# Some constraint qualifications for quasiconvex vector-valued systems 

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#### Abstract

In this paper, we consider minimization problems with a quasiconvex vector-valued inequality constraint. We propose two constraint qualifications, the closed cone constraint qualification for vector-valued quasiconvex programming (the VQ-CCCQ) and the basic constraint qualification for vector-valued quasiconvex programming (the VQ-BCQ). Based on previous results by Benoist, Borwein, and Popovici (Proc. Amer. Math. Soc. 13: 1109-1113, 2002), and the authors (J. Optim. Theory Appl. 149: 554-563, 2011 and Nonlinear Anal. 74: 1279-1285, 2011), we show that the VQ-CCCQ (resp. the VQ-BCQ) is the weakest constraint qualification for Lagrangian-type strong (resp. min-max) duality. As consequences of the main results, we study semi-definite quasiconvex programming problems in our scheme, and we observe the weakest constraint qualifications for Lagrangiantype strong and min-max dualities. Finally, we summarize the characterizations of the weakest constraint qualifications for convex and quasiconvex programming.


Keywords quasiconvex programming • quasiaffine functions • vector-valued • constraint qualification

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## 1 Introduction

In mathematical programming, constraint qualifications are essential elements for duality theory. The best-known constraint qualifications are the interior point conditions, also known as the Slater-type constraint qualifications. Often, however, such constraint qualifications are not satisfied for problems that arise in applications. The lack of a constraint qualification can cause theoretical and numerical

[^0]difficulties in applications. In convex programming, research on the weakest constraint qualifications for Lagrangian strong and min-max dualities has been carried out in many studies. Jeyakumar, Dinh and Lee developed the closed cone constraint qualification (the CCCQ) involving epigraphs and extending the Slater-type conditions in [1], and Jeyakumar [2] demonstrated that the CCCQ is the weakest constraint qualification for Lagrangian (strong) duality (in [2], the CCCQ is called [CQ1]). Also, Li, Ng and Pong established the basic constraint qualification (the BCQ ) as the weakest constraint qualification for the Lagrangian min-max duality in [3]: let $I$ be an arbitrary set, $f$ and $g_{i}$ be proper lsc convex functions from $X$ to $\overline{\mathbb{R}}$, and $A=\left\{x \in X \mid \forall i \in I, g_{i}(x) \leq 0\right\}$; then $x_{0} \in A$ is a global minimizer of $f$ in $A$ if and only if there exists $\lambda \in \mathbb{R}_{+}^{(I)}$ such that $0 \in \partial f\left(x_{0}\right)+\sum_{i \in I} \lambda_{i} \partial g_{i}\left(x_{0}\right)$. Moreover, for a vector-valued convex inequality system, Jeyakumar established the BCQ, which is called [CQ2], for Lagrangian min-max duality as the weakest constraint qualification (see [2]).

In recent research $[4,5]$ on the weakest constraint qualifications for Lagrangiantype strong and min-max dualities in quasiconvex programming, we established constraint qualifications for real-valued quasiconvex inequality systems (the QCCCQ and the Q-BCQ) and proved that these constraint qualifications were the weakest ones for certain Lagrangian-type dualities. To define these constraint qualifications, we introduced the notion of a generator of quasiconvex functions, based on Penot and Volle's interesting result that each lower semi-continuous quasiconvex function consists of a supremum of some family of lower semi-continuous quasiaffine functions [6].

In the present paper, we consider the weakest constraint qualifications for Lagrangian-type dualities of the following minimization programming problem with a quasiconvex vector-valued inequality constraint:

$$
\left\{\begin{array}{l}
\operatorname{minimize} f(x), \\
\text { subject to } g(x) \in-K,
\end{array}\right.
$$

where $X$ and $Y$ are Banach spaces, $K$ is a closed convex cone in $Y, f$ is a function from $X$ to $\overline{\mathbb{R}}=[-\infty,+\infty]$, and $g$ is a $K$-quasiconvex function from $X$ to $Y$. We propose and investigate constraint qualifications for a vector-valued quasiconvex inequality system as generalizations of the results in [4] and [5], and show that these qualifications are the weakest ones for certain Lagrangian-type dualities.

The remainder of the present paper is organized as follows. In Section 2, we give some preliminaries and notation. In Section 3, we propose two constraint qualifications, the closed cone constraint qualification for vector-valued quasiconvex programming (the VQ-CCCQ) and the basic constraint qualification for vector-valued quasiconvex programming (the VQ-BCQ). Based on previous results by Benoist, Borwein, and Popovici [7], and the authors [4,5], we show that the VQ-CCCQ (resp. the VQ-BCQ) is the weakest constraint qualification for Lagrangian-type strong (resp. min-max) duality. As consequences of the main results, in Section 4, we study semi-definite quasiconvex programming problems in our scheme, and we observe the weakest constraint qualifications for Lagrangian-type strong and min-max dualities. Finally, in Section 5, we summarize the characterizations of the weakest constraint qualifications for convex and quasiconvex programming.

