Jensen's inequality for nonconvex functions*

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Abstract. Jensen's inequality is formulated for convexifiable (generally nonconvex) functions.

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1. Introduction

Jensen's inequality is 100 years old, e.g., [1, 2, 3]. It says that the value of a convex function at a point, which is a convex combination of finitely many points, is less than or equal to the convex combination of values of the function at these points. Using symbols: If: $\mathbb{R}^n \to \mathbb{R}$ is convex then

$$f\left(\sum_{i=1}^{p} \lambda_i \mathbf{x}^i\right) \le \sum_{i=1}^{p} \lambda_i f(\mathbf{x}^i) \tag{1}$$

for every set of p points $\mathbf{x}^i, i = 1, \dots, p$, in the Euclidean space \mathbb{R}^n and for all real scalars $\lambda_i \geq 0, i = 1, \dots, p$, such that $\sum_{i=1}^p \lambda_i = 1$.

In this note the inequality (1) is extended from convex to convexifiable functions, e.g., [4, 5]. These include all twice continuously differentiable functions, all once continuously differentiable functions with Lipschitz derivative and all analytic functions. As a special case we obtain a new form of the arithmetic mean theorem.

2. Convexifiable functions

If $f: \mathbb{R}^n \to \mathbb{R}$ is a continuous function in n variables defined on a convex set C of \mathbb{R}^n , then the function is said to be convex on C if

$$f(\lambda \mathbf{x} + (1 - \lambda)\mathbf{y}) \le \lambda f(\mathbf{x}) + (1 - \lambda)f(\mathbf{y})$$
 (2)

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for every $\mathbf{x}, \mathbf{y} \in C$ and scalar $0 \le \lambda \le 1$. Note that this is (1) for p = 2. Let us recall several recent results.

Definition 1 [[5]]. Given a continuous $f: \mathbb{R}^n \to \mathbb{R}$ defined on a convex set C, consider the function $\varphi: \mathbb{R}^{n+1} \to \mathbb{R}$ defined by $\varphi(\mathbf{x}, \alpha) = f(\mathbf{x}) - \frac{1}{2}\alpha\mathbf{x}^T\mathbf{x}$, where \mathbf{x}^T is the transposed of \mathbf{x} . If $\varphi(\mathbf{x}, \alpha)$ is a convex function on C for some $\alpha = \alpha^*$, then $\varphi(\mathbf{x}, \alpha)$ is a convexification of f and α^* is its convexifier on C. Function f is convexifiable if it has a convexification.

Observation 1. If α^* is a convexifier of f, then so is every $\alpha \leq \alpha^*$.

In order to characterize a convexifiable function, the $\mathsf{mid}\text{-}\mathsf{point}$ acceleration function

$$\Psi(\mathbf{x}, \mathbf{y}) = \frac{4}{\|\mathbf{x} - \mathbf{y}\|^2} \left(f(\mathbf{x}) + f(\mathbf{y}) - 2f\left(\frac{\mathbf{x} + \mathbf{y}}{2}\right) \right), \quad \mathbf{x}, \mathbf{y} \in C, \mathbf{x} \neq \mathbf{y}$$

was introduced in [5]. There it was shown that a continuous $f: \mathbb{R}^n \to \mathbb{R}$, defined on a nontrivial convex set C (i.e., a convex set with at least two distinct points) in \mathbb{R}^n is convexifiable on C if, and only if, its mid-point acceleration function Ψ is bounded from below on C.

For two important classes of functions a convexifier α can be given explicitly. If f is twice continuously differentiable then its second derivative at \mathbf{x} is represented by the Hessian matrix $H(\mathbf{x}) = (\partial^2 f(\mathbf{x})/\partial \mathbf{x}_i \partial \mathbf{x}_j)$. This is a symmetric matrix with real eigenvalues. Denote its smallest eigenvalue by $\lambda(\mathbf{x})$ and its "globally" smallest eigenvalue over a compact convex set C by

$$\lambda^* = \min_{\mathbf{x} \in C} \lambda(\mathbf{x}).$$

Lemma 1 [[4, 5]]. Given a twice continuously differentiable function $f : \mathbb{R}^n \to \mathbb{R}$ on a nontrivial compact convex set C in \mathbb{R}^n . Then $\alpha = \lambda^*$ is a convexifier.

