The concavity of generalized entropy powers

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Abstract

In this note we introduce a new family of entropy powers which are related to generalized entropies, called Sharma-Mittal entropies, and we prove their concavity along diffusion processes generated by $L^2$-Wasserstein gradient flows of corresponding entropy functionals. This result extends the result of Savaré and Toscani on the concavity of Rényi entropy powers (IEEE Trans. Inf. Theory, 2014) and reveals a connection to Rényi entropy power inequalities by Bobkov and Marsiglietti (IEEE Trans. Inf. Theory, 2017).

Index Terms

entropy power, Sharma-Mittal entropy, concavity, $L^2$-Wasserstein gradient flow

I. INTRODUCTION

ENTROPY power has been introduced by Shannon in his seminal paper [18]. Given a continuous random vector $X$ with values in $\mathbb{R}^d$, the entropy power $\mathcal{N}(X)$ is defined by

$$\mathcal{N}(X) = \exp \left( \frac{2}{d} \mathcal{H}(X) \right)$$

where

$$\mathcal{H}(X) = -\int_{\mathbb{R}^d} u(x) \log u(x) \, dx,$$

is the Shannon (also known as Boltzmann-Gibbs) entropy and $u$ is the probability density of $X$ [6]. The entropy power is a superadditive functional, i.e. for any two independent random vectors $X$ and $Y$ it holds

$$\mathcal{N}(X + Y) \geq \mathcal{N}(X) + \mathcal{N}(Y),$$

with equality if and only if $X$ and $Y$ are Gaussian random vectors with independent identically distributed components. This is the famous entropy power inequality (EPI), which was partially proved already by Shannon in [18], who used it to obtain a lower bound on the channel capacity, but the complete proof was given later by Stam [20]. The EPI and its refinements were subject of extensive research in information theory, to name only few [7], [8], [14], [16], [25]. Particularly important case are Gaussian perturbations of a random vector $X$. Let $Z$ be distributed according to the standard Gaussian, and let us denote $X_t = X + \sqrt{t}Z$ for $t \geq 0$, then the following refinement of the EPI has been proved by Costa [5]

$$\mathcal{N}(X_t) \geq (1 - t)\mathcal{N}(X_0) + t\mathcal{N}(X_1), \quad \forall t \in [0, 1].$$

The latter can be rephrased as the concavity of the entropy power along the stochastic process $(X_t)_{t>0}$, i.e.

$$\frac{d^2}{dt^2} \mathcal{N}(X_t) \leq 0.$$

Since $(X_t)_{t>0}$ is a diffusion process whose probability densities $u_t(x)$, $t > 0$, are governed by the linear diffusion or heat equation

$$\partial_t u_t = \Delta u_t, \quad (x, t) \in \mathbb{R}^d \times (0, +\infty),$$

one can also say that $\mathcal{N}(X_t)$ is concave along solutions to (3), where $\Delta$ denotes the Laplace operator in $\mathbb{R}^d$. A remarkable proof of the concavity of the EPI was given by Villani in [26], and we will explore this idea to prove our main result in Section III.

The entropy $\mathcal{H}(X)$ and the heat equation (3) are intimately related in a geometric sense. Namely, it has been proved in [10] that the heat equation constitutes a gradient flow of the entropy functional $\mathcal{H}(u) = -\mathcal{H}(X)$ with respect to a transport distance called $L^2$-Wasserstein distance on the space of probability densities of finite second moment. In notation $\mathcal{H}(u) = -\mathcal{H}(X)$, $u$ always denotes the density of the corresponding random vector $X$. Informally speaking, we can write equation (3) in its Wasserstein gradient flow form [10]

$$\partial_t u_t = \text{div} \left( u_t \nabla \frac{\delta \mathcal{H}(u_t)}{\delta u_t} \right), \quad \frac{\delta \mathcal{H}(u_t)}{\delta u_t} = \log u_t,$$

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where \( \delta \tilde{H}(u)/\delta u \) denotes the variational derivative of the functional \( \tilde{H}(u) \), \( \text{div} \) denotes the divergence operator, and \( \nabla \) the Euclidean gradient in \( \mathbb{R}^d \). Here we will use this gradient flow structure in a formal way and interested reader is referred to [1] for details.