## 2 Preliminaries

Let $X$ be a Banach space, let $X^{*}$ be the continuous dual space of $X$, and let $\left\langle x^{*}, x\right\rangle$ denote the value of a functional $x^{*} \in X^{*}$ at $x \in X$. Given a set $A^{*} \subset X^{*}$, we denote the $w^{*}$-closure, the convex hull and the conical hull generated by $A^{*}$, by $\operatorname{cl} A^{*}$, $\operatorname{co} A^{*}$ and cone $A^{*}$, respectively. The normal cone of $A \subset X$ at $z_{0} \in A$ is denoted by $N_{A}\left(z_{0}\right)=\left\{x^{*} \in X^{*} \mid \forall y \in A,\left\langle x^{*}, y-z_{0}\right\rangle \leq 0\right\}$. The indicator function $\delta_{A}$ of $A$ is defined by

$$
\delta_{A}(x):= \begin{cases}0 & x \in A, \\ \infty & \text { otherwise } .\end{cases}
$$

Throughout the present paper, let $f$ be a function from $X$ to $\overline{\mathbb{R}}$, where $\overline{\mathbb{R}}=$ $[-\infty, \infty]$. Here, $f$ is said to be proper if for all $x \in X, f(x)>-\infty$ and there exists $x_{0} \in X$ such that $f\left(x_{0}\right) \in \mathbb{R}$. We denote the domain of $f$ by $\operatorname{dom} f$, that is, $\operatorname{dom} f=$ $\{x \in X \mid f(x)<\infty\}$. The epigraph of $f$, epi $f$, is defined as epi $f=\{(x, r) \in X \times \mathbb{R} \mid$ $f(x) \leq r\}$, and $f$ is said to be convex if epi $f$ is convex. In addition, the Fenchel conjugate of $f, f^{*}: X^{*} \rightarrow \overline{\mathbb{R}}$, is defined as $f^{*}(u)=\sup _{x \in \operatorname{dom} f}\{\langle u, x\rangle-f(x)\}$. Recall that $f$ is said to be quasiconvex if for all $x_{1}, x_{2} \in X$ and $\alpha \in(0,1)$,

$$
f\left((1-\alpha) x_{1}+\alpha x_{2}\right) \leq \max \left\{f\left(x_{1}\right), f\left(x_{2}\right)\right\} .
$$

Define level sets of $f$ with respect to a binary relation $\diamond$ on $\overline{\mathbb{R}}$ as

$$
L(f, \diamond, \alpha)=\{x \in X \mid f(x) \diamond \alpha\}
$$

for any $\alpha \in \mathbb{R}$. Then, $f$ is quasiconvex if and only if for any $\alpha \in \mathbb{R}, L(f, \leq, \alpha)$ is a convex set, or equivalently, for any $\alpha \in \mathbb{R}, L(f,<, \alpha)$ is a convex set. Any convex function is quasiconvex, but the opposite is not true.

Recall $\Gamma_{0}(X)$, the set of all proper lsc convex functions. It is well known that a function in $\Gamma_{0}(X)$ consists of a supremum of some family of affine functions. In the case of quasiconvex functions, a similar result was also proved by Penot and Volle [6]. A function $f$ is said to be quasiaffine if quasiconvex and quasiconcave. It is important that $f$ is lsc quasiaffine if and only if there exists $k \in Q$ and $w \in X^{*}$ such that $f=k \circ w$, where $Q=\{k: \mathbb{R} \rightarrow \overline{\mathbb{R}} \mid k$ is lsc and non-decreasing $\}$. By using the notion of quasiaffine function, Penot and Volle proved that $f$ is lsc quasiconvex if and only if there exists $\left\{\left(k_{i}, w_{i}\right) \mid i \in I\right\} \subset Q \times X^{*}$ such that $f=\sup _{i \in I} k_{i} \circ w_{i}$. This result indicates that a lsc quasiconvex function $f$ consists of a supremum of some family of lsc quasiaffine functions. Based on this result, we define a notion of generator for quasiconvex functions in [4], that is, $G=\left\{\left(k_{i}, w_{i}\right) \mid i \in I\right\} \subset Q \times X^{*}$ is said to be a generator of $f$ if $f=\sup _{i \in I} k_{i} \circ w_{i}$. From Penot and Volle's result, all lsc quasiconvex functions have at least one generator. Also, it is clear that all proper lsc convex functions have a generator which consists of continuous affine functions.

Moreover, we introduce a generalized notion of inverse function of $k \in Q$. The following function $k^{-1}$ is said to be the hypo-epi-inverse of $k$ :

$$
k^{-1}(a)=\inf \{b \in \mathbb{R} \mid a<k(b)\}=\sup \{b \in \mathbb{R} \mid k(b) \leq a\}
$$

If $k$ has the inverse function, then the inverse and the hypo-epi-inverse of $k$ are the same, in detail see [6]. In this paper, we denote the hypo-epi-inverse of $k$ by $k^{-1}$.

Also, we denote the lower left-hand Dini derivative of $k \in Q$ at $t$ by $D_{-} k(t)$, that is $D_{-} k(t)=\liminf _{\alpha \rightarrow 0-} \frac{k(t+\alpha)-k(t)}{\alpha}$. A function $k$ is said to be lower left-hand Dini differentiable if $D_{-} k(t)$ is finite for all $t \in \mathbb{R}$.

Let $Y$ be a Banach space, and $K$ a nonempty closed convex cone of $Y$. By introducing the binary relation $\leq_{K}$ on $Y$ by $y \leq_{K} z$ if and only if $z-y \in K$, $\left(Y, \leq_{K}\right)$ becomes a partially ordered set. Also, $\left(Y, \leq_{K}\right)$ is said to be directed if for all $y_{1}, y_{2} \in Y$, there exists $z \in Y$ such that $y_{1} \leq_{K} z$ and $y_{2} \leq_{K} z$. Let $Y^{*}$ be the continuous dual space of $Y, K^{+}$the positive polar cone of $K$, and extd $K^{+}$the set of all extreme directions of $K^{+}$. Recall that $K^{+}=\left\{y^{*} \in Y^{*} \mid \forall y \in K,\left\langle y^{*}, y\right\rangle \geq 0\right\}$ and $y^{*} \in \operatorname{extd} K^{+}$if and only if $y^{*} \in K^{+} \backslash\{0\}$ and for all $y_{1}^{*}, y_{2}^{*} \in K^{+}$with $y^{*}=y_{1}^{*}+y_{2}^{*}, y_{1}^{*}, y_{2}^{*} \in \mathbb{R}_{+}\left\{y^{*}\right\}$. A function $g$ is said to be $K$-convex if for all $x_{1}$, $x_{2} \in X$, and $\alpha \in(0,1),(1-\alpha) g\left(x_{1}\right)+\alpha g\left(x_{2}\right) \in g\left((1-\alpha) x_{1}+\alpha x_{2}\right)+K$. It is well known that $g$ is $K$-convex if and only if $y^{*} \circ g$ is convex for all $y^{*} \in K^{+}$. Also, a function $g$ is said to be $K$-quasiconvex if for all $y \in Y, x_{1}, x_{2} \in X$, and $\alpha \in(0,1)$ with $y \in\left(g\left(x_{1}\right)+K\right) \cap\left(g\left(x_{2}\right)+K\right), y \in g\left((1-\alpha) x_{1}+\alpha x_{2}\right)+K$. In [7], $K$-quasiconvexity is characterized as follows:

Theorem 2.1 [7] Let $D$ be a nonempty convex subset of a vector space $Z$, and $K$ be a closed convex cone in Banach space $Y$ satisfying $\left(Y, \leq_{K}\right)$ is directed and $K^{+}=\mathrm{cl}$ coextd $K^{+}$. For a function $g$ from $D$ to $Y$, the following conditions are equivalent:
(i) $g$ is $K$-quasiconvex,
(ii) $y^{*} \circ g$ is quasiconvex for all $y^{*} \in \operatorname{extd} K^{+}$.

Next we mention about previous results concerned with the weakest constraint qualifications. In convex optimization, the closed cone constraint qualification (the CCCQ) have been investigated extensively as the weakest one for the Lagrangian strong duality see $[1,3,8,9]$, and the basic constraint qualification (the BCQ) for the Lagrangian min-max duality, see [2,3]. In quasiconvex optimization, the closed cone constraint qualification for quasiconvex programming (the Q-CCCQ) has been established in [4], as the weakest one for the Lagrangian-type strong duality:

Definition 2.1 [4] Let $\left\{g_{i} \mid i \in I\right\}$ be a family of lsc quasiconvex functions from $X$ to $\overline{\mathbb{R}},\left\{\left(k_{(i, j)}, w_{(i, j)}\right) \mid j \in J_{i}\right\} \subset Q \times X^{*}$ be a generator of $g_{i}$ for each $i \in I$ and $T=\left\{t=(i, j) \mid i \in I, j \in J_{i}\right\}$. Assume that $A=\left\{x \in X \mid \forall i \in I, g_{i}(x) \leq 0\right\}$ is non-empty set. Then, the quasiconvex system $\left\{g_{i}(x) \leq 0 \mid i \in I\right\}$ satisfies the closed cone constraint qualification for quasiconvex programming (the Q-CCCQ) w.r.t. $\left\{\left(k_{t}, w_{t}\right) \mid t \in T\right\}$ if

$$
\text { cone co } \bigcup_{t \in T}\left\{\left(w_{t}, \delta\right) \in X^{*} \times \mathbb{R} \mid k_{t}^{-1}(0) \leq \delta\right\}+\{0\} \times[0, \infty)
$$

is $w^{*}$-closed.
Theorem 2.2 [4] Let $f$ be a lsc quasiconvex function from $X$ to $\overline{\mathbb{R}}$ with generator $\left\{\left(k_{i}, w_{i}\right) \mid i \in I\right\} \subset Q \times X^{*}$. Assume that $A=\{x \in X \mid f(x) \leq 0\}$ is non-empty set. Then, the following statements are equivalent:
(i) $\{f(x) \leq 0\}$ satisfies the $Q$-CCCQ w.r.t. $\left\{\left(k_{i}, w_{i}\right) \mid i \in I\right\}$,
(ii) for all $h \in \Gamma_{0}(X)$ with $\operatorname{dom} h \cap A \neq \emptyset$ and $\operatorname{epi} h^{*}+\operatorname{epi}_{A}^{*}$ is $w^{*}$-closed,

$$
\inf _{x \in A} h(x)=\max _{\lambda \in \mathbb{R}_{+}^{(I)}} \inf _{x \in X}\left\{h(x)+\sum_{i \in I} \lambda_{i}\left(w_{i}(x)-k_{i}^{-1}(0)\right)\right\} .
$$

Also the basic constraint qualification for quasiconvex programming (the $\mathrm{Q}-\mathrm{BCQ}$ ) has been established as the weakest one for Lagrangian-type min-max duality by the authors, see [5]:

Definition 2.2 [5] Let $\left\{g_{i} \mid i \in I\right\}$ be a family of lsc quasiconvex functions from $X$ to $\overline{\mathbb{R}},\left\{\left(k_{(i, j)}, w_{(i, j)}\right) \mid j \in J_{i}\right\} \subset Q \times X^{*}$ be a generator of $g_{i}$ for each $i \in I$, $T=\left\{t=(i, j) \mid i \in I, j \in J_{i}\right\}, T(x)=\left\{t \in T \mid k_{t}\left(\left\langle w_{t}, x\right\rangle\right)=0, k_{t}^{-1}(0)=\left\langle w_{t}, x\right\rangle\right\}$, and $A=\left\{x \in X \mid \forall i \in I, g_{i}(x) \leq 0\right\}$.

