# A Companion Inequality to Jensen's Inequality 

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## 1. Introduction

In this paper we consider a convex function $f:(a, b) \rightarrow R$, where $-\infty \leqslant a<b \leqslant \infty$. We recall that $f$ is convex if for all $x_{1}, x_{2} \in(a, b)$ and $\alpha_{1}, \alpha_{2} \geqslant 0$ such that $\alpha_{1}+\alpha_{2}=1$,

$$
f\left(\alpha_{1} x_{1}+\alpha_{2} x_{2}\right) \leqslant \alpha_{1} f\left(x_{1}\right)+\alpha_{2} f\left(x_{2}\right) .
$$

$f$ is strictly convex if, in addition, whenever $\alpha_{1} x_{1}+\alpha_{2} x_{2}$ is strictly between $x_{1}$ and $x_{2}$,

$$
f\left(\alpha_{1} x_{1}+\alpha_{2} x_{2}\right)<\alpha_{1} f\left(x_{1}\right)+\alpha_{2} f\left(x_{2}\right) .
$$

If $f$ is convex, it is continuous; its left and right derivatives, $f_{-}^{\prime}$ and ${\dot{f^{\prime}}}_{+}^{\prime}$, exist, are finite, and are non-decreasing; $f_{-}^{\prime} \leqslant f_{+}^{\prime}$, and except for at most countably many $x \in(a, b), f_{-}^{\prime}(x)=f^{\prime}(x)=f_{+}^{\prime}(x)$.

Jensen's inequality can be stated as follows:
Suppose that $f$ is convex on $(a, b)$. Then, for $x_{1}, \ldots, x_{n}$ in $(a, b)$ and $p_{1}, \ldots, p_{n} \geqslant 0, p_{1}+\cdots+p_{n}>0$,

$$
\begin{equation*}
f\left(\frac{p_{1} x_{1}+\cdots+p_{n} x_{n}}{p_{1}+\cdots+p_{n}}\right) \leqslant \frac{p_{1} f\left(x_{1}\right)+\cdots+p_{n} f\left(x_{n}\right)}{p_{1}+\cdots+p_{n}} . \tag{1}
\end{equation*}
$$

The inequality in the title of the paper states that, under the additional assumption of monotonicity of $f$, there is a specific point in $(a, b)$ at which the value of $f$ is greater than or equal to the right-hand side of (1).

Theorem 1. Suppose that $f$ is convex and increasing on $(a, b)$. Then for $x_{1}, \ldots, x_{n} \in(a, b), \quad p_{1}, \ldots, p_{n} \geqslant 0, \quad p_{1}+\cdots+p_{n}>0, \quad$ and $p_{1} f_{+}^{\prime}\left(x_{1}\right)+\cdots+p_{n} f_{+}^{\prime}\left(x_{n}\right)>0$, we have

$$
\begin{equation*}
\frac{p_{1} f\left(x_{1}\right)+\cdots+p_{n} f\left(x_{n}\right)}{p_{1}+\cdots+p_{n}} \leqslant f\left(\frac{p_{1} f_{+}^{\prime}\left(x_{1}\right) x_{1}+\cdots+p_{n} f_{+}^{\prime}\left(x_{n}\right) x_{n}}{p_{1} f_{+}^{\prime}\left(x_{1}\right)+\cdots+p_{n} f_{+}^{\prime}\left(x_{n}\right)}\right) . \tag{2}
\end{equation*}
$$

A more general version of Jensen's inequality can be stated as follows:
Suppose that $f$ is convex on $(a, b)$, that $(E, \mathscr{E}, \mu)$ is a probability measure space, and that $X:(E, \mathscr{E}) \rightarrow(a, b)$ is measurable. If $X$ and $f \circ X$ are in $L(\mu)$, then

$$
\begin{equation*}
f\left(\int_{E} X d \mu\right) \leqslant \int_{E}(f \circ X) d \mu \tag{3}
\end{equation*}
$$

The corresponding generalization of Theorem 1 is

Theorem 2. Suppose that $f$ is convex and increasing on $(a, b)$. Suppose also that $X:(E, \mathscr{E}) \rightarrow(a, b)$ is measurable, that $f \circ X, f_{+}^{\prime} \circ X$, and $\left(f_{+}^{\prime} \circ X\right) X$ are in $L(\mu)$ and that $\int_{E}\left(f_{+}^{\prime} \circ X\right) d \mu>0$. Then

$$
\begin{equation*}
\int_{E}(f \circ X) d \mu \leqslant f\left(\frac{\int_{E}\left(f_{+}^{\prime} \circ X\right) X d \mu}{\int_{E}\left(f_{+}^{\prime} \circ X\right) d \mu}\right) . \tag{4}
\end{equation*}
$$

$I f$, in addition we assume that $f$ is strictly convex, then equality holds in (4) if and only if $X$ is constant $\mu$ a.e.

