# KUBO-ANDO THEORY FOR CONVEX FUNCTIONAL MEANS 

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#### Abstract

Inspired by the geometric mean due to Atteia and Raïssouli, we discuss a general theory of convex functional means on a Hilbert space like the Kubo-Ando theory of operator means. Though our construction is based on the integral representation in Kubo-Ando theory, it is an exact extension not only for operator means but also for Atteia-Raïssouli's ones. We give an example where our geometric mean can be defined even if their geometric one cannot. We show that our convex functional means satisfy monotonicity, semi-continuity, homogeneity, subadditivity, joint concavity, transformer inequality and normalization. One of outstanding properties of these means appeares in those for constant functions, which suggests us to weights for operator means.


## 1. Introduction.

Since we met Ando's lecture note [2], we have been studying operator means, which is now known as the Kubo-Ando theory. For positive operators on a (complex) Hilbert space $H$, the theory of operator means is established axiomatically by Kubo and Ando [11]: An (operator) connection $m$ is a binary operation on positive operators satisfying the following axioms:

> monotonicity: $\quad A_{1} \leq A_{2}$ and $B_{1} \leq B_{2}$ imply $A_{1} m B_{1} \leq A_{2} m B_{2}$.
> semi-continuity: $\quad A_{n} \downarrow A$ and $B_{n} \downarrow B$ imply $A_{n} m B_{n} \downarrow m B$.
> transformer inequality: $\quad T^{*}(A m B) T \leq\left(T^{*} A T\right) m\left(T^{*} B T\right)$.

An operator mean is a connection $m$ satisfying
normalization: $\quad A m A=A$.
It is easy to show the transformer equality if $T$ is invertible. In particular, we have:
homogeneity: $\alpha(A m B)=(\alpha A) m(\alpha B) \quad$ for every positive number $\alpha$.
For an operator mean $m$, the corresponding numerical function $f_{m}(x)=1 m x$ is operator monotone:

$$
0 \leq A \leq B \quad \text { implies } \quad f_{m}(A) \leq f_{m}(B)
$$

This correspondence $m \mapsto f_{m}$ is bijective. In fact, if $f$ is a continuous nonnegative opeartor monotone functional on $[0, \infty)$ with $f(1)=1$, then a binary operation $m$ defined by

$$
A m B=A^{1 / 2} f\left(A^{-1 / 2} B A^{-1 / 2}\right) A^{1 / 2}
$$

for positive invertible operators $A$ and $B$ induces an operator mean $A m B$. As in [7], the operator meams bijectively correspond to the operator concave functionals on $[0,1]$ with $F(1 / 2)=1 / 2$ by

$$
F(x)=(1-x) m x
$$

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which shows another construction of operator means by Izumino's method [8]. This fact is also a bridge between means for operators and ones for positive forms by Pusz and Woronowicz $[14,15]$, and two theories are essentially equivalent.

Recently, Atteia and Raïssouli introduced the geometric convex functional mean in [3] which is characterized as the arithmetico-harmonic mean like the iteration method in [6]. As in [10] for example, the notion 'Legendre-Fenchel conjugate' $f$ ' of a convex functional $f$ on a real vector space

$$
f^{*}\left(y^{*}\right)=\sup _{x}\langle x, y\rangle-f(x)
$$

is considered as that of 'inverse' in some sense. So they used essentially the harmonic mean $\tau_{h}$ defined by

$$
f \tau_{h} g=\left(\frac{f^{*}+g^{*}}{2}\right)^{*}
$$

(In [16], they redefine these means on a complex Hilbert space.) Inspired by their convex functional means, we introduce a class of convex functional means $\sigma_{m}$ corresponding to the operator means $m$ and show similar properties of $\sigma_{m}$. Conversely we may define a general class $\Sigma_{0}$ of convex functional means axiomatically satisfying these common properties for $\sigma_{m}$ like the Kubo-Ando theory [11]. Finally we see what subclass $\Sigma$ corresponds to that of operator means.

## 2. The class $\Gamma$.

For our discussion, we summerize properties of a class of convex functionals. Let $f$ be a lower-bounded convex functional on a (complex) Hilbert space $H$ and

$$
\operatorname{dom} f=\{x \in H \mid f(x)<\infty\}
$$

the domain of $f$. If dom $f=H$, then $f$ is called finite. Throughout this note, we assume that $f$ is proper, i.e., dom $f$ is not empty. and $f$ is lower semi-continuous, i.e., the epigraph

$$
\text { epi } f=\{(x, \alpha) \in H \times \mathbb{R} \mid f(x) \leq \alpha\}
$$

of $f$ is closed. Then let $\Gamma=\Gamma(H)$ be the proper lower-bounded lower semi-continuous convex functionals on $H$. An indicator $\mathbf{1}_{C}$ for a closed convex subset $C$ of $H$ is defined by $\mathbf{1}_{C}(x)=0$ if $x \in C$ and $\mathbf{1}_{C}(x)=\infty$ otherwise and it is a simple example in $\Gamma$. A typical and important example in $\Gamma$ is $f_{A}$ for a bounded linear positive operator $A$ :

$$
f_{A}(x)=\frac{1}{2}\langle A x, x\rangle .
$$

The functional $f_{A}$ is called quadratic in the sense that $f(\gamma x)=|\gamma|^{2} f(x)$ for all complex number $\gamma$. For a subspace $C$, the indicator $\mathbf{1}_{C}$ is also quadratic.

As the researchers on convex functionals have been discussing, the class $\Gamma$ is stable in certain algebraic and topological senses. First it is closed for an key operation called Legendre-Fenchel conjugate $f^{*}$. Here, to condider complex spaces, it is defined as

$$
f^{*}\left(y^{*}\right)=\sup _{x \in H} \operatorname{Re}\langle x, y\rangle-f(x)
$$

where we write $y^{*}$ if $y$ is considered as a functional on $H$. Then $f^{*}$ is also a lower semicontinuous lower-bounded convex functional and $f^{* *}=\left(f^{*}\right)^{*}=f$. Note that $f^{*}$ is lower semi-continuous even if $f$ is not. $f^{* *}$ is often called the closure of $f$, denote by clf, since epi $\left(f^{* *}\right)$ coincides with the closure of epi $(f)$.