The family $\left\{g_{i} \mid i \in I\right\}$ is said to satisfy the basic constraint qualification for quasiconvex programming (the Q-BCQ) with respect to $\left\{\left(k_{t}, w_{t}\right) \mid t \in T\right\}$ at $x \in A$ if

$$
N_{A}(x)=\text { cone co } \bigcup_{t \in T(x)}\left\{w_{t}\right\} .
$$

Theorem 2.3 [5] Let $\left\{g_{i} \mid i \in I\right\}$ be a family of lsc quasiconvex functions from $X$ to $\overline{\mathbb{R}},\left\{\left(k_{(i, j)}, w_{(i, j)}\right) \mid j \in J_{i}\right\} \subset Q \times X^{*}$ be a generator of $g_{i}$ for each $i \in I$, $T=\left\{t=(i, j) \mid i \in I, j \in J_{i}\right\}, T(x)=\left\{t \in T \mid k_{t}\left(\left\langle w_{t}, x\right\rangle\right)=0, k_{t}^{-1}(0)=\left\langle w_{t}, x\right\rangle\right\}$, $A=\left\{x \in X \mid \forall i \in I, g_{i}(x) \leq 0\right\}$ and $x_{0} \in A$. Then, the following statements (i), (ii) and (iii) are equivalent:
(i) $\left\{g_{i}(x) \leq 0 \mid i \in I\right\}$ satisfies the $Q-B C Q$ w.r.t. $\left\{\left(k_{t}, w_{t}\right) \mid t \in T\right\}$ at $x_{0}$,
(ii) for each $f \in \Gamma_{0}(X)$ with $\operatorname{dom} f \cap A \neq \emptyset$ and epif $f^{*}+\mathrm{epid}_{A}^{*}$ is $w^{*}$-closed, $x_{0}$ is a minimizer of $f$ in $A$ if and only if there exists $\lambda \in \mathbb{R}_{+}^{(T)}$ such that $\lambda_{t}=0$ for all $t \in T \backslash T\left(x_{0}\right)$, and

$$
0 \in \partial f\left(x_{0}\right)+\sum_{t \in T} \lambda_{t} w_{t},
$$

(iii) for all $f \in Q_{F}(X) \cup Q_{C}(X)$ with a generator $G$, if $x_{0}$ is a local minimizer of $f$ in $A$, then, there exists $\lambda \in \mathbb{R}_{+}^{(T)}$ such that $\lambda_{t}=0$ for all $t \in T \backslash T\left(x_{0}\right)$, and

$$
0 \in \partial_{G} f\left(x_{0}\right)+\sum_{t \in T} \lambda_{t} w_{t},
$$

where $\partial_{G} f\left(x_{0}\right)$ is the subdifferential in [5] of $f$ at $x_{0}$, that is, $\partial_{G} f\left(x_{0}\right)=\mathrm{cl} \operatorname{co}\left\{D_{-} k_{s}\left(\left\langle w_{s}, x_{0}\right\rangle\right) w_{s} \mid\right.$ $\left.s \in S\left(x_{0}\right)\right\}$, and $Q_{F}(X)$ and $Q_{C}(X)$ are the following families of quasiconvex functions:

$$
\begin{gathered}
Q_{F}(X)=\left\{\begin{array}{l|l}
\left.\sup _{s \in S} k_{s} \circ w_{s} \left\lvert\, \begin{array}{l}
\left\{\left(k_{s}, w_{s}\right) \mid s \in S\right\} \subset Q \times X^{*}, S: \text { finite, } \\
\forall s \in S, k_{s}: \text { continuous and lower left-hand Dini diff. }
\end{array}\right.\right\}, \\
Q_{C}(X)=\left\{\begin{array}{ll}
\sup _{s \in S} k_{s} \circ w_{s} & \begin{array}{l}
\left\{\left(k_{s}, w_{s}\right) \mid s \in S\right\} \subset Q \times X^{*}, S: \text { compact, } \\
s \mapsto w_{s}: \text { continuous },(s, t) \mapsto k_{s}(t): \text { usc }, \\
D_{-} k_{s}(t) \in \mathbb{R} \text { and }(s, t) \mapsto D_{-}(t): \text { continuous. }
\end{array}
\end{array}\right\} .
\end{array}\right.
\end{gathered}
$$

## 3 Constraint qualifications for vector-valued quasiconvex functions

As mention in Section 1, we consider the weakest constraint qualifications for Lagrangian-type dualities of the following minimization programming problem with a quasiconvex vector-valued inequality constraint:

$$
\left\{\begin{array}{l}
\operatorname{minimize} f(x), \\
\text { subject to } g(x) \in-K,
\end{array}\right.
$$

where $X$ and $Y$ are Banach spaces, $K$ is a closed convex cone in $Y, f$ is a function from $X$ to $\overline{\mathbb{R}}=[-\infty,+\infty], g$ is a $K$-quasiconvex function from $X$ to $Y$, and $A=\{x \in X \mid g(x) \in-K\}$. Also, assume that $\left(Y, \leq_{K}\right)$ is directed and $K^{+}=$ cl co extd $K^{+}$.

Definition 3.1 Let $g$ be a continuous $K$-quasiconvex function, $\left\{\left(k_{\left(y^{*}, j\right)}, w_{\left(y^{*}, j\right)}\right) \mid\right.$ $\left.j \in J_{y^{*}}\right\} \subset Q \times X^{*}$ be a generator of $y^{*} \circ g$ for each $y^{*} \in \operatorname{extd} K^{+}$, and $T=$ $\left\{t=\left(y^{*}, j\right) \mid y^{*} \in \operatorname{extd} K^{+}, j \in J_{y^{*}}\right\}$. Assume that $A$ is non-empty set. Then, the vector-valued quasiconvex system $\{g(x) \in-K\}$ satisfies the closed cone constraint qualification for vector-valued quasiconvex programming (the VQ-CCCQ) with respect to $\left\{\left(k_{t}, w_{t}\right) \mid t \in T\right\}$ if

$$
\text { cone co } \bigcup_{t \in T}\left\{\left(w_{t}, \delta\right) \in X^{*} \times \mathbb{R} \mid k_{t}^{-1}(0) \leq \delta\right\}+\{0\} \times[0, \infty)
$$

is $w^{*}$-closed.
By the similar way in [4], we can check that $\{g(x) \in-K\}$ satisfies the VQ-CCCQ if and only if the alternative form of the VQ-CCCQ,

$$
\operatorname{epi} \delta_{A}^{*} \subset \text { cone co } \bigcup_{t \in T}\left\{\left(w_{t}, \delta\right) \in X^{*} \times \mathbb{R} \mid k_{t}^{-1}(0) \leq \delta\right\}+\{0\} \times[0, \infty)
$$

holds. In the next theorem, we can see that the VQ-CCCQ is the weakest constraint qualification of a vector-valued quasiconvex inequality system for Lagrangian-type duality.