Both theorems remain valid if at any occurrence of $f_{+}^{\prime}(x)$ we write instead any value in the interval $\left[f_{-}^{\prime}(x), f_{+}^{\prime}(x)\right]$. Note merely that inequalities (5) and (6) below, on which the proofs depend, continue to hold with such replacement. (" $f_{+}^{\prime}$ " was chosen for ease in stating the theorems.) Furthermore, both theorems are also true if $f$ is a convex and decreasing function; the only changes needed in the hypotheses are to require $p_{1} f_{+}^{\prime}\left(x_{1}\right)+\cdots+p_{n} f_{+}^{\prime}\left(x_{n}\right)<0$ in Theorem 1, and $\int_{E} f_{+}^{\prime} \circ X d \mu<0$ in Theorem 2. For if $f:(a, b) \rightarrow R$ is a convex and decreasing function, then $f:(-b,-a) \rightarrow R$, defined by $f(x)=f(-x)$, is a convex and increasing function, and Theorems 1 and 2 applied to $f$ yield the same theorems with $f$, just as before.

In Section 2, inequalities (2) and (4) will be established and the case of equality will be discussed. Section 3 contains several applications. Section 4 contains concluding remarks.

## 2. The Proofs of the Companion Inequalities

The proofs of (2) and (4) are based on the inequality

$$
\begin{equation*}
f(y) \geqslant f(x)+(y-x) f_{+}^{\prime}(x), \tag{5}
\end{equation*}
$$

valid for any convex function on $(a, b)$ and arbitrary $x, y \in(a, b)$. The proof
that (in the case of strict convexity) equality holds in (4) only if $X$ is constant $\mu$ a.e. depends on the inequality

$$
\begin{equation*}
f(y)>f(x)+(y-x) f_{+}^{\prime}(x) \tag{6}
\end{equation*}
$$

valid for any strictly convex function on $(a, b)$ and arbitrary distinct $x, y \in(a, b)$.

To simplify typesetting, set

$$
A=\frac{p_{1} f_{+}^{\prime}\left(x_{1}\right) x_{1}+\cdots+p_{n} f_{+}^{\prime}\left(x_{n}\right) x_{n}}{p_{1} f_{+}^{\prime}\left(x_{1}\right)+\cdots+p_{n} f_{+}^{\prime}\left(x_{n}\right)}
$$

To prove Theorem 1, observe that $A \in(a, b)$ since $A$ is a convex combination of $x_{1}, \ldots, x_{n}$, and so by (5),

$$
f(A) \geqslant f\left(x_{k}\right)+\left(A-x_{k}\right) f_{+}^{\prime}\left(x_{k}\right), \quad k=1, \ldots, n
$$

Multiply the $k$ th inequality by $p_{k}$ and add the inequalities thus obtained; we find

$$
\left(p_{1}+\cdots+p_{n}\right) f(A) \geqslant \sum_{k=1}^{n} p_{k} f\left(x_{k}\right)+A \sum_{k=1}^{n} p_{k} f_{+}^{\prime}\left(x_{k}\right)-\sum_{k=1}^{n} p_{k} f_{+}^{\prime}\left(x_{k}\right) x_{k}
$$

and Theorem 1 is proved since

$$
A \sum_{k=1}^{n} p_{k} f_{+}^{\prime}\left(x_{k}\right)-\sum_{k=1}^{n} p_{k} f_{+}^{\prime}\left(x_{k}\right) x_{k}=0
$$

We begin the proof of Theorem 2 by establishing inequality (4). The proof is similar to that of (2). Set

$$
A=\frac{\int_{E}\left(f_{+}^{\prime} \circ X\right) X d \mu}{\int_{E}\left(f_{+}^{\prime} \circ X\right) d \mu}
$$

$A \in(a, b)$, and by (5), for every $t \in E$,

$$
f(A) \geqslant\left(f_{+}^{\prime} \circ X\right)(t)+(A-X(t))\left(f_{+}^{\prime} \circ X\right)(t)
$$

Integrate this inequality with respect to $\mu$ and observe that $A \int_{E}\left(f_{+}^{\prime} \circ X\right) d \mu-\int_{E}\left(f_{+}^{\prime} \circ X\right) X d \mu=0$; (4) is immediate.