Moreover, note that the conjugate operation preserves the quadraticity and if it is induced by a positive invertible operator, then the conjugate is by its inverse, which is confirmed in [16]:

Lemma 2.1. If $f$ is quadratic, then so is $f^{*}$ and $f_{A}^{*}=f_{A^{-1}}$.
Proof. Here we show the former statement. Suppose $f(\gamma x)=|\gamma|^{2} f(x)$. Then

$$
f^{*}\left(\gamma y^{*}\right)=\sup _{x} \operatorname{Re}\langle x, \bar{\gamma} y\rangle-f(x)=|\gamma|^{2} \sup _{x} \operatorname{Re}\left\langle\frac{1}{\bar{\gamma}} x, y\right\rangle-f\left(\frac{1}{\bar{\gamma}} x\right)=|\gamma|^{2} f^{*}\left(y^{*}\right)
$$

Thus this conjugate operation * is considered as the inverse in the sense that $f_{A}^{*}=f_{A^{-1}}$. The following properties are easily obtained: For $f, g \in \Gamma$ and every positive number $\alpha>0$, we have
$\left(1^{*}\right) \quad f \leq g$ implies $\quad f^{*} \geq g^{*}$.
$\left(2^{*}\right) \quad(f \pm \alpha)^{*}=f^{*} \mp \alpha$ 。
$\left(3^{*}\right) \quad(\alpha f)^{*}\left(\alpha y^{*}\right)=\alpha f^{*}\left(y^{*}\right)$.
For a finite dimensional case, the epi-convergence, which is characterized by

$$
\text { epi }\left(\lim _{n \rightarrow \infty} f_{n}\right)=\lim _{n \rightarrow \infty}\left(\operatorname{epi} f_{n}\right)
$$

for a product topology in $H \times \mathbb{R}$, has been often discussed in this class $\Gamma$ as in a standard text like [17] or [10]. Moreover the Legendre-Fenchel conjugate preserves this convergence. To extend this property to an infinite dimensional case, Mosco [13] introduced the convergence $\mathrm{M}-\lim _{n \rightarrow \infty} f_{n}$, which is now called Mosco convergence, if the following conditions are satisfied:
(i) for each $x \in H$, there exists $x_{n} \in H$ with $\underset{n \rightarrow \infty}{\operatorname{s-lim}} x_{n}=x$ and $f(x)=\lim _{n \rightarrow \infty} f_{n}\left(x_{n}\right)$,
(ii) $f(x) \leq \liminf _{n \rightarrow \infty} f_{n}\left(x_{n}\right)$ for $\underset{n \rightarrow \infty}{\mathrm{w}-\lim _{n \rightarrow \infty} x_{n}=x}$.

Then it is also shown that

$$
\underset{n \rightarrow \infty}{\mathrm{M}-\lim _{n}} f_{n}=f \quad \text { if and only if } \quad \underset{n \rightarrow \infty}{\mathrm{M}-\lim _{n}} f_{n}^{*}=f^{*}
$$

Considering such stability for $\Gamma$, Atteia and Raissouli [3] introduced convex functional means as the geometric mean is expressed by the limit in monotone convergence as the arithmetico-harmonic one.

## 3. Parallel addition.

The parallel addition for operators is introduced by Anderson-Duffin [1] and FillmoreWilliams [5]. For invertible operators, it is represented by

$$
A: B=\left(A^{-1}+B^{-1}\right)^{-1}
$$

and in general it is characterized by the formula:

$$
\begin{equation*}
\langle A: B x, x\rangle=\inf _{y+z=x}\langle A y, y\rangle+\langle B z, z\rangle \tag{*}
\end{equation*}
$$

Typical extremal operator means are the arithmetic one $A \nabla B$ and the harmonic one $A!B=2(A: B)$. By Löwner's theory [12], Kubo and Ando [11] showed that the operator means correspond also bijectively onto the Radon probability measures on $[0, \infty]$ by the following integral representation: For an operator mean $m$, there exists the Radon probability measure $\mu_{m}$ on $[0, \infty]$ with

$$
A m B=a A+b B+\int_{(0, \infty)}(t A): B \frac{1+t}{t} d \mu_{m}(t)
$$

where $a=\mu_{m}(\{0\})$ and $b=\mu_{m}(\{\infty\})$. Thus, roughly speaking, an operator mean is considered as a convex combination of the arithmetic mean and the harmonic one, or equivalently the addition and the parallel addition.

The above formula (*) suggests us that we can make the similar discussion in convex functionals since the notion 'inf-convolution' $f * g$ is defined as

$$
f * g(x)=\inf _{y+z=x} f(y)+g(z) .
$$

Unfortunately $\Gamma$ is not closed under this operation. In fact,

$$
\operatorname{dom}(f * g)=\operatorname{dom}(f)+\operatorname{dom}(g)
$$

and consequently

$$
\mathbf{1}_{C_{1}} * \mathbf{1}_{C_{2}}=\mathbf{1}_{C_{1}+C_{2}}
$$

but it is known that, in an infinite space, $C_{1}+C_{2}$ is not always closed even if both $C_{1}$ and $C_{2}$ is closed. For example, for a bounded linear operator $A$ on a Hilbert space $H$ whose range is not closed, put $C_{1}=H \oplus\{0\}$ and $C_{2}=\{(x, A x) \mid x \in H\}$ the graph of $A$. Then both $C_{1}$ and $C_{2}$ are closed in $H \oplus H$ but $C_{1}+C_{2}$ is not, see [9]. Thus $f * g$ is not always lower semi-continuous. Moreover, even if $f * g$ is lower semi-continuous, we find it unsuitable for a parallel addition for convex functionals in spite of the formula $(*)$ since $2(f * f) \neq f$ and $2 f * g \neq f \tau_{h} g$ if convex functions are not quadratic. However the following result is known for a real case. Moreover we pay attention to the following useful method to prove inequality or equality for parallel additions. So we give a proof of it:
Lemma 3.1. If $f, g \in \Gamma$, then $\mathrm{cl} f * g \equiv(f * g)^{* *}=\left(f^{*}+g^{*}\right)^{*}$.
Proof. Since

$$
\begin{aligned}
\left(f^{*}+g^{*}\right)\left(y^{*}\right) & =\sup _{v, w} \operatorname{Re}\langle v+w, y\rangle-(f(v)+g(w)) \\
& \leq \sup _{v, w} \operatorname{Re}\langle v+w, y\rangle-(f * g)(v+w)=(f * g)^{*}\left(y^{*}\right)
\end{aligned}
$$