Theorem 3.1 Let $g$ be a continuous K-quasiconvex function, $\left\{\left(k_{\left(y^{*}, j\right)}, w_{\left(y^{*}, j\right)}\right) \mid\right.$ $\left.j \in J_{y^{*}}\right\} \subset Q \times X^{*}$ be a generator of $y^{*} \circ g$ for each $y^{*} \in \operatorname{extd} K^{+}$, and $T=$ $\left\{t=\left(y^{*}, j\right) \mid y^{*} \in \operatorname{extd} K^{+}, j \in J_{y^{*}}\right\}$. Assume that $A$ is non-empty set. Then, the following statements are equivalent:
(i) $\{g(x) \in-K\}$ satisfies the VQ-CCCQ w.r.t. $\left\{\left(k_{t}, w_{t}\right) \mid t \in T\right\}$,
(ii) for all $f \in \Gamma_{0}(X)$ with epif $f^{*}+\operatorname{epi} \delta_{A}^{*}$ is $w^{*}$-closed, there exists $\lambda \in \mathbb{R}_{+}^{(T)}$ such that

$$
\inf _{x \in A} f(x)=\inf _{x \in X}\left\{f(x)+\sum_{t \in T} \lambda_{t}\left(w_{t}(x)-k_{t}^{-1}(0)\right)\right\} .
$$

Proof We can verify that

$$
A=\{x \in X \mid g(x) \in-K\}=\left\{x \in X \mid \forall y^{*} \in \operatorname{extd} K^{+}, y^{*} \circ g(x) \leq 0\right\}
$$

since $K^{+}=\mathrm{cl}$ co extd $K^{+}$and $K=\left\{y \in Y \mid \forall y^{*} \in K^{+},\left\langle y^{*}, y\right\rangle \geq 0\right\}$, and we have

$$
A=\left\{x \in X \mid \forall t \in T, k_{t} \circ w_{t}(x) \leq 0\right\}
$$

from the definition of $T$. Put $h=\sup _{t \in T} k_{t} \circ w_{t}$, then $h$ is a lsc quasiconvex function with generator $\left\{\left(k_{t}, w_{t}\right) \mid t \in T\right\}$. Hence, by using Theorem 2.2, we can prove this theorem.

Next, we introduce the following constraint qualification for Lagrangian-type min-max duality:

Definition 3.2 Let $g$ be a continuous $K$-quasiconvex function, $\left\{\left(k_{\left(y^{*}, j\right)}, w_{\left(y^{*}, j\right)}\right) \mid\right.$ $\left.j \in J_{y^{*}}\right\} \subset Q \times X^{*}$ be a generator of $y^{*} \circ g$ for each $y^{*} \in \operatorname{extd} K^{+}, T=\left\{t=\left(y^{*}, j\right) \mid\right.$ $\left.y^{*} \in \operatorname{extd} K^{+}, j \in J_{y^{*}}\right\}$, and $T(x)=\left\{t \in T \mid k_{t}\left(\left\langle w_{t}, x\right\rangle\right)=0, k_{t}^{-1}(0)=\left\langle w_{t}, x\right\rangle\right\}$. Assume that $A$ is non-empty set. Then, the vector-valued quasiconvex system $\{g(x) \in-K\}$ is said to satisfy the basic constraint qualification for vector-valued quasiconvex programming (the VQ-BCQ) with respect to $\left\{\left(k_{t}, w_{t}\right) \mid t \in T\right\}$ at $x \in A$ if

$$
N_{A}(x)=\text { cone co } \bigcup_{t \in T(x)}\left\{w_{t}\right\} .
$$

Also, we can check that the $\mathrm{VQ}-\mathrm{BCQ}$ is equivalent to the following inclusion

$$
N_{A}(x) \subset \text { cone co } \bigcup_{t \in T(x)}\left\{w_{t}\right\}
$$

The following result shows us the VQ-BCQ is the weakest constraint qualification of a vector-valued quasiconvex inequality system for Lagrangian-type minmax duality:

Theorem 3.2 Let $g$ be a continuous $K$-quasiconvex function, $\left\{\left(k_{\left(y^{*}, j\right)}, w_{\left(y^{*}, j\right)}\right) \mid\right.$ $\left.j \in J_{y^{*}}\right\} \subset Q \times X^{*}$ be a generator of $y^{*} \circ g$ for each $y^{*} \in \operatorname{extd} K^{+}, T=\left\{t=\left(y^{*}, j\right) \mid\right.$ $\left.y^{*} \in \operatorname{extd} K^{+}, j \in J_{y^{*}}\right\}$, and $T(x)=\left\{t \in T \mid k_{t}\left(\left\langle w_{t}, x\right\rangle\right)=0, k_{t}^{-1}(0)=\left\langle w_{t}, x\right\rangle\right\}$. and $x_{0} \in A$. Then, the following statements (i), (ii) and (iii) are equivalent:
(i) $\{g(x) \in-K\}$ satisfies the VQ-BCQ w.r.t. $\left\{\left(k_{t}, w_{t}\right) \mid t \in T\right\}$ at $x_{0}$,
(ii) for each $f \in \Gamma_{0}(X)$ with $\operatorname{dom} f \cap A \neq \emptyset$ and epi $f^{*}+\operatorname{epi} \delta_{A}^{*}$ is $w^{*}$-closed, $x_{0}$ is a minimizer of $f$ in $A$ if and only if there exists $\lambda \in \mathbb{R}_{+}^{(T)}$ such that $\lambda_{t}=0$ for all $t \in T \backslash T\left(x_{0}\right)$, and