To complete the proof of Theorem 2, we consider the case of equality in (4). If $X$ is constant $\mu$ a.e., then (4) is an obvious equality. We now show that if we require $f$ to be strictly convex, then for equality to hold in (4), it is also necessary that $X$ be constant $\mu$ a.e.

Since $f$ is increasing and convexity is strict on ( $a, b$ ), have $f_{+}^{\prime}>0$, and $Z: E \rightarrow R$ as defined by the equation

$$
\begin{equation*}
\int_{E} f \circ X d \mu=(f \circ X)(t)+\left(f_{+}^{\prime} \circ X\right)(t)(Z(t)-X(t)), \tag{7}
\end{equation*}
$$

is measurable on $E$.
From (3) and (4), and the continuity and strict monotonicity of $f$, we see that there is a unique $x_{0} \in(a, b)$ such that $f\left(x_{0}\right)=\int_{E} f \circ X d \mu$. We rewrite (7) as

$$
\begin{equation*}
Z(t)-x_{0}=\frac{f\left(x_{0}\right)-\left((f \circ X)(t)+\left(x_{0}-X(t)\right)\left(f_{+}^{\prime} \circ X\right)(t)\right)}{\left(f_{+}^{\prime} \circ X\right)(t)} . \tag{8}
\end{equation*}
$$

From (6), and the fact that $f_{+}^{\prime}>0$, we see that for all $t \in E, X(t) \neq x_{0}$ if and only if $Z(t)>x_{0}$, while $X(t)=x_{0}$ if and only if $Z(t)=x_{0}$.

Integrate both sides of (7) with respect to $\mu$ and divide by $\int_{E} f_{+}^{\prime} \circ X d \mu$. We find

$$
\begin{equation*}
A=\frac{\int_{E}\left(f_{+}^{\prime} \circ X\right) Z d \mu}{\int_{E}\left(f_{+}^{\prime} \circ X\right) d \mu}, \tag{9}
\end{equation*}
$$

where $A$ is as defined above. If $X$ is not constant $\mu$ a.e., then $E_{1}=\left\{t \in E: X(t) \neq x_{0}\right\}=\left\{t \in E: Z(t)>x_{0}\right\}$ has positive measure. Since $E \backslash E_{1}=\left\{t \in E: Z(t)=x_{0}\right\}$, we see from (9) that $A>x_{0}$, and so $\int_{E} f \circ X d \mu=f\left(x_{0}\right)<f(A)$; thus the inequality in (4) is strict and the proof of Theorem 2 is complete.

The following example illustrates one of the possibilities for equality in (4) (more specifically, in (2)) when convexity is not strict. (A sketch will make the construction clear.)

Suppose $f_{i}(x)=m_{i} x+b_{i}, i=1,2$ and $x \in R$, where $0<m_{1}<m_{2}$. Define $f=\operatorname{Sup}\left(f_{1}, f_{2}\right)$ and let $\bar{x}, \bar{y}$ be defined by $f_{1}(\bar{x})=f_{2}(\bar{x})=\bar{y}$. Suppose $x_{1}$ and $x_{2}$ are fixed points such that $x_{1}<\bar{x}<x_{2} . \bar{y}$ is in the open interval $\left(f\left(x_{1}\right), f\left(x_{2}\right)\right)$ and so there exist $\alpha_{1}, \alpha_{2}>0$ such that $\alpha_{1}+\alpha_{2}=1$ and $\alpha_{1} f\left(x_{1}\right)+\alpha_{2} f\left(x_{2}\right)=\bar{y}$. An easy calculation shows that

$$
\bar{x}=\frac{\alpha_{1} f^{\prime}\left(x_{1}\right) x_{1}+\alpha_{2} f^{\prime}\left(x_{2}\right) x_{2}}{\alpha_{1} f^{\prime}\left(x_{1}\right)+\alpha_{2} f^{\prime}\left(x_{2}\right)} .
$$