we have $\left(f^{*}+g^{*}\right)^{*} \geq f * g^{* *}$ by $\left(1^{*}\right)$. Conversely

$$
\begin{aligned}
f * g(x) & =\inf _{y+z=x} f^{* *}(y)+g^{* *}(z) \\
& =\inf _{y+z=x}\left(\sup _{v} \operatorname{Re}\langle y, v\rangle-f^{*}\left(v^{*}\right)\right)+\left(\sup _{w} \operatorname{Re}\langle z, w\rangle-g^{*}\left(w^{*}\right)\right) \\
& \geq \inf _{y+z=x}\left(\sup _{v} \operatorname{Re}\langle y+z, v\rangle-\left(f^{*}+g^{*}\right)\left(v^{*}\right)\right) \\
& =\sup _{v} \operatorname{Re}\langle x, v\rangle-\left(f^{*}+g^{*}\right)\left(v^{*}\right)=\left(f^{*}+g^{*}\right)^{*}(x) .
\end{aligned}
$$

Taking the conjugation twice, we have $(f * g)^{* *} \geq\left(f^{*}+g^{*}\right)^{* * *}=\left(f^{*}+g^{*}\right)^{*}$. Thus $(f * g)^{* *}=$ $\left(f^{*}+g^{*}\right)^{*}$.

Considering these and $\left(3^{*}\right)$, we define the parallel addition for $f$ and $g$ by

$$
(f: g)(x)=\frac{1}{4}_{2}\left(\left(f^{*}+g^{*}\right)^{*}\right)(x) \equiv \frac{1}{4}\left(f^{*}+g^{*}\right)^{*}(2 x)
$$

Note that a functional ${ }_{T} f$ defined by ${ }_{T} f(x) \equiv f(T x)$ for a bounded linear operator $T$ also belongs to $\Gamma$ for $f \in \Gamma$. By $\left(3^{*}\right)$, this implies the following formulae:

$$
(f: g)(x)=\frac{1}{4}(f * g)^{* *}(2 x)=\frac{1}{2}\left(\frac{f^{*}+g^{*}}{2}\right)^{*}(x)=\frac{1}{2}\left(f \tau_{h} g\right)(x)
$$

Contrary to such differences, all of them are extensions for the parallel addition for operators:
Theorem 3.2. If $A$ and $B$ are positive operators, then

$$
f_{A}: f_{B}=f_{A} * f_{B}=\left(f_{A}^{*}+f_{B}^{*}\right)^{*}=f_{A: B}
$$

Proof. By the formula (*), we have

$$
\begin{aligned}
f_{A} * f_{B}(x) & =\inf _{y+z=x} f_{A}(y)+f_{B}(z)=\frac{1}{2} \inf _{y+z=x}(\langle A y, y\rangle+\langle B z, z\rangle) \\
& =\frac{1}{2}\langle(A: B) x, x\rangle=f_{A: B}(x),
\end{aligned}
$$

so that $f_{A} * f_{B}$ is lower semi-continuous and hence $\left(f_{A}^{*}+f_{B}^{*}\right)^{*}=f_{A} * f_{B}=f_{A: B}$ by Lemma 3.1. In addition,

$$
f_{A}: f_{B}(x)=\frac{1}{4}\left(f_{A}^{*}+f_{B}^{*}\right)^{*}(2 x)=\frac{1}{4}\langle(A: B) 2 x, 2 x\rangle=\langle(A: B) x, x\rangle=f_{A: B}(x) .
$$

Since the harmonic mean should be twice the parallel addition, we claim $2(f: f)=f$. In general, $2(f * f)=f$ does not always holds, but $2(f: f)=f$ does, which is the reason we need the above definition:
Lemma 3.3. $f: f=\frac{1}{2} f \quad$ for all $f \in \Gamma$.
Proof. By $\left(3^{*}\right)$, we have $2(f: f)(x)=\frac{1}{2}\left(2 f^{*}\right)^{*}(2 x)=\frac{1}{2} \times 2 f^{* *}(x)=f(x)$.
The following estimation, which shows that the parallel addition is an operation in $\Gamma$, is obtained immediately by $4(f: g)={ }_{2}(f * g)^{* *} \leq_{2}(f * g)$ :
Theorem 3.4. If $t \in \operatorname{dom} f$ and $s \in \operatorname{dom} g$ for $f, g \in \Gamma$, then

$$
4(f: g)(x) \leq f(2 x-s)+g(s) \quad \text { and } \quad 4(f: g)(x) \leq f(t)+g(2 x-t)
$$

In particular, $f($ resp, $g$ ) are normalized in the sense $f(0)=0$, then

$$
f: g(x) \leq \frac{1}{4} g(2 x) \quad\left(\text { resp., } \quad f: g(x) \leq \frac{1}{4} f(2 x) .\right)
$$

The last inequality is represented into a simple one if $g$ (resp., $f$ ) is quadratic in the sense that $f(\gamma x)=|\gamma|^{2} f(x)$ for every complex number $\gamma$ :

$$
f: g \leq g \quad(\text { resp. }, f: g \leq f)
$$

We are very interseted in quadratic functionals, so we discuss them in the next section.

## 4. Convex functional means via operator ones.

To consider general means, we show that the parallel addition has properties like operator means:

Lemma 4.1. For $f, g, h, k, f_{k}, g_{k} \in \Gamma$ and $\beta \in(0,1)$, the parallel addition satisfies
monotonicity: $\quad f \leq h$ and $g \leq k$ imply $f: g \leq h: k$.
semi-continuity: $\quad f_{n} \downarrow f$ and $g_{n} \downarrow g$ imply $f_{n}: g_{n} \downarrow f: g$.
subadditivity: $\quad(f+h):(g+k) \geq f: g+h: k$.
joint concavity: $(\beta f+(1-\beta) h):(\beta g+(1-\beta) k) \geq \beta(f: g)+(1-\beta)(h: k)$.
Proof. If $f \leq h$ and $g \leq k$, then $\left(1^{*}\right)$ shows

$$
4(f: g)(x)=\left(f^{*}+g^{*}\right)^{*}(2 x) \leq\left(h^{*}+k^{*}\right)^{*}(2 x)=4(h: k)(x)
$$