$$
0 \in \partial f\left(x_{0}\right)+\sum_{t \in T} \lambda_{t} w_{t}
$$

(iii) for all $f \in Q_{F}(X) \cup Q_{C}(X)$ with a generator $G$, if $x_{0}$ is a local minimizer of $f$ in $A$, then, there exists $\lambda \in \mathbb{R}_{+}^{(T)}$ such that $\lambda_{t}=0$ for all $t \in T \backslash T\left(x_{0}\right)$, and

$$
0 \in \partial_{G} f\left(x_{0}\right)+\sum_{t \in T} \lambda_{t} w_{t}
$$

Proof In the similar way of the proof of Theorem 3.1, we have $A=\{x \in X \mid$ $\left.\forall t \in T, k_{t} \circ w_{t}(x) \leq 0\right\}$, and $h=\sup _{t \in T} k_{t} \circ w_{t}$ is a lsc quasiconvex function with generator $\left\{\left(k_{t}, w_{t}\right) \mid t \in T\right\}$. By using Theorem 2.3, we can prove this theorem.

## 4 Applications

In this section, we investigate some applications of results in this paper. Let $S^{n}$ be the vector space of $(n \times n)$ symmetric matrices with the trace inner product which is partially ordered by $\succeq$, that is, for $M, N \in S^{n}, M \succeq N$ if and only if $(M-N)$ is positive semidefinite. Consider the following minimization programming problem:

$$
\left\{\begin{array}{l}
\operatorname{minimize} f(x), \\
\text { subject to } G(x) \preceq 0,
\end{array}\right.
$$

where $X$ is a Banach space, $f$ is a function from $X$ to $\overline{\mathbb{R}}$, and $G$ is a function from $X$ to $S^{n}$. Let $S_{+}^{n}=\left\{M \in S^{n} \mid M \succeq 0\right\}$. Then $S_{+}^{n}$ is a closed convex cone, $\left(S_{+}^{n}\right)^{+}=S_{+}^{n}$ and int $S_{+}^{n}=\left\{M \in S^{n} \mid M\right.$ is positive definite $\}$, in detail, see [2]. Since int $S_{+}^{n}$ is nonempty, cl co extd $S_{+}^{n}=S_{+}^{n}$ and ( $S^{n}, \leq_{S_{+}^{n}}$ ) is directed, in detail, see [7]. Hence, Theorem 3.1 and 3.2 apply to any $S_{+}^{n}$-quasiconvex functions.

In the rest of this section, we consider the case when $n=2$. Let $M=\left(\begin{array}{ll}a & b \\ b & c\end{array}\right) \in$ $S^{2}$. Then, $M \in S_{+}^{2}$ if and only if

$$
(a=0, b=0 \text { and } c \geq 0) \text { or }\left(a>0 \text { and } c \geq \frac{b^{2}}{a}\right) .
$$

Now we show that

$$
\operatorname{extd} S_{+}^{2}=\operatorname{cone}\left(\left\{\left(\begin{array}{ll}
0 & 0 \\
0 & 1
\end{array}\right)\right\} \bigcup\left\{\left.\left(\begin{array}{ll}
1 & b \\
b & b^{2}
\end{array}\right) \right\rvert\, b \in \mathbb{R}\right\}\right) \backslash\left\{\left(\begin{array}{ll}
0 & 0 \\
0 & 0
\end{array}\right)\right\}
$$

Let $A=\left(\begin{array}{ll}0 & 0 \\ 0 & 1\end{array}\right), B_{b}=\left(\begin{array}{cc}1 & b \\ b & b^{2}\end{array}\right)$ for all $b \in \mathbb{R}, M=\left(\begin{array}{ll}a_{1} & b_{1} \\ b_{1} & c_{1}\end{array}\right)$, and $N=\left(\begin{array}{ll}a_{2} & b_{2} \\ b_{2} & c_{2}\end{array}\right) \in$ $S_{+}^{2}$. First we show that $A \in \operatorname{extd} S_{+}^{2}$. If $M+N=A$, then $a_{1}+a_{2}=0$ and $a_{1}$, $a_{2} \geq 0$ since $M$ and $N \in S_{+}^{2}$. This implies that $a_{1}=a_{2}=0$. Similarly, $b_{1}=b_{2}=0$. Also, clearly $c_{1}, c_{2} \geq 0$, this implies that $M, N \in \mathbb{R}_{+}\{A\}$, that is, $A \in \operatorname{extd} S_{+}^{2}$. Next we show that $B_{b} \in \operatorname{extd} S_{+}^{2}$ for each $b \in \mathbb{R}$. Assume that $M+N=B_{b}$. If $a_{2}=0$, then $1=a_{1}, b=b_{1}$, and $b^{2}=c_{1}+c_{2}$. Since $a_{1}=1$ and $c_{2} \geq 0$,

$$
b^{2}=c_{1}+c_{2} \geq c_{1} \geq \frac{b_{1}^{2}}{a_{1}}=b^{2}
$$

that is $c_{1}=b^{2}$ and $c_{2}=0$. Hence, $M=\left(\begin{array}{cc}1 & b \\ b & b^{2}\end{array}\right)$ and $N=0 \in \mathbb{R}\{B\}$. If $a_{1} \neq 0$ and $a_{2} \neq 0$, then $1=a_{1}+a_{2}, b=b_{1}+b_{2}, b^{2}=c_{1}+c_{2}, c_{1} \geq \frac{b_{1}^{2}}{a_{1}}$ and $c_{2} \geq \frac{b_{2}^{2}}{a_{2}}$. Clearly, $1>a_{1}>0$. Then,

$$
\frac{b_{1}^{2}}{a_{1}}+\frac{b_{2}^{2}}{a_{2}}-b^{2}=\frac{b_{1}^{2}}{a_{1}}+\frac{\left(b-b_{1}\right)^{2}}{1-a_{1}}-b^{2}=\frac{\left(b_{1}-b a_{1}\right)^{2}}{a_{1}\left(1-a_{1}\right)} \geq 0
$$