We say that a continuously differentiable function f has Lipschitz derivative if $|\nabla f(\mathbf{x}) - \nabla f(\mathbf{y})| (\mathbf{x} - \mathbf{y})| \le L ||\mathbf{x} - \mathbf{y}||$ for every $\mathbf{x}, \mathbf{y} \in C$ and some constant L. Here $\nabla f(\mathbf{u})$ is the (Fréchet) derivative of f at \mathbf{u} and $||\mathbf{u}|| = (\mathbf{u}^T \mathbf{u})^{1/2}$ is the Euclidean norm. We represent the derivative at \mathbf{x} as a row n-tuple gradient $\nabla f(\mathbf{x}) = (\partial f(\mathbf{x})/\partial \mathbf{x}_i)$.

Lemma 2 [[5]]. Given a continuously differentiable function $f : \mathbb{R}^n \to \mathbb{R}$ with Lipschitz derivative and a constant L on a nontrivial compact convex set C in \mathbb{R}^n . Then $\alpha = -L$ is a convexifier.

One can show that every convexifiable scalar function $f: \mathbb{R} \to \mathbb{R}$ is Lipschitz, i.e, $|f(s) - f(t)| \leq K|s - t|$ for every s and t and some constant K. This means that a scalar non-Lipschitz function is not convexifiable. However, almost all smooth functions of practical interest are convexifiable; e.g., [5].

3. Jensen's inequality for convexifiable functions

In this section we formulate (1) for convexifiable functions.

Theorem 1 [Jensen's inequality for convexifiable functions]. Consider a convexifiable function $f: \mathbb{R}^n \to \mathbb{R}$ on a bounded nontrivial convex set C of \mathbb{R}^n and its convexifier α . Then

$$f\left(\sum_{i=1}^{p} \lambda_i \mathbf{x}^i\right) \le \sum_{i=1}^{p} \lambda_i f(\mathbf{x}^i) - \frac{\alpha}{2} \left(\sum_{\substack{i,j=1\\i \le j}}^{p} \lambda_i \lambda_j \|\mathbf{x}^i - \mathbf{x}^j\|^2\right)$$
(3)

for every set of p points \mathbf{x}^i , i = 1, ..., p, in C and all real scalars $\lambda_i \geq 0$, i = 1, ..., p, with $\sum_{i=1}^p \lambda_i = 1$.

Proof. Since f is convexifiable, $\varphi(\mathbf{x}, \alpha) = f(\mathbf{x}) - \frac{1}{2}\alpha\mathbf{x}^T\mathbf{x}$ is a convex function for every convexifier α . Hence Jensen's inequality works for $\varphi(\mathbf{x}, \alpha)$. After substitution one obtains

$$f\left(\sum_{i=1}^{p} \lambda_i \mathbf{x}^i\right) \le \sum_{i=1}^{p} \lambda_i f(\mathbf{x}^i) - \frac{\alpha}{2} \left(\sum_{i,j=1}^{p} \lambda_i \lambda_j (\mathbf{x}^i)^T (\mathbf{x}^i - \mathbf{x}^j)\right).$$

After more rearranging the more pleasing form (3) follows.

Using the fact that for a convex function f one can choose the convexifier $\alpha=0$, one recovers (1). For a twice continuously differentiable function one can specify $\alpha=\lambda^*$ (by $Lemma\ 1$) and for a continuously differentiable function with Lipschitz derivative and its constant L, one can specify $\alpha=-L$ (by $Lemma\ 2$). Hence we have, respectively, the following special cases:

Corollary 1 [Jensen's inequality for twice continuously differentiable functions]. Given a twice continuously differentiable function $f: \mathbb{R}^n \to \mathbb{R}$ on a nontrivial compact convex set C in \mathbb{R}^n . Then

$$f\left(\sum_{i=1}^{p} \lambda_i \mathbf{x}^i\right) \le \sum_{i=1}^{p} \lambda_i f(\mathbf{x}^i) - \frac{\lambda^*}{2} \left(\sum_{\substack{i,j=1\\i < j}}^{p} \lambda_i \lambda_j \|\mathbf{x}^i - \mathbf{x}^j\|^2\right)$$
(4)

for every set of p points $\mathbf{x}^i, i = 1, \dots, p$, in C and all real scalars $\lambda_i \geq 0$, $i = 1, \dots, p$, with $\sum_{i=1}^p \lambda_i = 1$.