The above described link between the entropy power \( \mathcal{N}(X) \) and the heat equation [3] was a cornerstone for an extension of the entropy power for Rényi entropies proposed by Savaré and Toscani in [17]. Defining \( p \)-th Rényi entropy power (for \( p > 1 - 2/d \)) as

\[
\mathcal{P}_p(X) = \exp \left( \sigma_p \mathcal{R}_p(X) \right),
\]

where \( \sigma_p = 2/d + p - 1 \) and

\[
\mathcal{R}_p(X) = \frac{1}{1 - p} \log \left( \int_{\mathbb{R}^d} u^p(x) dx \right)
\]

is the Rényi entropy of order \( p > 0, p \neq 1 \), they proved the concavity of \( \mathcal{P}_p(X_t) \), i.e.

\[
\frac{d^2}{dt^2} \mathcal{P}(X_t) \leq 0,
\]

along the diffusion process whose probability densities \( u_t(x), t > 0 \), solve the nonlinear diffusion equation

\[
\partial_t u_t = \Delta u_t^p, \quad (x, t) \in \mathbb{R}^d \times (0, +\infty).
\]

This result was fruitful for obtaining Gagliardo-Nirenberg type functional inequalities with sharp constants [22] and improved decay rates for convergence of solutions of [6] to the self-similar profile [4, 21]. In fact the choice of \( \sigma_p \) in (4) comes from the requirement that the functional

\[
\mathcal{Q}_p(u_t) = \mathcal{P}(X_t) \frac{d}{dt} \mathcal{R}_p(X_t),
\]

with the time derivative along solutions to [6], is invariant with respect to mass conservative dilations, i.e. \( \mathcal{Q}_p(\lambda^d u(\lambda x)) = \mathcal{Q}_p(u(x)) \) holds for all \( \lambda > 0 \). Dilation invariance jointly with the self-similarity structure of solutions to [6] [24] provides the isoperimetric inequality [17] cf. Theorem 2], which is the basis of Gagliardo-Nirenberg type inequalities.

Equation (6) is a well known and studied partial differential equation with plethora of applications [24]. It also has a geometric interpretation in terms of the \( L^2 \)-Wasserstein distance. For \( p > \max(1 - 1/d, d/(d + 2)) \), \( p \neq 1 \), it constitutes the \( L^2 \)-Wasserstein gradient flow of the entropy functional \( \mathcal{T}_p(u) = -\mathcal{T}_p(X) \), where

\[
\mathcal{T}_p(X) = \frac{1}{1 - p} \left( \int_{\mathbb{R}^d} u^p(x) dx - 1 \right), \quad p \neq 1,
\]

is the Tsallis entropy of order \( p \) [12]. In the gradient flow form, equation (6) reads

\[
\partial_t u_t = \text{div} \left( u_t \nabla \frac{\delta \mathcal{T}_p(u_t)}{\delta u_t} \right), \quad \frac{\delta \mathcal{T}_p(u_t)}{\delta u_t} = \frac{p}{p - 1} u_t^{p-1}.
\]

Hence, the Rényi entropy power \( \mathcal{P}_p(X) \) is related to both Rényi and Tsallis entropies. We will resolve this ambiguity by introducing generalized entropy powers which will enable us to see a broad picture.

On the other hand, Bobkov and Chistyakov introduced another version of Rényi entropy power [2], which is a straightforward extension of the Shannon’s entropy power,

\[
\mathcal{B}_p(X) = \exp \left( \frac{2}{d} \mathcal{R}_p(X) \right).
\]

Factor \( 2/d \) in the exponent makes the functional \( \mathcal{B}_p(X) \), likewise \( \mathcal{N}(X) \), homogeneous of order two, i.e. \( \mathcal{B}_p(\lambda X) = \lambda^2 \mathcal{B}_p(X) \) for all \( \lambda \in \mathbb{R} \). They proved the following entropy power inequality: for \( p > 1 \) and \( n \geq 3 \), let \( X_1, X_2, \ldots, X_n \) be independent continuous random vectors in \( \mathbb{R}^d \), then

\[
\mathcal{B}_p(X_1 + X_2 + \ldots + X_n) \geq \frac{1}{n} p^n \sum_{k=1}^n \mathcal{B}_p(X_k).
\]

