

# Convergence Rates of Characteristic Functions

M. Cramer

Institut für Theoretische Physik, Albert-Einstein Allee 11, Universität Ulm, Germany

## Abstract

This is a small note on the rate of convergence in the quantum central limit theorem. So far it only contains the 0-local and product state case, so in a sense the “classical” case. The more general, “truly quantum” case, will be added later.

## 1 Setting

We let  $\mathcal{X}$  a collection of lattices sites equipped with a distance  $d : \mathcal{X} \times \mathcal{X} \rightarrow \mathbb{N}$  and consider observables of the form

$$\hat{X} = \sum_{i \in \mathcal{X}} \hat{X}_i \quad (1)$$

with  $\hat{X}_i$  hermitian and  $R$ -local, i.e., it acts only on sites  $\{j \in \mathcal{X} \mid d(i, j) \leq R\}$ . For a given state  $\hat{\rho}$  we write  $\langle \cdot \rangle = \text{tr}[\hat{\rho} \cdot]$  and

$$\mu = \langle \hat{X} \rangle, \quad \sigma = \langle (\hat{X} - \mu)^2 \rangle. \quad (2)$$

We will be concerned with the characteristic function  $\phi : \mathbb{R} \rightarrow \mathbb{C}$ ,

$$\phi(t) = \langle e^{it\hat{X}} \rangle, \quad (3)$$

and its distance to the corresponding Gaussian characteristic function,

$$\Delta(t) = |\phi(t) - e^{i\mu t - \sigma^2 t^2/2}|. \quad (4)$$

We will write  $|\mathcal{X}| = N$ .

## 2 Product States and 0-local Observables

For 0-local observables and product states we have the following.

**Theorem 1** *Let  $R = 0$  and  $\hat{\rho} = \otimes_{i \in \mathcal{X}} \hat{\rho}_i$ . If*

$$|t| \sum_{i \in \mathcal{X}} \langle (\hat{X}_i - \langle \hat{X}_i \rangle)^2 \rangle^{1/2} \langle (\hat{X}_i - \mu_i)^4 \rangle^{1/2} \leq \frac{3}{5} \sigma^3 \quad (5)$$

*then*

$$\Delta(t/\sigma) \leq \frac{5}{12} e^{-t^2/4} |t|^3 \frac{\sum_{i \in \mathcal{X}} \langle (\hat{X}_i - \langle \hat{X}_i \rangle)^2 \rangle^{1/2} \langle (\hat{X}_i - \mu_i)^4 \rangle^{1/2}}{\sigma^3}. \quad (6)$$

*Remark 1:* Eq. (6) implies

$$\Delta(t/\sigma) \leq \frac{5}{12} e^{-t^2/4} |t|^3 \sqrt{\frac{\sum_{i \in \mathcal{X}} \langle (\hat{X}_i - \mu_i)^4 \rangle}{\sigma^2}}. \quad (7)$$

*Remark 2:* If  $\langle (\hat{X}_i - \langle \hat{X}_i \rangle)^2 \rangle = s^2$  and  $\langle (\hat{X}_i - \mu_i)^4 \rangle^{1/2} = \beta$ , one recovers the familiar i.i.d. case,

$$\Delta(t/\sigma) \leq \frac{5\beta}{12s^2} e^{-t^2/4} |t|^3 \frac{1}{\sqrt{N}} \quad \text{if } |t| \leq \frac{3s^2}{5\beta} N^{1/2}. \quad (8)$$

*Remark 3:* If the  $\hat{X}_i$  are bounded,  $\|\hat{X}_i\| \leq x$  then

$$\Delta(t/\sigma) \leq \frac{5}{12} e^{-t^2/4} |t|^3 \left( \frac{x^2 N}{\sigma^2} \right)^{3/2} \frac{1}{\sqrt{N}} \quad \text{if } |t| \leq \frac{3}{5} \left( \frac{\sigma^2}{x^2 N} \right)^{3/2} \sqrt{N}. \quad (9)$$

The proof of Theorem 1 is completely analogous to the classical case, see, e.g., pretty much any book on probability theory. We give it for completeness.

*Proof.* For  $i \in \mathcal{X}$  denote

$$\mu_i = \langle \hat{X}_i \rangle, \quad \sigma_i^2 = \langle (\hat{X}_i - \mu_i)^2 \rangle. \quad (10)$$

As  $\hat{\rho}$  is a product state, we have

$$\sigma^2 = \langle (\hat{X} - \mu)^2 \rangle = \sum_{i,j \in \mathcal{X}} \langle (\hat{X}_i - \mu_i)(\hat{X}_j - \mu_j) \rangle = \sum_{i \in \mathcal{X}} \sigma_i^2 \quad (11)$$

and

$$\Delta(t) = e^{-\sigma^2 t^2/2} \left| e^{\sigma^2 t^2/2} \prod_{i \in \mathcal{X}} \langle e^{it(\hat{X}_i - \mu_i)} \rangle - 1 \right|. \quad (12)$$

Now,

$$\langle e^{it(\hat{X}_i - \mu_i)} \rangle - 1 = - \int_0^t dx \int_0^x dy \langle (\hat{X}_i - \mu_i) e^{it(\hat{X}_i - \mu_i)} (\hat{X}_i - \mu_i) \rangle. \quad (13)$$

