Optimization Lab.

Nonlinear programming

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Solution of homework II

Problem 0.1. (Text, 3.1)

(a) By the assumption, $\frac{b-x}{b-a} \ge 0$ and $\frac{x-a}{b-a} \ge 0$. Moreover, $\frac{b-x}{b-a} + \frac{x-a}{b-a} = 1$. Let $\lambda = \frac{b-x}{b-a}$. Then, $x = b - \lambda(b-a) = (1-\lambda)b + \lambda a$. Thus,

$$f(x) = f((1-\lambda)b + \lambda a) \le \lambda f(a) + (1-\lambda)f(b) = \frac{b-x}{b-a}f(a) + \frac{x-a}{b-a}f(b)$$

(b) $\frac{f(x) - f(a)}{x - a} \le \frac{f(b) - f(a)}{b - a}$:

$$\frac{f(x)-f(a)}{x-a} \le \frac{f(b)-f(a)}{b-a} \quad \Leftrightarrow f(x) - f(a) \le \frac{x-a}{b-a}f(b) - \frac{x-a}{b-a}f(a) \\ \Leftrightarrow f(x) \le \frac{x-a}{b-a}f(b) + \frac{b-x}{b-a}f(a).$$

Thus, the first inequality holds by (a). The second inequality can be proved in a similar way.

(c) Since
$$\frac{f(x) - f(a)}{x - a} \le \frac{f(b) - f(a)}{b - a}$$
 for all $x \in (a, b)$,
$$\lim_{x \to a} \frac{f(x) - f(a)}{x - a} = f'(a) \le \lim_{x \to a} \frac{f(b) - f(a)}{b - a} = \frac{f(b) - f(a)}{b - a}.$$

Similarly, $f'(b) \ge \frac{f(b) - f(a)}{b - a}$.

(d) From (c),

$$f(b) \ge f(a) + f'(a)(b-a), \quad f(a) \ge f(b) + f'(b)(a-b).$$

By summing these two inequalities we have

$$(b-a)(f'(b) - f'(a)) \ge 0 \implies \frac{f'(b) - f'(a)}{b-a} \ge 0.$$

By limiting $b \to a$, we obtain $f''(a) \ge 0$. $f''(b) \ge 0$ can be proved in a similar way.

Problem 0.2. (Text, 3.2)

Problem 0.3. (Text, 3.3) First, since f is increasing, $f(x_1) = y_1$, $f(x_2) = y_2$, and $y_1 < y_2$ implies $x_1 < x_2$. Thus, g is an increasing function. Moreover,

$$g((1 - \lambda)y_1 + \lambda y_2) = g((1 - \lambda)f(x_1) + \lambda f(x_2))$$

$$\geq g(f((1 - \lambda)x_1 + \lambda x_2))$$

$$= (1 - \lambda)x_1 + \lambda x_2$$

$$= (1 - \lambda)g(y_1) + \lambda g(y_2).$$

Thus, g is a concave function.

Problem 0.4. (Text, 3.6)

- (i) epif is a halfspace iff hypof is a halfspace. Thus, f is both convex and concave, and hence affine function.
- (ii) If $\operatorname{epi} f$ is a convex cone, for $\alpha, \beta \geq 0$ and $(x, f(x)), (y, f(y)) \in \operatorname{epi} f$,

$$f(\alpha x + \beta y) \le \alpha f(x) + \beta f(y).$$

It implies that f is a sublinear function.

(iii) epif is a polyhedron iff

$$epif = \{(x,t) | \langle s_j, x \rangle + \alpha_j t \le b_j \text{ for } j \in J \}.$$

For this set to be an epigraph each α_j should be nonpositive, so we can assume that $\alpha_j = -1$ for all j. Furthermore, we may denote by $\{1, \ldots, m\} \subseteq J$ such that $\alpha_j = -1$ and by $\{m + 1, \ldots, m + p\}$ the rest. With these notations,

$$f(x) := \max\{\langle s_j, x \rangle - b_j : j = 1, \dots, m\}$$

defines a polyhedron epigraph.

Problem 0.5. (Text, 3.7) Suppose that f(x) < M for all $x \in \mathbb{R}^n$. By the first-order conditions for convexity,

$$f(y) \ge f(x) + \nabla f(x)^T (y - x), \quad \forall x, y \in \mathbb{R}^n$$

Suppose that f is not constant, which means that there exists a $\bar{x} \in \mathbb{R}^n$ such that $\nabla f(\bar{x}) \neq 0$. For any positive $\alpha > 0$ let $y := \bar{x} + \alpha \nabla f(\bar{x})$. Then, first-order condition implies that

$$f(y) \ge f(\bar{x}) + \alpha \|\nabla f(\bar{x})\|_2^2$$

Let $\alpha = \frac{M - f(\bar{x})}{\|\nabla f(\bar{x}\|_2^2)}$. Then,

$$f(\frac{M - f(\bar{x})}{\|\nabla f(\bar{x})\|_2^2} \nabla f(\bar{x})) \ge M,$$

a contradiction.