Contrary to [2], the latter does not hold for \( n = 2 \). A counterexample can be found in [2]. The case of \( p \in (0, 1) \) has been recently discussed in [11]. Furthermore, in [3] Bobkov and Marsiglietti extended the Rényi entropy power inequality to the following form: given independent continuous random vectors \( X \) and \( Y \) in \( \mathbb{R}^d \), then

\[
\mathcal{B}_p^\alpha(X + Y) \geq \mathcal{B}_p^\alpha(X) + \mathcal{B}_p^\alpha(Y)
\]

for all \( \alpha \geq (p + 1)/2 \) and \( p > 1 \).
Remark 1. Observe that for $\alpha = d(p−1)/2 + 1$ the $\alpha$-power of the functional $B_p(X)$ coincides with the functional $P_p(X)$, i.e. according to (7), for $p > 1$ we have the EPI
\[ P_p(X + Y) \geq P_p(X) + P_p(Y). \] (8)

Inspired by (7) we introduce a two-parameter generalization of the entropy power that we will call Sharma-Mittal entropy power of order $(p, q)$ (or simply Sharma-Mittal entropy power). Its relation with Sharma-Mittal entropies [19] will be clarified below.

Definition 1. Let $X$ be a continuous random vector in $\mathbb{R}^d$. For $p > 1−2/d$ and $q > 0$ we define Sharma-Mittal entropy power of order $(p, q)$ as
\[ N_{p,q}(X) = \exp \left( \sigma_q R_p(X) \right), \] (9)
where $\sigma_q = 2/d + q - 1$, and $R_p(X)$ is the Rényi entropy of order $p$.

It is apparent from the definition that setting $q = 2(\alpha−1)/d + 1$, for $\alpha \geq (p+1)/2$ and $p > 1$, inequality (7) reads as the EPI for Sharma-Mittal entropy powers
\[ N_{p,q}(X + Y) \geq N_{p,q}(X) + N_{p,q}(Y). \] (10)

In particular, for $q = p$, the Sharma-Mittal entropy power $N_{p,q}(X)$ coincides with the functional $P_p(X)$, and inequality (10) reduces to (8). If $q = 1$, then $N_{p,q}(X)$ coincides with the functional $B_p(X)$, but in this case $q = 1$ implies $\alpha = 1$, which in further requires $p = 1$, hence (10) actually reduces to (2).

While the entropy power inequality (10) is an immediate consequence of (7) and the definition of $N_{p,q}(X)$, our aim in studying this subject was to complement (10) with the concavity property of $N_{p,q}(X_t)$. Thus, to extend the result of Savaré and Toscani to the wider class of functionals. In light of the above interpretation of the concavity of entropy powers along the gradient flows of respective entropy functionals, it is appealing to formally consider the $L^2$-Wasserstein gradient flow
\[ \partial_t u_t = \text{div} \left( u_t \nabla \frac{\delta S_{p,q}(u_t)}{\delta u_t} \right), \] (11)
of functional $\tilde{S}_{p,q}(u) = -S_{p,q}(X)$, where
\[ S_{p,q}(X) = \frac{1}{1−q} \left[ \left( \int_{\mathbb{R}^d} u(x)^p \, dx \right)^{\frac{1}{p−1}} - 1 \right], \quad p, q \neq 1, \]
is the Sharma-Mittal entropy of order $(p, q)$ [19]. Since the variational derivative of $\tilde{S}_{p,q}(u)$ equals
\[ \frac{\delta \tilde{S}_{p,q}(u)}{\delta u} = \frac{p}{p−1} \left( \int_{\mathbb{R}^d} u^p \, dx \right)^{\frac{p−1}{p}} u^{p−1}, \]
partial differential equation (11) is for smooth positive solutions equivalent to the following nonlinear and non-local diffusion equation
\[ \partial_t u_t = \left( \int_{\mathbb{R}^d} u_t^p \, dx \right)^{\frac{p−1}{p}} \Delta u_t^p, \quad (x, t) \in \mathbb{R}^d \times (0, +\infty). \] (12)

This equation appeared in the literature [2] in studying related diffusion processes. Now we can state our main result which provides a remarkable geometric relation between generalized entropies and their powers.

