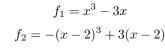
Fundamentals of Optimization Homework 1

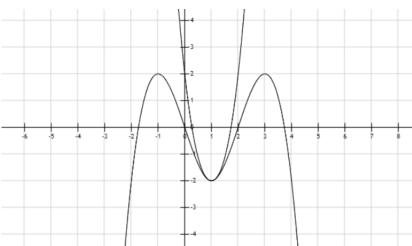
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(5.1)

Give a counterexample here (to prove that the proposition is wrong)





both of them are non-convex functions while their epigraphs' intersection is a convex set.

(5.2)

1. f is convex function -> a >= 0

because f is convex, for x1, x2 (in domain), $\lambda \in [0, 1]$

$$f(\lambda x_1 + (1 - \lambda)x_2) = a(\lambda x_1 + (1 - \lambda)x_2)^2 + b(\lambda x_1 + (1 - \lambda)x_2) + c < = \lambda ax_1^2 + (1 - \lambda)ax_2^2 + b(\lambda x_1 + (1 - \lambda)x_2) + c = \lambda f(x_1) + (1 - \lambda)f(x_2)$$

since from the deature of quadratic function, we know $(\lambda x_1 + (1 - \lambda)x_2)^2 \le \lambda x_1^2 + (1 - \lambda)x_2^2$. so only when a >= 0, the inequality is tenable.

2. a >= 0 -> f is convex function

when a > 0, consider the feature of quadratic function and the inequality function above, we can prove that f is convex. when a = 0, f(x) = bx + c, and either $b \neq 0$ (linear function) or b = 0, f(x) = c (constant function), f(x) is always convex. In this case, the proposition is right.

(6.1)

for any $x_1, x_2 \in L_a(f)$, due to the feature that $ifx \neq y$, $then f(x) \neq f(y)$, we know $x_1 = x_2$ so that $f(x_1) = f(x_2) = a$ in this case, for $\lambda \in [0, 1]$, we have $f(\lambda x_1 + (1 - \lambda)x_2) = \lambda f(x_1) + (1 - \lambda)f(x_2)$, so we prove that the level set for an injective function is convex set

(6.2)

we consider a counterexample that f is a constant function, like f = 0. linear function's level set is convex, and it's the same for our constant function f.

now we cannot say our f = 0 is a one-to-one fuction, since for any x in domain, f(x) will be the same value, 0. so this proposition is not true.