The semicontinuity is obtained by the fact that monotone convergence implies Mosco convergence, see [4]. The subadditivity of inf-convolution follows from

$$
\begin{aligned}
2((f+h) *(g+k))(x) & =\inf _{y+z=2 x}(f+h)(y)+(g+k)(z) \\
& \geq \inf _{y+z=2 x} f(y)+g(z)+\inf _{y+z=2 x} h(y)+k(z) \\
& =2(f * g)(x)+2(h * k)(x) .
\end{aligned}
$$

Taking double conjugate, we have

$$
4(f+h):(g+k) \geq 4(f: g)+4(h: k)
$$

Combining the homogeneity and the subadditivity, we have the joint concavity.
Since the product operation is lacking in the convex functionals, we use instead the operation

$$
{ }_{T} f(x)=f(T x)
$$

for $f \in \Gamma$ and a (bounded linear) operator $T$ on $H$. Then we show another inverse property of conjugate:
Lemma 4.2. $\left({ }_{T} f\right)^{*}=\left(T^{-1}\right)^{*}\left(f^{*}\right)$ for invertible $T$.
Proof. The required formula follows from
$\left({ }_{T} f\right)^{*}\left(y^{*}\right)=\sup _{x} \operatorname{Re}\langle x, y\rangle-f(T x)=\sup _{x} \operatorname{Re}\left\langle T x,\left(T^{-1}\right)^{*} y\right\rangle-f(T x)=f\left(\left(T^{-1}\right)^{*} y^{*}\right)$.
Now, considering ${ }_{T} f_{A}=f_{T^{*} A T}$, we can discuss the transformer inequality:
Lemma 4.3. For $f, g \in \Gamma$, the parallel addition satisfies
transformer inequality: $\quad T(f: g) \leq(T f):\left({ }_{T} g\right)$.
transformer equality: $\quad T(f: g)=\left({ }_{T} f\right):\left({ }_{T} g\right)$ if $T$ is invertible.
homogeneity: $\quad(\alpha f: \alpha g)=\alpha(f: g)$ for $\alpha>0$.
quadratic preserving: If $f$ and $g$ is quadratic, then so is $f: g$.
Proof. Since the range of $T$ is a subspace of $H$, we have

$$
\begin{aligned}
4_{T}(f: g)(x) & =4(f: g)(T x) \leq_{2}(f * g)(T x)=\inf _{v+w=2 T x} f(v)+g(w) \\
& \leq \inf _{y+z=2 x} f(T y)+g(T z)={ }_{2}\left(\left({ }_{T} f\right) *\left({ }_{T} g\right)\right)(x),
\end{aligned}
$$

which implies the required inequality by taking the conjugation twice. Since $x(f+g)=X$ $f+x g$, Lemma 4.2 implies

$$
\begin{aligned}
4_{T}(f: g)(x) & ={ }_{T}\left(\left(f^{*}+g^{*}\right)^{*}\right)(2 x)=\left({\left(T^{-1}\right)^{*}}\left(f^{*}+g^{*}\right)\right)^{*}(2 x) \\
& =\left(\left({ }_{T} f\right)^{*}+\left({ }_{T} g\right)^{*}\right)^{*}(2 x)=4\left(_{T} f\right):\left({ }_{T} g\right)(x) .
\end{aligned}
$$

The homogeneity follows from

$$
\begin{aligned}
4(\alpha f: \alpha g)(x) & =\left((\alpha f)^{*}+(\alpha g)^{*}\right)^{*}(2 x)=\left(\alpha_{1 / \alpha} f^{*}+\alpha_{1 / \alpha} g^{*}\right)^{*}(2 x) \\
& =\alpha\left(1 / \alpha f^{*}+_{1 / \alpha} g^{*}\right)^{*}(2 x / \alpha)=\alpha_{\alpha}\left(f^{*}+g^{*}\right)^{*}(2 x / \alpha) \\
& =\alpha\left(f^{*}+g^{*}\right)^{*}(2 x)=4 \alpha(f: g)(x) .
\end{aligned}
$$

and the quadraticity is preserved by

$$
|\gamma|^{2}(f: g)=\left(|\gamma|^{2} f\right):\left(|\gamma|^{2} g\right)={ }_{\gamma} f:_{\gamma} g=\gamma_{\gamma}(f: g)
$$

Now we can define a class of convex functional means $\sigma_{m}$ corresponding to that of operator means $m$ : First note that $f, g \in \Gamma$ does not always imply $f+g \in \Gamma$. In fact, if dom $f \cap \operatorname{dom} g=\emptyset$, then $f+g \equiv \infty$. So, we assume here

$$
\operatorname{dom} f \cap \operatorname{dom} g \neq \emptyset
$$

when we discuss means of functionals in $\Gamma$. Such a pair $(f, g)$ is denoted by $(f, g) \in \Gamma^{2}$. Let $m$ be an operator mean in the sense of Kubo and Ando. Considering its integral representation, we define the convex functional mean $\sigma_{m}$ associated with $m$ by

$$
\left(f \sigma_{m} g\right)(x)=a f(x)+b g(x)+\int_{(0, \infty)}\left(\frac{1+t}{2} f: t_{\frac{1+t}{2 t}} g\right)(x) \frac{4}{1+t} d \mu_{m}(t)
$$

If $f$ and $g$ are quadratic, then we easily have

$$
f \sigma_{m} g(x)=a f(x)+b g(x)+\int_{(0, \infty)}((t f): g)(x) \frac{1+t}{t} d \mu_{m}(t)
$$

In fact, since ${ }_{s} f(x)=f(s x)=s^{2} f(x)$ for quadratic $f$ and $s>0$, we have

$$
\left(\frac{1+t}{2} f: t_{\frac{1+t}{2 t}} g\right)(x) \frac{4}{1+t}=\left(\frac{(1+t)^{2}}{4} f: \frac{(1+t)^{2}}{4 t} g\right)(x) \frac{4}{1+t}=(t f: g) \frac{1+t}{t} .
$$

The above modification is needed for the normalization of the mean as we show in the below.