Hence,

$$
b^{2}=c_{1}+c_{2} \geq \frac{b_{1}^{2}}{a_{1}}+\frac{b_{2}^{2}}{a_{2}} \geq b^{2}
$$

that is $c_{1}=\frac{b_{1}^{2}}{a_{1}}$ and $c_{2}=\frac{b_{2}^{2}}{a_{2}}$. Therefore, $M=a_{1}\left(\begin{array}{cc}1 & \frac{b_{1}}{a_{1}} \\ \frac{b_{1}}{a_{1}} & \frac{b_{1}}{a_{1}^{2}}\end{array}\right), N=a_{2}\left(\begin{array}{cc}1 & \frac{b_{2}}{a_{2}} \\ \frac{b_{2}}{a_{2}} & \frac{b_{2}}{a_{2}^{2}}\end{array}\right) \in$ $\mathbb{R}_{+}\left\{B_{b}\right\}$. This implies that $B_{b} \in \operatorname{extd} S_{+}^{2}$. Since $\lambda M \in \operatorname{extd} S_{+}^{2}$ holds for any $M \in \operatorname{extd} S_{+}^{2}$ and $\lambda>0$, we have

$$
\operatorname{extd} S_{+}^{2} \supset \operatorname{cone}\left(\left\{\left(\begin{array}{ll}
0 & 0 \\
0 & 1
\end{array}\right)\right\} \bigcup\left\{\left.\left(\begin{array}{ll}
1 & b \\
b & b^{2}
\end{array}\right) \right\rvert\, b \in \mathbb{R}\right\}\right) \backslash\left\{\left(\begin{array}{ll}
0 & 0 \\
0 & 0
\end{array}\right)\right\}
$$

Conversely, let $M=\left(\begin{array}{ll}a & b \\ b & c\end{array}\right) \in \operatorname{extd} S_{+}^{2}$. If $a=0$, then $a=b=0$ and $c>0$ since $M \in S_{+}^{2}$ and $M \neq 0$, that is, $M=c A$. If $a \neq 0$, then $a>0$ and $c \geq \frac{b^{2}}{a}$ since $M \in S_{+}^{2}$. Then, $c=\frac{b^{2}}{a}$ and $M$ is an element of the right-hand side. Actually, if $c>\frac{b^{2}}{a}$ then

$$
M=\left(\begin{array}{cc}
a & b \\
b & \frac{b^{2}}{a}
\end{array}\right)+\left(\begin{array}{cc}
0 & 0 \\
0 & c-\frac{b^{2}}{a}
\end{array}\right) .
$$

This is a contradiction. Therefore, the equality holds. Let $G$ be a function from $X$ to $S^{2}$ and $g_{1}, g_{2}$, and $g_{3}$ be functions from $X$ to $\overline{\mathbb{R}}$ as follows:

$$
g_{1}=\operatorname{Tr}\left[\left(\begin{array}{ll}
1 & 0 \\
0 & 0
\end{array}\right) G\right], g_{2}=\operatorname{Tr}\left[\left(\begin{array}{ll}
0 & 0 \\
1 & 0
\end{array}\right) G\right], g_{3}=\operatorname{Tr}\left[\left(\begin{array}{ll}
0 & 0 \\
0 & 1
\end{array}\right) G\right]
$$

where $\operatorname{Tr} M$ is the trace of a matrix $M$. Then, we can verify that $G=\left(\begin{array}{ll}g_{1} & g_{2} \\ g_{2} & g_{3}\end{array}\right)$, and by Theorem 2.1, $G$ is $S_{+}^{2}$-quasiconvex if and only if $g_{3}$ and $g_{1}+2 b g_{2}+b^{2} g_{3}$ ( $b \in \mathbb{R}$ ) are quasiconvex.

Assume that $G$ is $S_{+}^{2}$-quasiconvex. Then, for each $b \in \mathbb{R}$, there exists a generator $\left\{\left(k_{(b, j)}, w_{(b, j)}\right) \mid j \in J_{b}\right\} \subset Q \times X^{*}$ of $g_{1}+2 b g_{2}+b^{2} g_{3}$ and there exists a generator $\left\{\left(k_{(\gamma, j)}, w_{(\gamma, j)}\right) \mid j \in J_{\gamma}\right\} \subset Q \times X^{*}$ of $g_{3}$, where $\gamma$ is an index. We can reput generators $\left\{\left(k_{(b, w)}, w\right) \mid w \in S_{X^{*}}\right\} \subset Q \times X^{*}$ of $g_{1}+$ $2 b g_{2}+a^{2} g_{3}$ where $S_{X^{*}}=\left\{x^{*} \in X^{*} \mid\left\|x^{*}\right\|=1\right\}$. Actually, let $J_{(b, 0)}=\{j \in$ $\left.J_{b} \mid w_{(b, j)}=0\right\}, J_{(b, w)}=\left\{j \in J_{b} \left\lvert\, w=\frac{w_{(b, j)}}{\left\|w_{(b, j)}\right\|}\right.\right\}$ for all $w \in S_{X^{*}}$ and let $k_{(b, w)}(t)=\max \left\{\sup _{j \in J_{(b, w)}} k_{(b, j)}\left(\left\|w_{(b, j)}\right\| t\right), \sup _{j \in J_{(b, 0)}} k_{(b, j)}(0)\right\}$ for all $w \in S_{X^{*}}$. Then, $k_{(b, w)} \in Q$ and $g_{1}+2 b g_{2}+b^{2} g_{3}=\sup _{w \in S_{X^{*}}} k_{(b, w)} \circ w$. Also, we can reput a generator of $g_{3}$ similarly. Let

$$
D=\text { cone co } \bigcup_{w \in S_{X^{*}}}\left\{(w, \delta) \mid \inf _{b \in \mathbb{R} \cup\{\gamma\}} k_{(b, w)}^{-1}(0) \leq \delta\right\}+\{0\} \times[0, \infty)
$$