$\left(f^{\prime}\left(x_{1}\right)=m_{1}\right.$ and $f^{\prime}\left(x_{2}\right)=m_{2}$.) Thus

$$
\alpha_{1} f\left(x_{1}\right)+\alpha_{2} f\left(x_{2}\right)=\bar{y}=f(\bar{x})=f\left(\frac{\alpha_{1} . f^{\prime}\left(x_{1}\right) x_{1}+\alpha_{2} f^{\prime}\left(x_{2}\right) x_{2}}{\alpha_{1} f^{\prime}\left(x_{1}\right)+\alpha_{2} f^{\prime}\left(x_{2}\right)}\right) .
$$

## 3. Applications

In this section we give two applications, one of inequality (4) (which yields a familiar result), and one of inequality (2) (which might be new). Since all integrals are over $E$, we will drop " $E$ " from the integral sign.
$1^{\circ}$. Define $f: R \rightarrow R$ by $f(x)=0$ for $x<0$ and $f(x)=x^{p}, p>1$, for $x \geqslant 0$. If $X: E \rightarrow R$ is non-negative, bounded, and measurable, and not a null function, then from (4)

$$
\begin{equation*}
\int X^{p} d \mu \leqslant\left(\frac{\int X^{p} d \mu}{\int X^{p-1} d \mu}\right)^{p} \tag{10}
\end{equation*}
$$

after rearranging we obtain

$$
\begin{equation*}
\left(\int X^{p-1} d \mu\right)^{1 /(p-1)} \leqslant\left(\int X^{p} d \mu\right)^{1 / p} \tag{11}
\end{equation*}
$$

By familiar approximation techniques of real analysis, notably the Lebesgue monotone convergence theorem, (11) can easily be shown to hold for all non-negative extended real valued measurable $X$. With $q>0$, replace $X$ by $X^{q}$ and raise both sides of (11) to the $1 / q$. If we now set $q=t-s$ and $p=t /(t-s)$, where $0<s<t$, we recover the well known result

$$
\begin{equation*}
\left(\int X^{s} d \mu\right)^{1 / s} \leqslant\left(\int X^{t} d \mu\right)^{1 / t}, \quad \text { if } \quad 0<s<t \tag{12}
\end{equation*}
$$

It is also well known that (12) holds if $s<t<0$; this too can be obtained from (11) by first replacing $X$ by $1 / X$, and then proceeding in a similar fashion.
$2^{\circ}$. Suppose $g:\{z:|z|<R\} \rightarrow C$ is analytic. Let $M(r)=$ $\operatorname{Max}\{|g(z)|:|z|=r\}$ for $0<r<R$. From the maximum modulus theorem and the Hadamard three circles theorem we are able to conclude that $\log M(r)$ is a convex increasing function of $\log r$. That is, the composite function $\log \circ M \circ \exp$ is convex and increasing on ( $0, R$ ). Since log and exp have strictly positive derivatives, we conclude that the one-sided derivatives $M_{+}^{\prime}$ and $M_{-}^{\prime}$ exist and satisfy $M_{-}^{\prime} \leqslant M_{+}^{\prime}$ and that they are equal except for at most countably many $r \in(0, R)$.

Take any $r_{1}, r_{2} \in(0, R)$ and $\alpha_{1}, \alpha_{2} \geqslant 0$ such that $\alpha_{1}+\alpha_{2}=1$. Hadamard's inequality states

$$
\begin{equation*}
M\left(r_{1}^{\alpha_{1}} r_{2}^{\alpha_{2}}\right) \leqslant\left(M\left(r_{1}\right)\right)^{\alpha_{1}}\left(M\left(r_{2}\right)\right)^{\alpha_{2}} \tag{13}
\end{equation*}
$$

the companion inequality, which will be proved below, states

$$
\begin{equation*}
\left(M\left(r_{1}\right)\right)^{\alpha_{1}}\left(M\left(r_{2}\right)\right)^{\alpha_{2}} \leqslant M\left(r_{1}^{\beta_{1}} r_{2}^{\beta_{2}}\right), \tag{14}
\end{equation*}
$$

where

$$
\beta_{i}=\frac{\alpha_{i} M_{+}^{\prime}\left(r_{i}\right) r_{i} / M\left(r_{i}\right)}{\alpha_{1} M_{+}^{\prime}\left(r_{1}\right) r_{1} / M\left(r_{1}\right)+\alpha_{2} M_{+}^{\prime}\left(r_{2}\right) r_{2} / M\left(r_{2}\right)}, \quad i=1,2 .
$$