For the arithmetic mean $\sigma_{\nabla}$, the measure is decided by $a=b=1 / 2$ and $\mu_{\nabla}((0, \infty))=0$. For the harmonic mean $\sigma_{!}$, we have $\mu_{!}=\delta_{\{1\}}$, the Dirac measure. For the geometric mean $\sigma_{\sharp}$, the corresponding measure $\mu_{\sharp}$ is decided by

$$
d \mu_{\sharp}(t)=\frac{\sin \frac{\pi}{2}}{\pi \sqrt{t}(1+t)} d t
$$

Here we denote the geometric operator mean by $\sharp$.
By the above lemmas, we have fundamental properties similarly to operator means, which forms basic part of an extension of the Kubo-Ando theory:

Theorem 4.4. For $(f, g),(h, k),\left(f_{k}, g_{k}\right) \in \Gamma^{2}, \alpha>0$ and $\beta \in(0,1)$, a convex functional mean $\sigma_{m}$ for an operator mean $m$ is LBPL preserving and has the following properties:

$$
\begin{array}{ll}
\text { monotonicity: } & f \leq h \text { and } g \leq k \text { imply } f \sigma_{m} g \leq h \sigma_{m} k . \\
\text { semi-continuity: } & f_{n} \downarrow f \text { and } g_{n} \downarrow g \text { imply } f_{n} \sigma_{m} g_{n} \downarrow f \sigma_{m} g . \\
\text { homogeneity: } & (\alpha f) \sigma_{m}(\alpha g)=\alpha\left(f \sigma_{m} g\right) . \\
\text { subadditivity: } & (f+h) \sigma_{m}(g+k) \geq f \sigma_{m} g+h \sigma_{m} k . \\
\text { joint concavity: } & (\beta f+(1-\beta) h) \sigma_{m}(\beta g+(1-\beta) k) \geq \beta\left(f \sigma_{m} g\right)+(1-\beta)\left(h \sigma_{m} k\right) \text {. } \\
\text { transformer inequality: } \quad T\left(f \sigma_{m} g\right) \leq(T f) \sigma_{m}(T g) . \\
\text { normalization: } \quad f \sigma_{m} f=f . \\
\text { quadratic preserving: } \quad \text { If } f \text { and } g \text { is quadratic, then so is } f \sigma g \text {. }
\end{array}
$$

Proof. It suffices to show the normalization. By Lemma 4.3, we have

$$
\begin{aligned}
\frac{4}{1+t}\left(\frac{1+t}{2} f: t_{\frac{1+t}{2 t}} f\right) & (x)=\frac{1}{1+t}\left(\left(\frac{1+t}{2} f\right)^{*}+\left(t_{\frac{1+t}{2 t}} f\right)^{*}\right)^{*}(2 x) \\
& =\frac{1}{1+t}\left(\left(\frac{1+t}{2} f\right)^{*}+t\left(\frac{1+t}{2} f\right)^{*}\right)^{*}(2 x)=\frac{1}{1+t}\left((1+t)\left(\frac{1+t}{2} f\right)^{*}\right)^{*}(2 x) \\
& =\left(\left(\frac{1+t}{2} f\right)^{*}\right)^{*}\left(\frac{2}{1+t} x\right)=\left(\frac{2}{1+t} f^{*}\right)^{*}\left(\frac{2}{1+t} x\right) \\
& =\frac{1+t}{2}\left(f^{* *}\right)\left(\frac{2}{1+t} x\right)=f^{* *}(x)=f(x)
\end{aligned}
$$

Since $\mu_{m}$ is a probability measure, we have $f \sigma_{m} f=f$.
Considering constant functions $m \mathbf{I}$ where $\mathbf{I}(x) \equiv 1$, we have:
Corollary 4.5. If $(f, g) \in \Gamma^{2}$, then $f \sigma_{m} g$ is also lower bounded;

$$
f, g \geq m \quad \text { implies } \quad f \sigma_{m} g \geq m
$$

and belongs to $\Gamma$. In particular, $f, g \geq 0$ implies $f \sigma_{m} g \geq 0$.
Recalling that

$$
A m B=A^{1 / 2} f\left(A^{-1 / 2} B A^{-1 / 2}\right) A^{1 / 2}
$$

for operator mean $m$ for invertible operators, we reconstruct convex functional ones:

Corollary 4.6. If $A$ is invertible, then

$$
f_{A m B}=A_{A^{1 / 2}} f_{I m\left(A^{-1 / 2} B A^{-1 / 2}\right)}=\left(A^{1 / 2} f_{I}\right) \sigma_{m}\left(A^{1 / 2} f_{A^{-1 / 2} B A^{-1 / 2}}\right)
$$

Here we discuss constant functions $\mathbf{I}$ including an extension of Corollary 4.5. Before considering means for constant functions generally, we give a case of the parallel addition:

Lemma 4.7. $(f+t \mathbf{I}):(g+t \mathbf{I})=f: g+t \mathbf{I}, \quad t \mathbf{I}: s \mathbf{I}=\frac{t+s}{4} \mathbf{I} \quad$ for $t, s \in \mathbb{R}$ and $f \in \Gamma$.
Proof. Observing the proof of the subadditivity, we easily have the former. By $(t \mathbf{I})^{*}=$ $\mathbf{1}_{\{0\}}-t \mathbf{I}$, we have

$$
4(t \mathbf{I}: s \mathbf{I})=\left((t \mathbf{I})^{*}+(s \mathbf{I})^{*}\right)^{*}=\left(\mathbf{1}_{\{0\}}-(t+s) \mathbf{I}\right)^{*}=(t+s) \mathbf{I}
$$

Thereby we have the additivity, or translation invariance of constants:
Theorem 4.8. If $c$ is a real number, $m$ is an operator mean and $(f, g) \in \Gamma^{2}$, then

$$
(f+c \mathbf{I}) \sigma_{m}(g+c \mathbf{I})=f \sigma_{m} g+c \mathbf{I}
$$

Proof. We obtain the result by

$$
(f+c \mathbf{I}) \sigma_{m}(g+c \mathbf{I})(x)
$$

$$
=a(f+c \mathbf{I})(x)+b(g+c \mathbf{I})(x)+\int_{(0, \infty)}\left(\frac{1+t}{2}(f+c \mathbf{I}): t_{\frac{1+t}{2 t}}(g+c \mathbf{I})\right)(x) \frac{4}{1+t} d \mu_{m}(t)
$$

$$
=a f(x)+b g(x)+(a+b) c \int_{(0, \infty)}\left(\left(_{\frac{1+t}{2}} f+c \mathbf{I}\right):\left(t_{\frac{1+t}{2 t}} g+t c \mathbf{I}\right)\right)(x) \frac{4}{1+t} d \mu_{m}(t)
$$

$$
=a f(x)+b g(x)+(a+b) c \int_{(0, \infty)}\left(\frac{1+t}{2} f: t_{\frac{1+t}{2 t}} g+\frac{c+t c}{4} \mathbf{I}\right)(x) \frac{4}{1+t} d \mu_{m}(t)
$$

$$
=\left(f \sigma_{m} g\right)(x)+(a+b) c+c \int_{(0, \infty)} d \mu_{m}(t)=\left(f \sigma_{m} g\right)(x)+c
$$