We can show the following corollary.
Corollary 4.1 Assume that $A=\{x \in X \mid G(x) \preceq 0\}$ is nonempty. Then, $D$ is closed if and only if for all $f \in \Gamma_{0}(X)$ with epif $f^{*}+\operatorname{epi}_{\delta_{A}^{*}}^{*}$ is $w^{*}$-closed, there exists $\lambda \in \mathbb{R}_{+}^{\left(S_{X^{*}}\right)}$ such that

$$
\inf _{x \in A} f(x)=\inf _{x \in X}\left\{f(x)+\sum_{w \in S_{X^{*}}} \lambda_{w}\left(\langle w, x\rangle-\inf _{b \in \mathbb{R} \cup\{\gamma\}} k_{(b, w)}^{-1}(0)\right)\right\} .
$$

Proof We can verify that

$$
\begin{aligned}
D: \text { closed } & \Longleftrightarrow\left\{g_{3}(x) \leq 0,\left(g_{1}+b g_{2}+b^{2} g_{3}\right)(x) \leq 0 \mid b \in \mathbb{R}\right\}: \mathrm{Q}-\mathrm{CCCQ} \\
& \Longleftrightarrow\{G(x) \preceq 0\}: \mathrm{VQ}-\mathrm{CCCQ} .
\end{aligned}
$$

By using Theorem 3.1, we can proof this corollary.
We compare the condition " $D$ is closed" with previous constraint qualifications. In [9], the following constraint qualification, Farkas-Minkowski (FM), was investigated. Let $I$ be an index set, for each $i \in I$, let $g_{i} \in \Gamma_{0}(X)$. The convex system $\left\{g_{i}(x) \leq 0 \mid i \in I\right\}$ is said to be FM if cone co $\bigcup_{i \in I}$ epi $g_{i}^{*}$ is $w^{*}$-closed. Also, FM is closely related to the closedness of $D$ and the Slater type constraint qualification, "there exists $x \in X$ such that for all $w \in S_{X^{*}}\langle w, x\rangle<\inf _{b \in \mathbb{R} \cup\{\gamma\}} k_{(b, w)}^{-1}(0)$ ", assures that $D$ is closed. Actually,

$$
\begin{aligned}
D: \text { closed } & \Longleftrightarrow\left\{g_{3}(x) \leq 0,\left(g_{1}+b g_{2}+b^{2} g_{3}\right)(x) \leq 0 \mid b \in \mathbb{R}\right\}: \text { Q-CCCQ } \\
& \Longleftrightarrow\left\{w(x)-k_{(b, w)}^{-1}(0) \leq 0 \mid b \in \mathbb{R} \cup\{\gamma\}, w \in S_{X^{*}}\right\}: \mathrm{FM} \\
& \Longleftrightarrow\left\{w(x)-\inf _{b \in \mathbb{R} \cup\{\gamma\}} k_{(b, w)}^{-1}(0) \leq 0 \mid w \in S_{X^{*}}\right\}: \mathrm{FM}
\end{aligned}
$$

and the Slater constraint qualification implies the FM, in detail see $[1,4,9]$.
Similarly, we can show the following corollary.
Corollary 4.2 Let $A=\{x \in X \mid G(x) \preceq 0\}$, $x_{0} \in A$, and $S_{X^{*}}\left(x_{0}\right)=\left\{w \in S_{X^{*}} \mid\right.$ $\left.\langle w, x\rangle=\inf _{b \in \mathbb{R} \cup\{\gamma\}} k_{(b, w)}^{-1}(0)\right\}$. Then, the following conditions are equivalent.
(i) $N_{A}\left(x_{0}\right)=\mathrm{cone} \operatorname{co} S_{X^{*}}\left(x_{0}\right)$,
(ii) for each $f \in \Gamma_{0}(X)$ with $\operatorname{dom} f \cap A \neq \emptyset$ and epif $f^{*}+\mathrm{epi} \delta_{A}^{*}$ is $w^{*}$-closed, $x_{0}$ is a minimizer of $f$ in $A$ if and only if there exists $\lambda \in \mathbb{R}_{+}^{\left(S_{X^{*}}\right)}$ such that $\lambda_{w}=0$ for all $w \in S_{X^{*}} \backslash S_{X^{*}}\left(x_{0}\right)$, and

$$
0 \in \partial f\left(x_{0}\right)+\sum_{w \in S_{X^{*}}} \lambda_{w} w
$$

Proof We can verify that $\{G(x) \preceq 0\}$ satisfies VQ-BCQ if and only if $N_{A}\left(x_{0}\right)=$ cone $\operatorname{co} S_{X^{*}}\left(x_{0}\right)$. By using Theorem 3.2, we can proof this corollary.

## 5 Conclusion

We define constraint qualifications the VQ-CCCQ and the VQ-BCQ for quasiconvex vector-valued systems, and we prove the VQ-CCCQ is the weakest constraint qualifications for Lagrangian-type strong duality (Theorem 3.1), and the VQ-BCQ is one for Lagrangian-type min-max duality (Theorem 3.2).

The following table shows the weakest constraint qualifications of real/vectorvalued convex/quasiconvex inequality systems for Lagrangian strong/min-max dualities.

|  | strong | min-max |
| :---: | :---: | :---: |
| real-valued convex | FM $[9]$ | BCQ [3] |
| vector-valued convex | CCCQ [1,2] | $[\mathrm{CQ} 2][2]$ |
| real-valued quasiconvex | Q-CCCQ [4] | Q-BCQ [5] |
| vector-valued quasiconvex | VQ-CCCQ(Theorem 3.1) | VQ-BCQ(Theorem 3.2) |

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