By the Cauchy–Schwarz inequality

$$|\langle \hat{A} \hat{B} \rangle|^2 = |\text{tr}[\hat{B} \sqrt{\hat{\rho}} \sqrt{\hat{\rho}} \hat{A}]|^2 \leq \text{tr}[\hat{B} \hat{\rho} \hat{B}^\dagger] \text{tr}[\hat{A}^\dagger \hat{\rho} \hat{A}] = \langle \hat{A} \hat{A}^\dagger \rangle \langle \hat{B}^\dagger \hat{B} \rangle \quad (14)$$

we have

$$|\langle e^{it(\hat{X}_i - \mu_i)} \rangle - 1| \leq \sigma_i^2 t^2/2 \leq 1/2, \quad (15)$$

where the second inequality is implied by Eq. (5).<sup>1</sup> Hence we may use the principle branch of the logarithm to write

$$\begin{aligned} \Delta(t) &= e^{-\sigma^2 t^2/2} \left| e^{\sigma^2 t^2/2} e^{\sum_{i \in \mathcal{X}} \log \langle e^{it(\hat{X}_i - \mu_i)} \rangle} - 1 \right| \\ &= e^{-\sigma^2 t^2/2} \left| \exp \left( \sum_{i \in \mathcal{X}} \left[ \sigma_i^2 t^2/2 + \log \langle e^{it(\hat{X}_i - \mu_i)} \rangle \right] \right) - 1 \right| \\ &\leq e^{F(t) - \sigma^2 t^2/2} F(t), \end{aligned} \quad (16)$$

where  $F(t) = \sum_i F_i(t)$  with

$$\begin{aligned} F_i(t) &= \left| \frac{\sigma_i^2 t^2}{2} + \log \langle e^{it(\hat{X}_i - \mu_i)} \rangle \right| \\ &\leq \left| \langle e^{it(\hat{X}_i - \mu_i)} \rangle - 1 + \frac{\sigma_i^2 t^2}{2} \right| + \left| \langle e^{it(\hat{X}_i - \mu_i)} \rangle - 1 - \log \langle e^{it(\hat{X}_i - \mu_i)} \rangle \right| \\ &=: G_i(t) + H_i(t), \end{aligned} \quad (17)$$

where we inserted a zero and used the triangle inequality. For the first term we have

$$G_i(t) = \left| \int_0^t dx \int_0^x dy \int_0^y dz \langle (\hat{X}_i - \mu_i)^2 e^{ix(\hat{X}_i - \mu_i)} (\hat{X}_i - \mu_i) \rangle \right| \quad (18)$$

such that by Eq. (14)

$$\begin{aligned} G_i(t) &\leq \int_0^t dx \int_0^x dy \int_0^y dz \sqrt{\langle (\hat{X}_i - \mu_i)^4 \rangle \langle (\hat{X}_i - \mu_i)^2 \rangle} \\ &= \sigma_i \langle (\hat{X}_i - \mu_i)^4 \rangle^{1/2} |t|^3/6. \end{aligned} \quad (19)$$

---

<sup>1</sup>  $\sigma_i |t| > 1$  implies the absurdity  $\frac{3}{5} \sum_{i \in \mathcal{X}} \langle (\hat{X}_i - \mu_i)^4 \rangle^{1/2} \geq \frac{3}{5} \sigma^2 > \sum_{i \in \mathcal{X}} \langle (\hat{X}_i - \mu_i)^4 \rangle^{1/2}$ .

For the second term we use the Mercator series and Eq. (15) to find

$$\begin{aligned}
H_i(t) &\leq \sum_{n=2}^{\infty} \frac{1}{n} |\langle e^{it(\hat{X}_i - \mu_i)} \rangle - 1|^n \leq \sum_{n=2}^{\infty} \frac{1}{n} \left( \frac{\sigma_i^2 t^2}{2} \right)^n \\
&= \frac{\sigma_i^4 t^4}{4} \sum_{n=0}^{\infty} \frac{1}{n+2} \left( \frac{\sigma_i^2 t^2}{2} \right)^n \\
&\leq \frac{\sigma_i^4 t^4}{4} \sum_{n=0}^{\infty} \frac{1}{n+2} 2^{-n} \\
&\leq \frac{\sigma_i^4 t^4}{4}.
\end{aligned} \tag{20}$$

Hence,

$$\begin{aligned}
F(t) &\leq \frac{|t|^3}{6} \sum_{i \in \mathcal{X}} \sigma_i \langle (\hat{X}_i - \mu_i)^4 \rangle^{1/2} + \frac{t^4}{4} \sum_{i \in \mathcal{X}} \sigma_i^4 \\
&\leq \frac{|t|^3}{6} \sum_{i \in \mathcal{X}} \sigma_i \langle (\hat{X}_i - \mu_i)^4 \rangle^{1/2} + \frac{t^4}{4} \sum_{i \in \mathcal{X}} \sigma_i^2 \langle (\hat{X}_i - \mu_i)^4 \rangle^{1/2} \\
&\leq \frac{5}{12} |t|^3 \sum_{i \in \mathcal{X}} \sigma_i \langle (\hat{X}_i - \mu_i)^4 \rangle^{1/2} \\
&\leq \sigma^2 \frac{t^2}{4},
\end{aligned} \tag{21}$$

where we used  $\sigma_i^2 \leq \langle (\hat{X}_i - \mu_i)^4 \rangle^{1/2}$  (which follows from Eq. (14)) to obtain the second line, Eq. (15) to obtain the third line, and Eq. (5) to obtain the last line. Thus

$$\Delta(t) \leq e^{-\sigma^2 t^2/4} F(t) \leq \frac{5}{12} e^{-\sigma^2 t^2/4} |t|^3 \sum_{i \in \mathcal{X}} \sigma_i \langle (\hat{X}_i - \mu_i)^4 \rangle^{1/2} \tag{22}$$

and Eq. (6) follows by using the bound in the third line of Eq. (21).  $\square$