This theorem shows that convex functional means are reduced to the case for positive functionals by translation. Recall that the order $m \leq n$ as operator means is defined by $A m B \leq A n B$ for all positive operators $A$ and $B$. By the definition via integral representation, we immediately have a map $m \mapsto \sigma_{m}$ is order-preserving for positive functionals, and hence for all functionals in $\Gamma$ :

Corollary 4.9. If $m \leq n$ as operator means, then $f \sigma_{m} g \leq f \sigma_{n} g$ for all $(f, g) \in \Gamma^{2}$.
Now we have a formula for means of constant functions:
Theorem 4.10. If $r$ and $s$ are real numbers and $m$ is an operator mean, then

$$
(r \mathbf{I}) \sigma_{m}(s \mathbf{I})=\left(\int_{[0, \infty]} \frac{1}{1+t} d \mu_{m}(t)\right) r \mathbf{I}+\left(\int_{[0, \infty]} \frac{t}{1+t} d \mu_{m}(t)\right) s \mathbf{I}
$$

In particular, if $m$ is symmetric, then $(r \mathbf{I}) \sigma_{m}(s \mathbf{I})=\frac{r+s}{2} \mathbf{I}$.

Proof. By Lemma 4.7, we have

$$
\begin{aligned}
(r \mathbf{I}) \sigma_{m}(s \mathbf{I})(x) & =a r \mathbf{I}(x)+b s \mathbf{I}(x)+\int_{(0, \infty)}\left(\frac{1+t}{2}(r \mathbf{I}): t_{\frac{1+t}{2 t}}(s \mathbf{I})\right)(x) \frac{4}{1+t} d \mu_{m}(t) \\
& =a r+b s+\int_{(0, \infty)}(r \mathbf{I}):(t s \mathbf{I})(x) \frac{4}{1+t} d \mu_{m}(t) \\
& =a r+b s+\int_{(0, \infty)} \frac{r+t s}{1+t} d \mu_{m}(t) \\
& =\left(\int_{[0, \infty]} \frac{1}{1+t} d \mu_{m}(t)\right) r+\left(\int_{[0, \infty]} \frac{t}{1+t} d \mu_{m}(t)\right) s .
\end{aligned}
$$

By the above, we can define the weight $\mathbf{W}\left(\sigma_{m}\right)$ by

$$
\mathbf{W}\left(\sigma_{m}\right)=\int_{[0, \infty]} \frac{t}{1+t} d \mu_{m}(t)
$$

In fact, we can confirm that the nonsymmetric degree for the weighted arithmetic, geometric or harmonic mean is equal to its weight respectively.

## 5. Quadratic functionals.

To observe the difference between operator means and convex functional ones, we confirm basic properties for quadratic functionals $f$ in $\Gamma$ and discuss when a convex functional mean is associated by some operator mean. Note that $f(0)=0, f(x)=f(-x)$ and hence $f$ is nonnegative by

$$
f(x)=\frac{f(x)+f(-x)}{2} \geq f(0)=0
$$

Now we characterize its local boundedness, which is rather 'boundedness' for functionals:
Lemma 5.1. Every quadratic functional $f \in \Gamma$ is locally bounded in the sense $\sup _{\|x\|=1} f(x)<$ $\infty$ if and only if $f$ is continuous at 0 . In this case, $f$ is finite.

Proof. Suppose $f$ is not locally bounded. Then there exist a sequence of unit vectors $x_{n}$ with $f\left(x_{n}\right)>n^{3}$. Putting $y_{n}=\frac{1}{n} x_{n}$, we have $y_{n} \rightarrow 0$ by $\left\|y_{n}\right\|=1 / n$ while $f\left(y_{n}\right) \rightarrow \infty$ by

$$
f\left(y_{n}\right)=\frac{1}{n^{2}} f\left(x_{n}\right)>n
$$

Conversely suppose there exists $M>0$ with $\sup _{x \neq 0} f(x /\|x\|) \leq M$. Then

$$
0 \leq f(x)=\|x\|^{2} f\left(\frac{x}{\|x\|}\right) \leq M\|x\|^{2}
$$

and hence $x_{n} \rightarrow 0$ implies $f\left(x_{n}\right) \rightarrow 0=f(0)$.
Here we give a proper example in an infinite space which is not locally bounded. Let $H=\ell^{2}$ with an orthonormal basis $\left\{e_{k}\right\}$. Putting

$$
f\left(\sum_{n=1}^{\infty} \gamma_{n} e_{n}\right)=\sum_{n=1}^{\infty} n^{3}\left|\gamma_{n}\right|^{2}
$$

we have $f$ is not locally bounded and hence not continuous at 0 . In fact, $x_{n}=(1 / n) e_{n} \rightarrow 0$ while $f\left(x_{n}\right) \rightarrow \infty$.

Note that the parallel addition is quadratic preserving:
Lemma 5.2. If $f$ and $g$ are quadratic functionals in $\Gamma$, then so is $f: g$.

Proof. Since $f$ and $g$ are quadratic, then

$$
{ }_{2}(f * g)(\gamma x)=\inf _{y+z=2 x} f(\gamma y)+g(\gamma z)=|\gamma|^{2} \inf _{y+z=2 x} f(y)+g(z)=|\gamma|^{2}{ }_{2}(f * g)(x)
$$

Taking the conjugation twice, we have $4(f: g)(\gamma x)=4|\gamma|^{2}(f: g)(x)$.
A quadratic functional $f \in \Gamma$ is called a $P L$ functional if $f$ satisfies

$$
\text { parallelogram law: } \quad f(x+y)+f(x-y)=2(f(x)+f(y))
$$

For example, consider $f(x)=\|x\|_{p}^{2}$ for the $p$-(semi)norm on $H$. Then $f$ is PL functional if and only if $p=2$.