Observation 2. If f in Corollary 1 is strictly convex, then the lowest eigenvalue of the Hessian is $\lambda^* \geq 0$ (often $\lambda^* > 0$) and (4) may provide a better bound than (1). Since every analytic function $f: \mathbb{R} \to \mathbb{R}$ is twice continuously differentiable, Corollary 1 holds, in particular, for analytic functions with $\lambda^* = \min_{t \in C} f''(t)$.

Corollary 2 [Jensen's inequality for once continuously differentiable functions with Lipschitz derivative]. Given a continuously differentiable function $f: \mathbb{R}^n \to \mathbb{R}$ with Lipschitz derivative and a constant L on a nontrivial compact convex set C in \mathbb{R}^n . Then

$$f\left(\sum_{i=1}^{p} \lambda_i \mathbf{x}^i\right) \le \sum_{i=1}^{p} \lambda_i f(\mathbf{x}^i) + \frac{L}{2} \left(\sum_{\substack{i,j=1\\i < j}}^{p} \lambda_i \lambda_j \|\mathbf{x}^i - \mathbf{x}^j\|^2\right)$$
 (5)

for every set of p points \mathbf{x}^i , i = 1, ..., p, in C and all real scalars $\lambda_i \geq 0$, i = 1, ..., p, with $\sum_{i=1}^p \lambda_i = 1$.

Special Case: For a scalar function $f:\mathbb{R}\to\mathbb{R}$ and two scalar points a and b Jensen's inequality is

$$f(\lambda a + (1 - \lambda)b) \le \lambda f(a) + (1 - \lambda)f(b)$$
, for every $0 \le \lambda \le 1$

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while for a convexifiable f, it is

$$f(\lambda a + (1 - \lambda)b) \le \lambda f(a) + (1 - \lambda)f(b) - \frac{\alpha}{2}\lambda(1 - \lambda)(a - b)^2$$

for every convexifier α and for every $0 \le \lambda \le 1$. We will use this special case to illustrate the basic difference between the two inequalities.

Illustration 1. Consider $f(t) = \sin t$ on $0 \le t \le 2\pi$. Take a = 0 and $b = 2\pi$. Then (1) and its extension yield, respectively

$$\sin(2\pi(1-\lambda)) \le 0, \qquad 0 \le \lambda \le 1 \tag{6}$$

and

$$\sin(2\pi(1-\lambda)) \le 2\pi^2 \lambda(1-\lambda), \qquad 0 \le \lambda \le 1. \tag{7}$$

Inequality (6) is not satisfied on the region where f(t) is not convex, i.e., $1/2 \le \lambda \le 1$. On the other hand the new upper bound in (7) holds (see Figure 1).

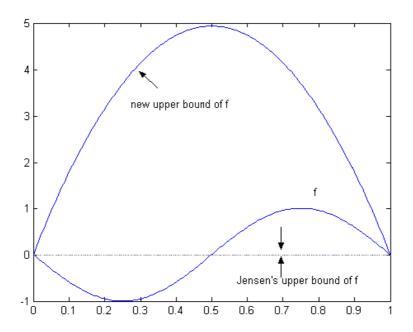


Figure 1. Jensen's inequality for a convexifiable function

A situation where the new bound is sharper than the one provided by Jensen's inequality for a convex function is illustrated in the following example.

Illustration 2. Consider $f(t) = t^4$ between a = 1 and b = 2. Then (1) and its extension yield $(2 - \lambda)^4 \le 16 - 15\lambda$ and $(2 - \lambda)^4 \le 16 - 9\lambda - 6\lambda^2$, $0 \le \lambda \le 1$, respectively. The upper bounds are compared against the original function in Figure 2.

