FROM JOINT CONVEXITY
OF QUANTUM RELATIVE ENTROPY
TO A CONCAVITY THEOREM OF LIEB

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ABSTRACT. This paper provides a succinct proof of a 1973 theorem of Lieb that establishes the concavity of a certain trace function. The development relies on a deep result from quantum information theory, the joint convexity of quantum relative entropy, as well as a recent argument due to Carlen and Lieb.

1. INTRODUCTION

In his 1973 paper on trace functions, Lieb establishes an important concavity theorem [Lie73, Thm. 6] concerning the trace exponential.

Theorem 1 (Lieb). Let \( H \) be a fixed self-adjoint matrix. The map

\[
A \mapsto \text{tr} \exp (H + \log A)
\]

is concave on the positive-definite cone.

The most direct proof of Theorem 1 is due to Epstein [Eps73]; see Ruskai’s papers [Rus02, Rus05] for a condensed version of this argument. Lieb’s original proof develops the concavity of the function \( 1 \) as a corollary of another deep concavity theorem [Lie73, Thm. 1]. In fact, many convexity and concavity theorems for trace functions are equivalent with each other, in the sense that the mutual implications follow from relatively easy arguments. See [Lie73, §5] and [CL08, §5] for discussions of this point.

The goal of this paper is to demonstrate that a modicum of convex analysis allows us to derive Theorem 1 directly from another major theorem, the joint convexity of the quantum relative entropy. The literature contains several elegant, conceptual proofs of the latter result; for example, see [Eff09]. These arguments now deliver Lieb’s theorem as an easy corollary.

The author’s interest in Theorem 1 stems from its striking applications in random matrix theory; refer to the paper [Tro10] for a detailed discussion. Researchers concerned with these developments may find the current approach to Lieb’s theorem more transparent than earlier treatments.
The main ideas in our presentation are drawn from the work of Carlen and Lieb [CL08], so this dispatch does not contain a truly novel technique. Nevertheless, this paper should be valuable because it provides a geometric intuition for Theorem 1 and connects it to another major result.

2. Background

Our argument rests on the properties of a function, called the quantum relative entropy, which can be interpreted as a measure of dissimilarity between two positive-definite matrices.

**Definition 2.** Let \( X, Y \) be positive-definite matrices. The quantum relative entropy of \( X \) with respect to \( Y \) is defined as

\[
D(X; Y) := \text{tr}(X \log X - X \log Y - (X - Y)).
\]

Other appellations for this function include quantum information divergence and von Neumann divergence.

The quantum relative entropy has a nice geometric interpretation [DT07, §2.2 and §2.6]. Define the quantum entropy function \( \varphi(X) := \text{tr}(X \log X) \) for a positive-definite argument. The divergence \( D(X; Y) \) can be viewed as the difference between \( \varphi(X) \) and the best affine approximation of the entropy \( \varphi \) at the matrix \( Y \). That is,

\[
D(X; Y) = \varphi(X) - [\varphi(Y) + \langle \nabla \varphi(Y), X - Y \rangle].
\]

The entropy \( \varphi \) is a strictly convex function, which implies that the affine approximation strictly underestimates \( \varphi \). This observation yields the following result.

**Fact 3.** The quantum relative entropy is nonnegative:

\[
D(X; Y) \geq 0.
\]

Equality holds if and only if \( X = Y \).

In quantum statistical mechanics, Fact 3 is usually called Klein’s inequality. Another proof proceeds by showing that certain functional relations for scalars extend to matrix trace functions [Pet94, §2].

The convexity properties of quantum relative entropy have paramount importance. We require a major theorem, due to Lindblad [Lin74, Lem. 2], which encapsulates the difficulties of the proof.

**Fact 4 (Lindblad).** The quantum relative entropy is a jointly convex function. That is,

\[
D(tX_1 + (1-t)X_2; tY_1 + (1-t)Y_2) \leq t \cdot D(X_1; Y_1) + (1-t) \cdot D(X_2; Y_2) \quad \text{for } t \in [0, 1],
\]

where \( X_i \) and \( Y_i \) are positive definite for \( i = 1, 2 \).

Fact 4 follows easily from Lieb’s main concavity theorem [Lie73, Thm. 1]; Bhatia’s book [Bha97, §IX.6 and Prob. IX.8.17] offers a clear account of this approach. The literature contains several other elegant proofs; see the papers [Uhl77, PW78, And79, Han06]. We single out Effros’ work [Eff09] because it is accessible to researchers with experience in matrix theory and convex analysis.

Our final tool is a basic result from convex analysis which ensures that partial maximization of a concave function produces a concave function [CL08, Lem. 2.3]. We include the simple proof.
Proposition 5. Let $f(\cdot: \cdot)$ be a jointly concave function. Then the function $y \mapsto \max_x f(x; y)$ obtained by partial maximization is concave, assuming the maximum is always attained.

Proof. For each pair of points $y_1$ and $y_2$, there are points $x_1$ and $x_2$ that satisfy

$$f(x_1; y_1) = \max_x f(x; y_1) \quad \text{and} \quad f(x_2; y_2) = \max_x f(x; y_2).$$

For each $t \in [0, 1]$, the joint concavity of $f$ implies that

$$\max_x f(x; ty_1 + (1-t)y_2) \geq f(tx_1 + (1-t)x_2; ty_1 + (1-t)y_2)$$

$$\geq t \cdot f(x_1; y_1) + (1-t) \cdot f(x_2; y_2)$$

$$= t \cdot \max_x f(x; y_1) + (1-t) \cdot \max_x f(x; y_2).$$

In words, the partial maximum is a concave function. \qed

3. PROOF OF LIEB’S THEOREM

We begin with a variational representation of the trace, which is a restatement of the fact that quantum relative entropy is nonnegative. The symbol $\succ$ denotes the positive-definite order.

Lemma 6 (Variational Formula for Trace). Let $Y$ be a positive-definite matrix. Then

$$\text{tr } Y = \max_{X \succ 0} \text{tr}(X \log Y - X \log X + X).$$

Proof. Introduce the definition of the quantum relative entropy into Fact and rearrange to reach

$$\text{tr } Y \geq \text{tr}(X \log Y - X \log X + X).$$

When $X = Y$, both sides are equal, which yields the advertised result. \qed

The main result follows quickly using the variational formula and the other tools we have assembled. As noted, the structure of this argument is parallel with the approach of Carlen–Lieb to another concavity theorem [CL08, Thm. 1.1].

Proof of Theorem In the variational formula, Lemma select $Y = \exp(H + \log A)$ to obtain

$$\text{tr } \exp(H + \log A) = \max_{X \succ 0} \text{tr}(X(H + \log A) - X \log X + X).$$

The latter expression can be written compactly using the quantum relative entropy:

$$\text{tr } \exp(H + \log A) = \max_{X \succ 0} [\text{tr}(XH) - (\text{D}(X; A) - \text{tr } A)].$$

For each self-adjoint matrix $H$, Fact implies that the bracket is a jointly concave function of the variables $A$ and $X$. It follows from Proposition that the right-hand side of defines a concave function of $A$. This observation establishes the theorem. \qed

Remark 7. The expression states that the function $f : H \mapsto \text{tr } \exp(H + \log A)$ is the Fenchel conjugate of $\text{D}(\cdot; A) - \text{tr } A$. This observation implies that $f$ is convex.
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