It is easy to see the conjugate preserves PL functionals:
Lemma 5.3. If $f$ is a $P L$ functional, then so is $f^{*}$.
Proof. For a PL functional $f \in \Gamma$, we have

$$
\begin{aligned}
f^{*}(x+y) & +f^{*}(x-y)=\sup _{v, w} \operatorname{Re}\langle x+y, v\rangle-f(v)+\langle x-y, w\rangle-f(w) \\
& =\sup _{z_{1}, z_{2}} \operatorname{Re}\left\langle x+y, z_{1}+z_{2}\right\rangle+\left\langle x-y, z_{1}-z_{2}\right\rangle-\left(f\left(z_{1}+z_{2}\right)+f\left(z_{1}-z_{2}\right)\right) \\
& =2 \sup _{z_{1}, z_{2}} \operatorname{Re}\left\langle x, z_{1}\right\rangle+\left\langle y, z_{2}\right\rangle-f\left(z_{1}\right)-f\left(z_{2}\right)=2\left(f^{*}(x)+f^{*}(y)\right) .
\end{aligned}
$$

In general, locally bounded PL functionals have good properties:
Lemma 5.4. A locally bounded $P L$ functional $f \in \Gamma$ is finite, positive and continuous.
Proof. It suffices to show the continuity. The parallelogram law shows

$$
f(x \pm 2 y)+f(x)=2(f(x \pm y)+f(y)),
$$

Considering the difference of both sides, we have

$$
f(x+2 y)-f(x-2 y)=2(f(x+y)-f(x-y)) .
$$

Then, it follows inductively that

$$
f(x+y)-f(x-y)=\frac{1}{2}(f(x+2 y)-f(f-2 y))=\cdots=\frac{1}{2^{k}}\left(f\left(x+2^{k} y\right)-f\left(f-2^{k} y\right)\right) .
$$

For a fixed vector $x$, any small number $\varepsilon>0$ and the bound $M=\sup _{\|z\|=1} f(z)$, there exists $K>0$ with

$$
\frac{M}{2^{K}}(\|x\|+1)<\frac{\varepsilon}{2}
$$

For a sequence $y_{n} \rightarrow 0$, there exists a number $N$ with $\left\|y_{n}\right\|<1 / 2^{K}$ for all $n \geq N$. Then

$$
\begin{aligned}
\mid f\left(x+y_{n}\right) & \left.-f\left(x-y_{n}\right)\left|=\frac{1}{2^{K}}\right| f\left(x+2^{K} y_{n}\right)-f\left(x-2^{K} y_{n}\right) \right\rvert\, \\
& \leq \frac{1}{2^{K}}\left(\left|f\left(x+2^{K} y_{n}\right)\right|+\left|f\left(x-2^{K} y_{n}\right)\right|\right) \leq \frac{M}{2^{K}}\left(\left\|x+2^{K} y_{n}\right\|+\left\|x-2^{K} y_{n}\right\|\right) \\
& \leq \frac{2 M}{2^{K}}\left(\|x\|+2^{K}\left\|y_{n}\right\|\right)<\frac{2 M}{2^{K}}(\|x\|+1)<\varepsilon .
\end{aligned}
$$

Thus $f\left(x+y_{n}\right)-f\left(x-y_{n}\right) \rightarrow 0$ as $n \rightarrow \infty$. Moreover, by the continuity at 0 ,

$$
f\left(x+y_{n}\right)+f\left(x-y_{n}\right)=2\left(f(x)+f\left(y_{n}\right)\right) \rightarrow 2 f(x),
$$

and consequently $f\left(x+y_{n}\right) \rightarrow f(x)$, which shows the continuity.

As usual, we can identfy a continuous PL functional with a quadratic one for positive operators:

Lemma 5.5. If $f$ is a locally bounded PL functional in $\Gamma$, then it is the quadratic functional $f_{A}$ for some positive operator $A$ on $H$.

Proof. Note that $f$ is continuous. So, similarly to the von Neumann-Jordan theorem, we can define a continuous positive sesquilinear form $\Phi$ by the polarization identity

$$
4 \Phi(x, y)=f(x+y)-f(x-y)+i f(x+i y)-i f(x-i y)
$$

Thereby there exists a bounded linear positive operator $A$ as its Radon-Nikodym type derivative with $f=f_{A}$.

One of the outstanding properties of the parallel addition is $L B P L$ preserving in the sense that $f: g$ is a locally bounded PL functional in $\Gamma$ if $f$ and $g$ are so:
Lemma 5.6. The parallel addition are $L B P L$ preserving.
Proof. By Lemma 5.3, the parallel addition is also PL preserving. The local boundedness follows from Theorem 3.4 and Lemma 5.1.

Theorem 5.7. Convex functional means associated by operator ones are $L B P L$ preserving.
Finally, like the Kubo-Ando theory, we attempt to define convex functional means axiomatically. Let $\Sigma_{0}$ be the class of convex functional means $f \sigma g \in \Gamma$ for $(f, g) \in \Gamma^{2}$ which satisfies the monotonicity, semi-continuity, homegeneity, transformer inequality and normalization. This class $\Sigma_{0}$ is worth considering by the following result for constant functionals corresponding to Theorem 4.10:
Theorem 5.8. $F=(\alpha \mathbf{I}) \sigma(\beta \mathbf{I})$ is a constant functional for $\sigma \in \Sigma_{0}$.
Proof. By the transformer equality for invertible $T$, we have

$$
F=(\alpha \mathbf{I}) \sigma(\beta \mathbf{I})=(T(\alpha \mathbf{I})) \sigma\left({ }_{T}(\beta \mathbf{I})\right)={ }_{T}((\alpha \mathbf{I}) \sigma(\beta \mathbf{I})) .
$$