Problem 0.6. (Text, 3.9)

(a) By the second order condition \tilde{f} is convex if and only if $\nabla^2 \tilde{f}(x) \succeq 0$ for all $x \in \text{dom}\tilde{f}$. Thus,

$$\nabla^2 \tilde{f}(x = Fz + \hat{x}) = \nabla^2 f(Fz + \hat{x}) = F^T f(Fz + \hat{x})F \succeq 0.$$

(b) From (a) \tilde{f} is convex if and only if $F^T \nabla^2 \tilde{f} (Fz + \hat{x}) F \succeq 0$. This means that

$$y^T F^T \nabla^2 \tilde{f} (Fz + \hat{x}) Fy \ge 0, \forall y \ge 0$$

Since the null space of A is equal to the range space of F,

$$w^T \nabla^2 \tilde{f}(Fz + \hat{x}) w \ge 0, \forall w \in \mathcal{N}(A).$$

Thus, by the hint, the proof is completed.

Problem 0.7. (Text, 3.20)

(a)

$$||A((1-\lambda)x + \lambda y) - b|| = ||(1-\lambda)(Ax - b) + \lambda(Ay - b)|| \leq (1-\lambda)||Ax - b|| + \lambda||Ay - b||.$$

where the second inequality is due to the homogeneity and metric property of $\|\cdot\|$.

- (b) Not yet...
- (c) Not yet...

Problem 0.8. (Text, 3.22) In this problem we will express f(x) as h(g(x)), and then apply the composition rules.

- (a) f(x) = h(g(x)) where $g(x) = -\log(\sum_{i=1}^{n} e^{a_i^T x + b_i})$ and $h(x) = -\log x$. Then, g(x) is concave, and h(x) is convex and nonincreasing. Thus, f(x) is convex.
- (b) f(x, u, v) = h(u, g(x, u, v)) where $g(x, u, v) = v x^T x/u$ and $h(u, y) = -\sqrt{uy}$. Then, g(x) is concave, and h(u, y) is convex and nonincreasing. Thus, f(x, u, v) is convex.
- (c) f(x, u, v) = h(g(x, u, v)) where g(x, u, v) is a concave function defined as the negative of (b), and $h(x) = -2 \log x$ which is a convex and nonincreasing. Thus, f(x, u, v) is convex.
- (d) f(x,t) = h(g(x,t),t) where $g(x,t) = t \frac{\|x\|_p^p}{t^{p-1}}$ and $h(x,y) = -x^{1/p}y^{1-1/p}$. Then, g(x,t) is concave, and h(x,y) is convex and nonincreasing. Thus, f(x,t) is convex.
- (e) f(x,t) = h(g(x,t)) where g(x,t) is a concave function defined as the negative of (d), and $h(x) = -p \log x$ which is a convex and nonincreasing. Thus, f(x,t) is convex.

Problem 0.9. (Text, 3.32)

- (a) (fg)'' = (f''g + 2f'g' + fg''). By assumption, f, g > 0 and $f'', g'' \ge 0$ and hence $f''g, fg'' \ge 0$. Since either $f', g' \ge 0$ or $f', g' \le 0$, $f'g' \ge 0$. Thus, $(fg)'' \ge 0$, and hence fg is convex.
- (b) Since $f'', g'' \leq 0$, and f, g > 0 and $f'g' \leq 0$, $(fg)'' \leq 0$, and hence fg is concave.

(c) Similar to (a) and (b).

Problem 0.10. (Text, 3.43) It suffices to prove the result for a function on \mathbb{R} . Recall that a function f is quasiconvex iff for all $x, y \in \text{dom} f$ and $0 < \lambda < 1$,

$$f((1-\lambda)x + \lambda y) \le \max\{f(x), f(y)\}.$$

only if part: Suppose that f is quasiconvex and $f(y) \leq f(x)$. By differentiability of f at x, we have

$$f((1-\lambda)x + \lambda y) - f(x) = \lambda f'(x)(y-x) + \lambda |y-x|\alpha(x;\lambda(y-x))$$

for $\lambda \in (0,1)$ where $\alpha(x; \lambda(y-x)) \to 0$ as $\lambda \to 0$. By the quasiconvexity of f, we have $f((1-\lambda)x + \lambda y) - f(x) = \lambda f'(x)(y-x) + \lambda |x-y|\alpha(x; \lambda(y-x)) \leq 0$. Dividing by λ and letting $\lambda \to 0$, we get

$$f'(x)(y-x) \le 0.$$

if part: Suppose that x and y are any two points such that $f(y) \leq f(x)$. We assume that y < x (A proof for y > x is similar). Suppose f is not quasiconvex, i.e., there exists $z \in (y, x)$ such that $f(z) > \max\{f(x), f(y)\}$. By continuity of f, there exists $z' \in (z, x)$ such that f'(z') < 0 and $f(y) \leq f(z')$. Then,

$$f'(z')(y-z') > 0,$$

a contradiction.