(x, y(x)) + V^T \nabla \phi_2(x, y(x)), \quad \text{for some } V \in \partial y(x), \quad (16)$$

with $\nabla \phi_{1,2}$ denoting the gradient of ϕ with respect to the first and second component, respectively.

Finally, let us apply these results to the Examples 1.9 and 1.10 of the previous section.

Example 2.8. Assume again that the sets Y_j are polyhedrons and that we can replace the followers game by condition (11). If π is give as a quadratic function of the form

$$\pi(y, x) = \frac{1}{2} \sum_{j=1}^M y_j^T Q_j y_j + b(x)^T y$$

with Q_j being symmetric and positive definite on the nullspace of A_j and T_j being defined as in Example 1.9, then (11) corresponds to the condition

$$Q_j^z z_j + b_j^z(x) = 0 \quad \forall j = 1, \dots, M,$$

where $b_j^z(x) = T_j^T Q_j \hat{y}_j + T_j^T b_j(x)$ and $Q_j^z = T_j^T Q_j T_j$ is regular. Therefore, the single-valued solution of (2) is given by $y_j(x) = T_j (Q_j^z)^{-1} b_j^z(x) + \hat{y}_j$. Hence, if $b(x)$ is a continuously differentiable, convex function, this property transfers to $y^*(x)$ such that Theorem 2.7 can be applied with $\partial y(x) = Dy^*(x)$ under suitable assumptions on the leaders objectives and the feasible set X .

Example 2.9. We consider again Example 1.10, where we have already derived the continuous solution mapping $y_j(x) = \max(0, Q_j^{-1} b(x))$ for all $j = 1, \dots, M$, with each Q_j being a diagonal, positive definite matrix. Hence, since taking the pointwise maximum of two convex functions yields again a convex function, if $b(x)$ is convex, so is $y(x)$. We can therefore apply Theorem 2.7 to obtain the necessary and sufficient conditions given by (16) with

$$\partial y(x) = \begin{cases} 0, & \text{if } Q_j^{-1} b(x) < 0 \\ Q_j^{-1} D b(x), & \text{if } Q_j^{-1} b(x) > 0 \\ \xi Q_j^{-1} D b(x), \quad \xi \in [0, 1], & \text{if } Q_j^{-1} b(x) = 0 \end{cases}.$$

REFERENCES

- [1] Didier Aussel, Michal Cervinka, and Matthieu Marechal. Deregulated electricity markets with thermal losses and production bounds: models and optimality conditions. *RAIRO-Oper. Res.*, 50(1):19–38, 2016.
- [2] Stephen Boyd and Lieven Vandenbergh. *Convex Optimization*. Cambridge university press, 2004.
- [3] F. Clarke. *Optimization and Nonsmooth Analysis*. Society for Industrial and Applied Mathematics, 1990.
- [4] S. Dempe. *Foundations of bilevel programming*. Nonconvex Optimization and Its Applications 61, Kluwer Academic Publishers, Dordrecht, 2002.
- [5] F. Facchinei and J.S. Pang. *Finite-dimensional variational inequalities and complementarity problems*, volume 1 and 2. Springer Series in Operations Research. New York, Springer, 2003.
- [6] M. Herty, S. Steffensen, and A. Thünen. Solving quadratic multi-leader-follower games by smoothing the follower’s best response, 2019.
- [7] M. Hu and M. Fukushima. Existence, uniqueness, and computation of robust nash equilibria in a class of multi-leader-follower games. *SIAM Journal on Optimization*, 23(2):894–916, 2013.
- [8] X.M. Hu and D. Ralph. Using epecs to model bilevel games in restructured electricity markets with locational prices. *Operations Research*, 55(5):809–827, 2007.
- [9] Andrew Koh and Simon Shepherd. Tolling, collusion and equilibrium problems with equilibrium constraints. 2010.
- [10] M. Kočvara, J. Outrata, and J. Zowe. *Nonsmooth approach to optimization problems with equilibrium constraints. Theory, applications and numerical results*. Nonconvex Optimization and Its Applications. 28. Dordrecht: Kluwer Academic Publishers, 1998.
- [11] Ankur A. Kulkarni and Uday V. Shanbhag. A shared-constraint approach to multi-leader multi-follower games. *Set-Valued and Variational Analysis*, 22(4):691–720, aug 2014.
- [12] Ankur A. Kulkarni and Uday V. Shanbhag. An existence result for hierarchical stackelberg v/s stackelberg games. *IEEE Transactions on Automatic Control*, 60(12):3379–3384, dec 2015.
- [13] Sven Leyffer and Todd Munson. Solving multi-leader-common-follower games. *Optimization Methods and Software*, 25(4):601–623, 2010.
- [14] Dov Monderer and Lloyd S. Shapley. Potential games. *Games and Economic Behavior*, 14(1):124 – 143, 1996.
- [15] J. F. Nash. Non-cooperative games. *Annals of Mathematics*, 54(2):286–295, 1951.
- [16] Hanif D. Sherali. A multiple leader stackelberg model and analysis. *Operations Research*, 32(2):390–404, 1984.
- [17] S. Steffensen and M. Bittner. Relaxation approach for equilibrium problems with equilibrium constraints. *Computers and Operations Research*, (41(1)):333–345, 2011.
- [18] S. Steffensen and A. Thünen. An explicit nash equilibrium to a multi-leader-follower game. In *PAMM-Proc. Appl. Math. Mech.* Wiley-VCH Verlag GmbH & Co. KGaA, Weinheim, 2019.
- [19] Che-Lin Su. Analysis on the forward market equilibrium model. *Operations Research Letters*, 35(1):74–82, jan 2007.
- [20] Jessie Hui Wang, Dah Ming Chiu, and John C. S. Lui. A game-theoretic analysis of the implications of overlay network traffic on ips peering. *Computer Networks*, 2008.