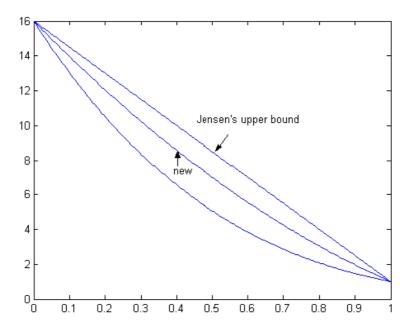


Figure 2. Improvement for a strictly convex function

Jensen's inequality is closely related to the arithmetic mean theorem for real numbers. The following theorem says that the value of a convex function at the arithmetic mean of p numbers is less than or equal to the arithmetic mean of the values of the function at these numbers.

Theorem 2 [Classic arithmetic mean theorem for convex functions, e.g., [3]]. Consider a convex scalar function $f: \mathbb{R} \to \mathbb{R}$ on a nontrivial compact interval [a,b]. Then

$$f\left(\frac{1}{p}\sum_{i=1}^{p}t_{i}\right) \leq \frac{1}{p}\sum_{i=1}^{p}f(t_{i})\tag{8}$$

for every set of p points $t_i \in [a, b], i = 1, ..., p$.

Specifying $\mathbf{x}^i = t_i$, $\lambda_i = 1/p$, $i = 1, \ldots, p$, in (3) one obtains, after rearrangement, the following extension:

Theorem 3 [Arithmetic mean theorem for convexifiable functions]. Consider a convexifiable scalar function $f : \mathbb{R} \to \mathbb{R}$ on a nontrivial compact interval [a,b] and its convexifier α . Then

$$f\left(\frac{1}{p}\sum_{i=1}^{p}t_{i}\right) \leq \frac{1}{p}\sum_{i=1}^{p}f(t_{i}) - \frac{\alpha}{2}\left(\frac{1}{p}\sum_{i=1}^{p}t_{i}^{2} - \left(\frac{1}{p}\sum_{i=1}^{p}t_{i}\right)^{2}\right)$$
(9)

for every set of p points $t_i \in [a, b], i = 1, \ldots, p$.

Observation 3. In (9) one can set $\alpha = 0$ if f is convex, $\alpha = \lambda^* = \min_{t \in [a,b]} f''(t)$ if f is twice continuously differentiable or $\alpha = -L$ if f is Lipschitz continuously differentiable with a constant L. The first special case recovers the classic result.

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Observation 4. The term corresponding to the convexifier is positive, provided that at least one t_i is non-zero. Indeed, denote $\mathbf{A} = (t_i) \in \mathbb{R}^p$, $\mathbf{E} = (1, \dots, 1)^T \in \mathbb{R}^p$. Then this term is $[(1/p)(\mathbf{A}, \mathbf{A}) - (1/p)^2(\mathbf{A}, \mathbf{E})^2]$. Since $(\mathbf{A}, \mathbf{E})^2 \leq \|\mathbf{A}\|^2 \|\mathbf{E}\|^2 = (\mathbf{A}, \mathbf{A}) \cdot p$ and $p < p^2$, the term is positive. Since for a twice continuously differentiable strictly convex f, we know that $\lambda^* = \min_{t \in [a,b]} f''(t) \geq 0$, it follows that (9) typically provides in this case a better estimate than (8).

Special Case: For a scalar function $f: \mathbb{R} \to \mathbb{R}$ and only two points t_1 and t_2 , (3) (and after some rearrangement (9)) yields

$$f\left(\frac{t_1+t_2}{2}\right) \le \frac{1}{2}(f(t_1)+f(t_2)) - \frac{\alpha}{8} \cdot (t_1-t_2)^2$$

References

- [1] J. L. W. V. Jensen, Om konvexe Funktioner og Uligheder mellem Middel-vaerdier, Nyt Tidsskr. Math. 16B(1905), 49–69.
- [2] J. L. W. V. Jensen, Sur les fonctions convexes et les inegalites entre les valeur moyennes, Acta Math. 30(1906), 175-193.
- [3] A. W. Roberts, D. E. Varberg, *Convex Functions*, Academic Press, New York, 1973.
- [4] S. Zlobec, Estimating convexifiers in continuous optimization, Mathematical Communications 8(2003), 129–137.
- [5] S. Zlobec, *Characterization of convexifiable functions*, McGill University, June 2004, to be published.