Theorem 1. Let $p > 1−2/d$ and $q > 0$, and let $(X_t)_{t \geq 0}$ be a diffusion process whose probability densities $u_t(x)$, $t > 0$, are smooth, strictly positive and rapidly decaying solutions to equation (12), then
\[ \frac{d^2}{dt^2} N_{p,q}(X_t) \leq 0, \quad t > 0. \]

We close this introductory section by few important remarks.

First, observe that for $p = q$, the Sharma-Mittal entropy $S_{p,q}(X)$ equals to the Tsallis entropy $T_p(X)$, hence equation (12) reduces to the nonlinear diffusion equation (6), which is the gradient flow of $T_p(u)$. Since the concavity of the Rényi entropy power $P_p(X)$, as called in [17], holds along the gradient flow of $T_p(u)$, from this perspective it could also be called the Tsallis entropy power.

Second, on the limit as $q \to 1$ the Sharma-Mittal entropy $S_{p,q}(X)$ becomes the Rényi entropy $R_p(X)$, and equation (12) reduces to
\[ \partial_t u_t = \left( \int_{\mathbb{R}^d} u_t^p \, dx \right)^{-1} \Delta u_t^p, \quad (x, t) \in \mathbb{R}^d \times (0, +\infty). \] (13)
which is formally the $L^2$-Wasserstein gradient flow of the Rényi entropy functional $\tilde{R}_p(u) = -R_p(X)$ \[4\]. Thus, according to Theorem \[1\], the Rényi entropy power $B_p(X)$ is concave along the gradient flow of $\tilde{R}_p(u)$.

Last, but not least, the choice of $\sigma_q = 2/d + q - 1$ in the definition of $N_{p,q}(X)$ follows an analogous argument like the choice of $\sigma_p$ in \[17\], as discussed above. Namely, we require that the functional

$$Q_{p,q}(u_t) = N_{p,q}(X_t) \frac{d}{dt} R_p(X_t),$$

with the time derivative along solutions to \[12\], is invariant with respect to mass conservative dilations.

In Section \[III\] we outline algebraic relation between generalized entropies and respective entropy powers, which complements the above stressed geometric relation. In addition, we provide sufficient conditions for the entropy functional $\tilde{S}_{p,q}(u)$ being geodesically convex which, according to the theory developed in \[1\], makes the gradient flow structure \[11\] rigorous. The proof of Theorem \[1\] is given in Section \[III\].

\section{II. Generalized entropies and respective entropy powers}

The Shannon’s idea of axiomatic foundation of the entropy \[18\] has become a fertile ground for its generalizations. In \[15\], Rényi proposed an alteration of the Fadeev’s set of axioms for the discrete Shannon’s entropy, which in the continuous setting leads to the following entropy of order $p \neq 1$:

$$R_p(X) = \frac{1}{1-p} \log \left( \int_{\mathbb{R}^d} u^p(x)dx \right).$$

Observe that $\lim_{p \to 1} R_p(X) = \mathcal{H}(X)$. Among others, further generalization has been proposed by Sharma and Mittal \[19\]. They introduce a two parameter entropy of order $(p, q)$ as

$$S_{p,q}(X) = \frac{1}{1-q} \left[ \left( \int_{\mathbb{R}^d} u(x)^p dx \right)^{\frac{1}{p}} - 1 \right], \quad p, q \neq 1.$$

Years later, in the framework of non-extensive thermodynamics, Tsallis proposed a new family of entropies \[23\]

$$T_p(X) = \frac{1}{1-p} \left( \int_{\mathbb{R}^d} u^p(x)dx - 1 \right), \quad p \neq 1.$$