For all nonzero vectors $x$ and $y$, we can take an invertible operator $T$ with $T x=y$, and consequently $F(x)=C$ for some $C \in \mathbb{R}$ for all nonzero vectors $x$. Suppose $F \neq C \mathbf{I}$. Then, by $F \in \Gamma$, we have $F(0)=C$ and $F(x)=\infty$ for $x \neq 0$. Then, by the homogeneity,

$$
F_{n}(x) \equiv\left(\frac{\alpha}{n} \mathbf{I}\right) \sigma\left(\frac{\beta}{n} \mathbf{I}\right)(x)=\frac{1}{n} F(x)=\infty
$$

for all nonzero $x$ while the semicontinuity implies $F_{n} \downarrow 0 \mathbf{I}$, which is the contradiction.
To connect convex functional means to operator ones, we consider a subclass $\Sigma$ of $\Sigma_{0}$ in which $\sigma$ is LBPL preserving. Let $\Gamma_{0}$ be the locally bounded PL functionals in $\Gamma$. Restricting ourselves to $\Gamma_{0}$, we have a final theorem like the Kubo- Ando theory (Here the order is defined as in Corollary 4.9) :
Theorem 5.9. The map, $\left.m \rightarrow \sigma_{m}\right|_{\Gamma_{0}}$, gives an order-preserving bijection from the class of operator means onto $\left.\Sigma\right|_{\Gamma_{0}}$.
Proof. Let $\sigma \in \Sigma$. Since $\sigma$ is LBPL preserving, then $f_{A} \sigma f_{B}$ is also locally bounded PL functional and hence equal to $f_{C}$ for some positive operator $C$, say $A m B$. Then the binary operation $m$ on positive operators satisfies also the monotonicity, the semi-continuity, the transformer inequality and the normalization in the operator sense, that is, $m$ is an operator mean by the Kubo-Ando theory [11]. The injectivity is clear since all functionals belong to $\Gamma_{0}$ here.

Finally, we pay attention to the relation to Atteia and Raïssouli's means [3]: Their harmonic mean $f \tau_{h} g$ is defind using the conjugates by

$$
f \tau_{h} g=\left(\frac{f^{*}+g^{*}}{2}\right)^{*}
$$

which is also available in a complex space in [16]. On the other hand, our harmonic mean $f \sigma_{!} g$ is $2(f: g)$, so that, they coincide. Thereby their convex functional means which are based on this harmonic one belongs to our class $\Sigma$ at least when we discuss functionals in $\Gamma_{0}$. For example, recall that their geometric mean $\tau$ in [3] is defined also as the arithmeticoharmonic mean:

Theorem 5.10. If $f$ and $g$ belong to $\Gamma_{0}$, then $f \tau g=f \sigma_{\sharp} g$.


$$
f \gamma_{0} g=f \sigma_{\nabla} g, f \gamma_{0}^{*} g=f \sigma_{!} g, f \gamma_{n} g=\left(f \gamma_{n-1} g\right) \sigma_{\nabla}\left(f \gamma_{n-1}^{*} g\right), f \gamma_{n}^{*} g=\left(f \gamma_{n-1} g\right) \sigma_{!}\left(f \gamma_{n-1}^{*} g\right)
$$

Clearly, we have $\gamma_{n}$ and $\gamma_{n}^{*}$ are convex functional means associated with the corresponding operator means $\nabla_{n}$ and $!_{n}$. Thus they have the integral representation with the corresponding measure $\sigma_{\nabla_{n}}$ and $\sigma_{!_{n}}$. Moreover they are determined only for operators. Since $f_{A} \tau f_{B}=f_{A \sharp B}=f_{A} \sigma_{\sharp} f_{B}$, we have they coinside.

In [16], the geometrico-harmonic mean $f \tau_{g h} g$ also intorduced, which is equal to $\sigma_{g h}$ associated with the geometrico-harmonic operator mean $g h$ for quadratic convex functionals with the same domain in $\Gamma_{0}$. Clearly other means in [3] coincide with our ones for convex functionals with the same domain. But the following example shows that our means can be defined even if dom $f \neq \operatorname{dom} g$. Thus we have an exact extension for their convex functional means, even for the geometric mean:

Example. Let $C$ and $D$ be one-dimensional subspaces orthogonal each other. Then $E=$ $C+D$ is the closed (2-dimensional) subspace. Thereby, similarly to the paragraph before Lemma 3.1, we have and hence

$$
\mathbf{1}_{C}: \mathbf{1}_{D}=\frac{1}{4} \mathbf{1}_{(C+D) / 2}=\mathbf{1}_{E / 2}=\mathbf{1}_{E}
$$

so that

$$
g_{1} \equiv \mathbf{1}_{C} \tau_{h} \mathbf{1}_{D}=\mathbf{1}_{E}
$$

by $s \mathbf{1}_{C}=\mathbf{1}_{C}$ for all $s>0$. On the other hand, it is easy to see

$$
\mathbf{1}_{C}+\mathbf{1}_{D}=\mathbf{1}_{C \cap D}=\mathbf{1}_{\{0\}} \quad \text { and hence } \quad f_{1} \equiv \mathbf{1}_{C} \tau_{a} \mathbf{1}_{D}=\mathbf{1}_{\{0\}}
$$

for the aritmetic mean $\tau_{a}$. Thereby

$$
g_{2} \equiv f_{1} \tau_{h} g_{1}=\mathbf{1}_{\{0\}+E}=\mathbf{1}_{E}=g_{1} \quad \text { and } \quad f_{2} \equiv f_{1} \tau_{a} g_{2}=\mathbf{1}_{\{0\} \cap E}=\mathbf{1}_{\{0\}}=f_{1}
$$

Thus this procedure shows $f_{n}=f_{1} \neq g_{1}=g_{n}$ for all $n$, and hence they cannot coincide and their geometric mean cannot be defined.

On the other hand, since

$$
\frac{1+t}{2} \mathbf{1}_{C}: t_{\frac{1+t}{2 t}} \mathbf{1}_{D}=\mathbf{1}_{\frac{2}{1+t} C}: \mathbf{1}_{\frac{2 t}{1+t} D}=\mathbf{1}_{C}: \mathbf{1}_{D}=\mathbf{1}_{E}
$$

for all $t>0$, then our geometric mean $\sigma_{\sharp}$ is

$$
\left(\mathbf{1}_{C} \sigma_{\sharp} \mathbf{1}_{D}\right)(x)=\int_{(0, \infty)} \mathbf{1}_{E}(x) \frac{4}{1+t} d \mu_{\sharp}(t)=\int_{(0, \infty)} \mathbf{1}_{E}(x) d \mu_{\sharp}(t)=\mathbf{1}_{E}
$$

Moreover this computation shows that

$$
\mathbf{1}_{C} \sigma_{m} \mathbf{1}_{D}=\mathbf{1}_{E}
$$

if the measure $\mu_{m}$ for an operator mean $m$ satisfies $\mu_{m}(\{0\})=\mu_{m}(\{\infty\})=0$ like the geometric operator mean $\sharp$.

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