We begin the proof by setting $u_{1}=\log r_{1}, u_{2}=\log r_{2}$, and with $f:(-\infty$, $\log R) \rightarrow R$ defined by $f(u)=\log M\left(e^{u}\right)$, we apply inequality (2):

$$
\begin{align*}
& \alpha_{1} \log M\left(e^{u_{1}}\right)+\alpha_{2} \log M\left(e^{u_{2}}\right)  \tag{16}\\
& \quad \leqslant \log M\left(\exp \left\{\frac{\alpha_{1}\left[M_{+}^{\prime}\left(e^{u_{1}}\right) / M\left(e^{u_{1}}\right)\right] e^{u_{1}} u_{1}+\alpha_{2}\left[M_{+}^{\prime}\left(e^{u_{2}}\right) / M\left(e^{u_{2}}\right)\right] e^{u_{2}} u_{2}}{\alpha_{1}\left[M_{+}^{\prime}\left(e^{u_{1}}\right) / M\left(e^{u_{1}}\right)\right] e^{u_{1}}+\alpha_{2}\left[M_{+}^{\prime}\left(e^{u_{2}}\right) / M\left(e^{u_{2}}\right)\right] e^{u_{2}}}\right\}\right)
\end{align*}
$$

If we now apply exp to both sides of (16) and replace $u_{1}$ and $u_{2}$ by $\log r_{1}$ and $\log r_{2}$ respectively, we find

$$
\begin{aligned}
& \left(M\left(r_{1}\right)\right)^{\alpha_{1}}\left(M\left(r_{2}\right)\right)^{\alpha_{2}} \\
& \quad \leqslant M\left(\exp \left\{\frac{\alpha_{1}\left[M_{+}^{\prime}\left(r_{1}\right) / M\left(r_{1}\right)\right] r_{1} \log r_{1}+\alpha_{2}\left[M_{+}^{\prime}\left(r_{2}\right) / M\left(r_{2}\right)\right] r_{2} \log r_{2}}{\alpha_{1}\left[M_{+}^{\prime}\left(r_{1}\right) / M\left(r_{1}\right)\right] r_{1}+\alpha_{2}\left[M_{+}^{\prime}\left(r_{2}\right) / M\left(r_{2}\right)\right] r_{2}}\right\}\right) \\
& \\
& \quad=M\left(\exp \left(\beta_{1} \log r_{1}+\beta_{2} \log r_{2}\right)\right) \\
& \quad=M\left(r_{1}^{B_{1}} r_{2}^{\beta_{2}}\right), \text { and the proof is complete. }
\end{aligned}
$$

## 4. Concluding Remarks

Hardy, Littlewood, and Polyà's "Inequalities," is still an excellent source; all our "well known" results are to be found there.

For a discussion of the differentiability of $M(r)$, Otto Blumenthal, "Über ganze transzendente Funktionen," Jahresbericht d. Deutschen Mathem.Vereinigung, XVI, Heft 2, 1905, seems to be the best we could find. $M_{+}^{\prime}(r)>M_{-}^{\prime}(r)$ can occur. We remark that in (14), any $M_{+}^{\prime}$ may be replaced by $M_{-}^{\prime}$ (or any value inbetween) and the inequality remains valid. (In this connection, compare the remarks following the statement of Theorem 2.)

Besides the elementary convex monotone functions (e.g., $f(x)=e^{x}$, $\left.f(x)=x\left(\log _{+} x\right)^{k}, k \geqslant 1\right)$ which often provide useful companion inequalities, there are other functions from analytic function theory whose companion inequalities may be worth investigating. For example, the Nevanlinna
characteristic $T(r, f)$ of a meromorphic $f$ (see Einar Hille, "Analytic Function Theory," Vol. II) is known to be an increasing convex function of $\log r$. Its differentiability properties were recently investigated by D. W. Townsend (Abstracts A. M. S. Vol. 1, No. 1, January 1980, Abstract 773-30-12).