Again observe that $\lim_{p \to 1} T_p(X) = \mathcal{H}(X)$, and furthermore $S_{p,p}(X) = T_p(X)$ for $p \neq 1$. In the context of Tsallis statistics \[13\], \[23\] it is customary to work with $q$-logarithm

$$\log_q(s) = \frac{1}{1-q} \left( s^{1-q} - 1 \right), \quad s > 0,$$

where $q \neq 1$, and its inverse, $q$-exponential

$$\exp_q(s) = \max \left( 1 + (1-q)s, 0 \right)^{\frac{1}{1-q}}.$$

In this notation we can write

$$S_{p,q}(X) = \log_q \left( \left( \int_{\mathbb{R}^d} u(x)^p dx \right)^{\frac{1}{p}} \right),$$

which better reveals similarities with Rényi entropies. Taking the $q$-exponential of the latter equation we find

$$\exp_q \left( S_{p,q}(X) \right) = \left( \int_{\mathbb{R}^d} u(x)^p dx \right)^{\frac{1}{p}},$$

and therefore, the Sharma-Mittal entropy power can be written as

$$N_{p,q}(X) = \left( \exp_q \left( S_{p,q}(X) \right) \right)^{\sigma_q},$$

where $\sigma_q = 2/d + q - 1$. This equation gives the direct relation between the Sharma-Mittal entropy and its power. In particular, for $q = p$ we can write

$$P_p(X) = \left( \exp_p \left( T_p(X) \right) \right)^{\sigma_p},$$

which gives the direct relation between the Rényi entropy power $P_p(X)$ and the Tsallis entropy. Both equations \[15\] and \[16\] resemble the original definition of the Shannon entropy power \[1\], which can be recovered on the limit as $p, q \to 1$:

$$\lim_{p, q \to 1} N_{p,q}(X) = \lim_{p \to 1} P_p(X) = N(X).$$

On the other hand, the Rényi entropy power $B_p(X)$ follows as

$$B_p(X) = \lim_{q \to 1} N_{p,q}(X).$$
Although we work only formally with gradient flows, let us briefly mention under which conditions the above gradient flow structure is rigorous. It has been shown in [1] Chapter 9 that the functional
\[ E_p(u) = \int_{\mathbb{R}^d} e_p(u(x))dx \quad \text{with} \quad e_p(z) = \frac{1}{p - 1} z^p, \]
is for \( p \geq 1 - 1/d, p \neq 1 \), geodesically convex on the space of probability measures of finite second moment \( \mathcal{P}_2(\mathbb{R}^d) \). Roughly speaking this means that \( E_p(u) \) is convex along the geodesic curve \( (u_t)_{0 \leq t \leq 1} \) connecting any two measures \( u_0, u_1 \in \mathcal{P}_2(\mathbb{R}^d) \), i.e.
\[ E_p(u_t) \leq (1 - t)E_p(u_0) + tE_p(u_1), \quad \forall t \in [0, 1]. \]
At the expense of rigor, but for the sake of simplicity of exposition, we denote both probability measures and their densities with respect to the Lebesgue measure simply by \( u \).

Now observe that for \( p > 1 \) the Sharma-Mittal entropy functional can be written as
\[ \tilde{S}_{p,q}(u) = -\log_q \left( \left( (p - 1)E_p(u) \right)^{\frac{1}{p-1}} \right) = s_{p,q}(E_p(u)), \]
where
\[ s_{p,q}(z) = -\log_q \left( ((p - 1)z)^{\frac{1}{p-1}} \right). \]
Easy calculation gives that \( s_{p,q}(z) \) is non-decreasing for \( p > 1 \) and convex for \( q \geq p > 1 \). Hence, the composition with \( E_p(u) \) and \( \mathcal{P}_2(\mathbb{R}^d) \) yield the geodesic convexity of \( \tilde{S}_{p,q}(u) \) when \( q \geq p > 1 \). Then, according to the theory developed in [1], in this range of parameters the gradient flow structure \( (11) \) is well-posed.

### III. Proof of Theorem [1]

In order to prove our main result, we closely follow the approach of Savaré and Toscani in [17]. After introducing auxiliary functional \( \mathcal{I}_p(u) \) in (17), they also introduce a generalization of the Fisher information
\[ \mathcal{I}_p(u) = \int_{\mathbb{R}^d} \frac{\lvert \nabla u \rvert}{u} dx = \int_{\mathbb{R}^d} u |\nabla e'_p(u)|^2 dx, \]
and the second-order functional
\[ \mathcal{J}_p(u) = 2 \int_{\mathbb{R}^d} u^p \left( |\nabla^2 e'_p(u)|^2 + (p - 1)(\Delta e'_p(u))^2 \right) dx. \]
The following proposition has been proved in [17 Proposition 3].

