
INEQUALITES COOKBOOK

A PREPRINT

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ABSTRACT

A collection of inequalities with some proofs.

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Keywords Inequalities

1 Convex Functions

If f is convex, then for all x, y and $t \in [0, 1]$

$$f(tx + (1-t)y) \leq tf(x) + (1-t)f(y) \quad (1)$$

If f is differentiable for all x, y , convexity is equivalent to

- The first-order condition

$$f(y) \geq f(x) + \langle \nabla f(x), y - x \rangle \quad (2)$$

- Montone gradient condition

$$\langle f(x) - \nabla f(y), x - y \rangle \geq 0 \quad (3)$$

2 Strong Convexity

A differentiable function is strongly convex if for some $m > 0$ and all x, y

$$f(y) \geq f(x) + \langle \nabla f(x), y - x \rangle + \frac{m}{2} \|y - x\|^2. \quad (4)$$

The following conditions are equivalent to strong-convexity

- Using the first-order condition of convexity, g is convex

$$g(x) = f(x) - \frac{\mu}{2} \|x\|^2 \quad (5)$$

- Using monotone gradient condition of convexity for g ,

$$\langle \nabla f(x) - \nabla f(y), x - y \rangle \geq \mu \|x - y\|^2 \quad (6)$$

- Using convexity of g , for $t \in [0, 1]$

$$f(tx + (1-t)y) \leq tf(x) + (1-t)f(y) - \frac{t(1-t)}{2} \|x - y\|^2. \quad (7)$$

The following are implied by strong convexity

- Minimizing the strong convexity condition w.r.t y , gives the Polyack-Lojasiewicz inequality

$$\frac{1}{2} \|\nabla f(x)\|^2 \geq \mu(f(x) - f(x^*)) \quad (8)$$

- Using Cauchy-Schwartz

$$\|\nabla f(x) - \nabla f(y)\| \geq \mu \|x - y\| \quad (9)$$

- Using $\phi_x(z) = f(z) - \langle \nabla f(x), z \rangle$ is strongly convex with the same μ ,

$$f(y) \leq f(x) + \langle \nabla f(x), y - x \rangle + \frac{1}{2\mu} \|\nabla f(y) - \nabla f(x)\|^2 \quad (10)$$

- Trivially,

$$\langle \nabla f(x) - \nabla f(y), x - y \rangle \leq \frac{1}{\mu} \|\nabla f(x) - \nabla f(y)\|^2 \quad (11)$$

If $\|\nabla^2 f\| \geq m$, then f is strongly convex.

3 Lipschitz Gradient

A differentiable function f has L -Lipschitz continuous gradient if for some $L > 0$

$$\|\nabla f(x) - \nabla f(y)\| \leq L\|x - y\| \quad (12)$$

This implies

$$g(x) = \frac{L}{2}\|x\|^2 - f(x) \quad (13)$$

is convex. As such, the following conditions hold

$$f(y) \leq f(x) + \langle \nabla f(x), y - x \rangle + \frac{L}{2}\|y - x\|^2 \quad (14)$$

$$\langle \nabla f(x) - \nabla f(y), x - y \rangle \leq L\|x - y\|^2 \quad (15)$$

$$f(tx + (1 - t)y) \geq tf(x) + (1 - t)f(y) - \frac{t(1 - t)L}{2}\|x - y\|^2 \quad (16)$$

If f is assumed to be convex, then the above relationships are equivalent. Moreover, the following conditions are also equivalent

$$f(y) \geq f(x) + \langle \nabla f(x), y - x \rangle + \frac{1}{2L}\|\nabla f(y) - \nabla f(x)\|^2 \quad (17)$$

$$\langle \nabla f(x) - \nabla f(y), x - y \rangle \geq \frac{1}{L}\|\nabla f(x) - \nabla f(y)\|^2 \quad (18)$$

$$f(tx + (1 - t)y) \leq tf(x) + (1 - t)f(y) - \frac{t(1 - t)}{2L}\|\nabla f(x) - \nabla f(y)\|^2 \quad (19)$$

4 Asymptotically Convex

4.1 Asymptotically Strongly Convex

A function is asymptotically strongly convex if for all $\|x_0\|, \|x_1\| \geq R$, for some $R > 0$

$$\langle \nabla f(x_0) - \nabla f(x_1), x_0 - x_1 \rangle \geq m\|x_0 - x_1\|^2. \quad (20)$$

That is, outside of a ball centered at the origin, f is strongly convex.

If f is assumed to have Lipschitz continuous gradient, then there exists some $M > 0$ such that

$$\langle \nabla f(x_0) - \nabla f(x_1), x_0 - x_1 \rangle \geq m\|x_0 - x_1\|^2 - M, \quad \forall x, y. \quad (21)$$

Proof. Since ∇f is L -Lipschitz

$$|\langle \nabla f(x_0) - \nabla f(x_1), x_0 - x_1 \rangle| \leq \|\nabla f(x_0) - \nabla f(x_1)\| \|x_0 - x_1\| \quad (22)$$

$$\leq L\|x_0 - x_1\|^2. \quad (23)$$

Then for $\|x_0\|, \|x_1\| \leq R$,

$$|\langle \nabla f(x_0) - \nabla f(x_1), x_0 - x_1 \rangle| \leq LR^2, \quad (24)$$

and

$$m\|x_0 - x_1\|^2 \leq mR^2. \quad (25)$$

Thus, for all x_0, x_1

$$\langle \nabla f(x_0) - \nabla f(x_1), x_0 - x_1 \rangle \geq m\|x_0 - x_1\|^2 - (m + L)R^2. \quad (26)$$

□

4.2 Asymptotically Strongly Co-coercive

This definition is given in [1]. There exists $R \geq 0, K, L_2 > 0$ such that

$$\langle x - y, \nabla f(x) - \nabla f(y) \rangle \geq K\|x - y\|^2 + \frac{1}{L_2} \|\nabla f(x) - \nabla f(y)\|, \quad (27)$$

for $\|x - y\| \geq R$.

References

- [1] Nawaf Bou-Rabee and Katharina Schuh. Nonlinear hamiltonian monte carlo and its particle approximation, 2023.