

Proposition 51.28. *For any proper convex function f on \mathbb{R}^n and for any vector $x \in \mathbb{R}^n$, the following conditions on a vector $y \in \mathbb{R}^n$ are equivalent.*

$$(a) \ y \in \partial f(x).$$

$$(b) \ \text{The function } \langle z, y \rangle - f(z) \text{ achieves its supremum in } z \text{ at } z = x.$$

$$(c) \ f(x) + f^*(y) \leq \langle x, y \rangle.$$

$$(d) \ f(x) + f^*(y) = \langle x, y \rangle.$$

If $(\text{cl}(f))(x) = f(x)$, then there are three more conditions all equivalent to the above conditions.

$$(a^*) \ x \in \partial f^*(y).$$

$$(b^*) \ \text{The function } \langle x, z \rangle - f^*(z) \text{ achieves its supremum in } z \text{ at } z = y.$$

$$(a^{**}) \ y \in \partial(\text{cl}(f))(x).$$

The following results are corollaries of Proposition 51.28; see Rockafellar [138] (Corollaries 23.5.1, 23.5.2, 23.5.3).

Corollary 51.29. *For any proper convex function f on \mathbb{R}^n , if f is closed, then $y \in \partial f(x)$ iff $x \in \partial f^*(y)$, for all $x, y \in \mathbb{R}^n$.*

Corollary 51.29 states a sort of adjunction property.

Corollary 51.30. *For any proper convex function f on \mathbb{R}^n , if f is subdifferentiable at $x \in \mathbb{R}^n$, then $(\text{cl}(f))(x) = f(x)$ and $\partial(\text{cl}(f))(x) = \partial f(x)$.*

Corollary 51.30 shows that the closure of a proper convex function f agrees with f wherever f is subdifferentiable.

Corollary 51.31. *For any proper convex function f on \mathbb{R}^n , for any nonempty closed convex subset C of \mathbb{R}^n , for any $y \in \mathbb{R}^n$, the set $\partial\delta^*(y|C) = \partial I_C^*(y)$ consists of the vectors $x \in \mathbb{R}^n$ (if any) where the linear form $z \mapsto \langle z, y \rangle$ achieves its maximum over C .*

There is a notion of approximate subgradient which turns out to be useful in optimization theory; see Bertsekas [19, 17].

Definition 51.17. Let $f: \mathbb{R}^n \rightarrow \mathbb{R} \cup \{+\infty\}$ be any proper convex function. For any $\epsilon > 0$, for any $x \in \mathbb{R}^n$, if $f(x)$ is finite, then an ϵ -subgradient of f at x is any vector $u \in \mathbb{R}^n$ such that

$$f(z) \geq f(x) - \epsilon + \langle z - x, u \rangle, \quad \text{for all } z \in \mathbb{R}^n.$$

See Figure 51.23. The set of all ϵ -subgradients of f at x is denoted $\partial_\epsilon f(x)$ and is called the ϵ -subdifferential of f at x .