**Proposition 2.** Let \( u_t(x), t > 0 \), be smooth, strictly positive and rapidly decaying probability densities solving the nonlinear diffusion equation (6), then
\[ -\frac{d}{dt} E_p(u_t) = \mathcal{I}_p(u_t), \quad t > 0, \tag{21} \]
\[ -\frac{d}{dt} \mathcal{I}_p(u_t) = \mathcal{J}_p(u_t), \quad t > 0. \tag{22} \]

Identity (21) actually says that the generalized Fisher information \( \mathcal{I}_p(u) \) equals to the production of the Tsallis entropy functional \( \tilde{E}_p(u) \) along its own gradient flow. Equation (22) can be interpreted as the production of the Fisher information along the gradient flow of \( \tilde{E}_p(u) \), which gives the second-order functional \( \mathcal{J}_p(u) \).

Using these results we can prove the following analogous statement.

**Corollary 3.** Let \( u_t(x), t > 0 \), be smooth, strictly positive and rapidly decaying probability densities solving the nonlinear diffusion equation (12), then
\[ -\frac{d}{dt} E_p(u_t) = \left( (p - 1)E_p(u_t) \right)^{\frac{1}{p-1}} \mathcal{I}_p(u_t), \quad t > 0, \tag{23} \]
\[ -\frac{d}{dt} \mathcal{I}_p(u_t) = \left( (p - 1)E_p(u_t) \right)^{\frac{p-2}{p-1}} \mathcal{J}_p(u_t), \quad t > 0. \tag{24} \]

**Proof.** Observe that for smooth and strictly positive solutions equation (12) can be equivalently written as
\[ \partial_t u_t = \left( \int_{\mathbb{R}^d} u_t^p dx \right)^{\frac{p}{p-1}} \nabla \cdot (u_t \nabla e'_p(u_t)). \]
Thus, assuming in addition rapid decay of solutions to (12) we can freely integrate by parts and calculate:

\[
\frac{d}{dt} \mathcal{E}_p(u_t) = \int_{\mathbb{R}^d} e_p'(u_t) \partial_t u_t \nabla x = \left( \int_{\mathbb{R}^d} u_t^p dx \right)^{\frac{p-1}{p}} \int_{\mathbb{R}^d} e_p'(u_t) \nabla \cdot (u_t \nabla e_p'(u_t)) dx \nabla x .
\]

Identity (23) then obviously follows from definitions (17) and (19). Identity (24) follows in analogous straightforward way from (22).

Recall the definition of the Sharma-Mittal entropy power (9),

\[
\mathcal{N}_{p,q}(X) = \exp (\sigma_q \mathcal{R}_p(X)),
\]

where \(\sigma_q = 2/d + q - 1\). Taking a stochastic process \((X_t)_{t \geq 0}\), we simply calculate

\[
\frac{d}{dt} \mathcal{N}_{p,q}(X_t) = \sigma_q \mathcal{N}_{p,q}(X_t) \frac{d}{dt} \mathcal{R}_p(X_t),
\]

\[
\frac{d^2}{dt^2} \mathcal{N}_{p,q}(X_t) = \sigma_q \mathcal{N}_{p,q}(X_t) \left( \sigma_q \left( \frac{d \mathcal{R}_p(X_t)}{dt} \right)^2 + \frac{d^2 \mathcal{R}_p(X_t)}{dt^2} \right).
\]

Therefore, \(\mathcal{N}_{p,q}(X_t)\) is concave, i.e. \(\frac{d^2 \mathcal{N}_{p,q}(X_t)}{dt^2} \leq 0\) if and only if

\[
- \frac{d^2 \mathcal{R}_p(X_t)}{dt^2} \geq \sigma_q \left( \frac{d \mathcal{R}_p(X_t)}{dt} \right)^2 .
\]

(25)

Writing the Rényi entropy \(\mathcal{R}_p(X)\) in terms of the functional \(\mathcal{E}_p(u)\) as

\[
\mathcal{R}_p(X) = \frac{1}{1 - p} \log \left( (p - 1) \mathcal{E}_p(u) \right),
\]

we further calculate

\[
\frac{d}{dt} \mathcal{R}_p(X_t) = \frac{1}{1 - p} \frac{d}{dt} \mathcal{E}_p(u_t),
\]

\[
\frac{d^2}{dt^2} \mathcal{R}_p(X_t) = \frac{1}{1 - p} \left( \frac{d^2}{dt^2} \mathcal{E}_p(u_t) - \left( \frac{d}{dt} \mathcal{E}_p(u_t) \right)^2 \right) .
\]

If \((X_t)_{t \geq 0}\) is a stochastic process whose density function \(u_t(x)\) solves (12), then employing the identities from Corollary [3] it follows

\[
\frac{d}{dt} \mathcal{R}_p(X_t) = \left( (p - 1) \mathcal{E}_p(u_t) \right)^{\frac{2p - q - 1}{p - 1}} \mathcal{I}_p(u_t),
\]

\[
\frac{d^2}{dt^2} \mathcal{R}_p(X_t) = \left( (p - 1) \mathcal{E}_p(u_t) \right)^{\frac{3p - 2q - 1}{p - 1}} \left( \frac{2p - q - 1}{p - 1} \frac{\mathcal{I}_p^2(u_t)}{\mathcal{E}_p(u_t)} - \mathcal{J}_p(u_t) \right) .
\]

Concavity condition (25) then becomes equivalent to

\[
\mathcal{J}_p(u_t) \geq (\sigma_q + 2p - q - 1) \frac{\mathcal{I}_p^2(u_t)}{(p - 1) \mathcal{E}_p(u_t)},
\]

which can be further written as

\[
\mathcal{J}_p(u_t) \left( \int_{\mathbb{R}^d} u_t^p dx \right) \geq 2 \left( \frac{1}{d} + p - 1 \right) \mathcal{I}_p^2(u_t) .
\]

(27)

The last inequality has been demonstrated in [17] cf. inequality (26), hence, the proof of Theorem [1] is finished.
for arbitrary $\lambda \in \mathbb{R}$, where $I$ denotes the identity matrix $d \times d$. Hence, for every $\lambda \in \mathbb{R}$ and $p > 1 - 1/d$ it holds
\[
0 \leq \int_{\mathbb{R}^d} u^p \left( |\nabla^2 e_p'(u) + \lambda I|^2 + (p - 1)(\Delta e_p'(u) + d\lambda)^2 \right) \, dx.
\]
Expanding the right hand side we obtain
\[
0 \leq \int_{\mathbb{R}^d} u^p \left( |\nabla^2 e_p'(u)|^2 + 2\lambda \Delta e_p'(u) + d\lambda^2 \right) \, dx
+ (p - 1) \int_{\mathbb{R}^d} u^p \left( (\Delta e_p'(u))^2 + 2d\lambda \Delta e_p'(u) + d^2\lambda^2 \right) \, dx.
\]
Integrating by parts and using the fact that $\lambda$ for arbitrary of Bobkov and Marsiglietti [3] by proving the concavity of Sharma-Mittal entropy powers along the flows of the corresponding Sharma-Mittal entropy functionals. Thus, we generalized the result of Savaré and Toscani [17], of the Rényi entropy power in the literature. To conclude, the contribution of our result and its settlement in the literature is best seen from the following table.

\[
\begin{align*}
\text{EPI} & \quad \mathcal{N}(X) & \quad \mathcal{P}_p(X) & \quad \mathcal{B}_p(X) & \quad \mathcal{N}_{p,q}(X) \\
\text{concavity} & \quad \text{Stam [29]} & \quad \text{Bobkov and Marsiglietti [5]} & \quad \text{Bobkov and Marsiglietti [5]} & \quad \text{Bobkov and Marsiglietti [5]}
\end{align*}
\]

\[
\begin{align*}
\text{Costa [5]} & \quad \text{Savaré and Toscani [17]} & \quad \text{Theorem 1} & \quad \text{Theorem 1}
\end{align*}
\]

\[
\begin{align*}
\text{TABLE I} & \quad \text{CONTRIBUTION AND SETTLEMENT OF OUR RESULT IN THE LITERATURE}
\end{align*}
\]

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