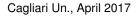
A new class of entropy-power-based uncertainty relations

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in collaboration with J.A. Dunningham





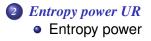




Outline



- Some history
- Why do we need ITUR?
- Rényi's entropy

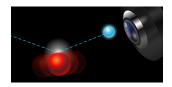






Introduction

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$$egin{array}{ll} \langle \Delta p_i^2
angle_{\psi} \langle \Delta x_j^2
angle_{\psi} &\geq \delta_{ij} rac{\hbar^2}{4} \ & \mathcal{H}(\mathcal{P}^{(1)}) \,+\, \mathcal{H}(\mathcal{P}^{(2)}) \,\geq \, -2\log c \end{array}$$

Quantum-mechanical URs place fundamental limits on the accuracy with which one is able to measure values of different physical quantities. This has profound implications not only on the microscopic but also on the macroscopic level of physical description.







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- Rényi's entropy
- *Entropy power UR*Entropy power
- **3** Applications in QM



Some history Why do we need ITUR? Rényi's entropy

History I

- **1927 Heisenberg's** intuitive derivation of UR $\delta p_x \, \delta x \approx \hbar$
- 1927 Kennard considers as δs as a standard deviation of s
- **1928 Dirac** uses Hausdorff-Young's inequality to prove HUR. δx and δp_x are half-widths of wave packet and its Fourier image
- 1929/30 **Rebertson** and **Schrödinger** reinterpret HUR in terms of statistical ensemble of identically prepared experiments. Both δp and δx are standard deviations. Schwarz inequality in the proof.
- 1945 Mandelstam and Tamm derive time-energy UR
- 1947 Landau derives time-energy UR
- 1968 Carruthers and Nietto angle-angular momentum UR



Some history Why do we need ITUR? Rényi's entropy

History II

1969 Hirschman first Shannon's entr. based UR (weaker than VUR)

- 1971 Synge's three-observable UR
- 1976 Lévy-Leblond improves angle-angular momentum UR
- 1980 Dodonov derives mixed-states UR

 $80-90^\prime \text{s}$ Most standard HUR's are re-derived from Cramér-Rao inequality using Fisher information

1983/84 Deutsch and Białynicky-Birula derive Shannon-entr.-based UR

80 - 90'**s Kraus, Maassen, etc.** derive Shannon-entropy-based UR with sharper bound than Deutsch and B-B

00's Uffink, Montgomery, Abe, etc. derive other non-Shannonian UR



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History III

2006/7s Ozawa's universal error-disturbance relations

2014 Dressel-Nori error-disturbance inequalities

nature

2012 – 15 Violations of Heisenberg's UR measured by number of groups

Experimental demonstration of a universally valid error-disturbance uncertainty relation in spin measurements

Jacqueline Erhart¹, Stephan Sponar¹, Georg Sulyok¹, Gerald Badurek¹, Masanao Ozawa² and Yuji Hasegawa^{1*}

The uncertainty principle generally prohibits simultaneous measurements of certain pairs of observables and forms the basis of indeterminacy in quantum mechanics¹. Heisenberg's original formulation, illustrated by the famous Y-ray microscope, sets a lower bound for the product of the measurement error and the disturbance². Later, the uncertainty relation was reformulated in terms of standard deviations³⁻⁶, where the focus was exclusively on the indeterminacy of predictions, whereas the unavoidable recoil in measuring devices has been ignored⁶. A correct formulation of the error-disturbance uncertainty relation, taking recoil into account, is essential for a deeper understanding of the uncertainty principle, as Heisenberg's original relation is valid only under specific circumstances7-10, A new error-disturbance relation, derived using the theory of general quantum measurements, has been claimed to be universally valid^{n-w}. Here, we report a neutronoptical experiment that records the error of a spin-component measurement as well as the disturbance caused on another spin-component. The results confirm that both error and disturbance obey the new relation but violate the old one in a wide range of an experimental parameter.

as $\sigma(A)^2 = (p(A^2|\phi) - (p(A|\phi)^2)$. Note that a positive definite covariance term can be added to the right-hand side of coquation (2), if squards, as discussed by Schrödinger². For our experimental string, this term vanishes. Robertson's relation (equation (2)) for standard deviations has been confirmed by many different experitions, as expressed in equation (2), has been confirmed. A trade-off and many experiment dimension than been confirmed out¹⁰.

Roberton's relation (equation (2)) has a mathematical basis, but has no immediate implications for limitations on measurements. This relation is naturally understood as limitations on state other hand, the proof of the reciprocal relation for the server (4A) of an A measurement and the distatione of (8) on observable B caused by the measurement, in a general form of Heisenberg's error-disturbance relation

 $\epsilon(A)\eta(B) \ge \frac{1}{2} ||\langle \psi ||[A, B]||\psi \rangle|$

is not straightforward, as Heisenberg's proof² used an unsupported



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PHYSICAL REVIEW A 88, 022110 (2013)

Violation of Heisenberg's error-disturbance uncertainty relation in neutron-spin measurements

Georg Sulyok, ¹ Stephan Sponar, ¹ Jacqueline Erhart, ¹ Cerald Badursk, ¹ Masanao Ozawa, ² and Yuji Hasegawa¹ ¹Institute of Austein and Stahomic Physics, ¹Orean University of Technology, 1020 Niema, Austria ²Gradaute School of Information Science, Nagora University, Chikasa Au, Nagora, Japan (Reverved 3 Jane 2015), publiched 14 August 2013)

In its original formalism, Neisobergy's uscertainty principle dark with the relationship between the error of a quartum measurement and the hereby hadeed burnary to the second s

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PACS number(s): 03.65.Ta, 03.75.Dg, 42.50.Xa, 03.67.-a

I. INTRODUCTION

The uncertainty principle, proposed by Heisenberg [1] in 1927, ranks without doubt among the most famous statements does not hold generally. Thus, his argument did not establish the universal validity of Eq. (1). In 1929, Robertson [19] extended Kennard's relation, Eq. (2), to an arbitrary pair of observables A and B as



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OPEN

SUBJECT AREAS: QUANTUM MECHANICS QUANTUM METROLOGY QUANTUM INFORMATION QUANTUM OPTICS

Experimental violation and reformulation of the Heisenberg's error-disturbance uncertainty relation

So-Young Baek1*, Fumihiro Kaneda1, Masanao Ozawa2 & Keiichi Edamatsu1

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icience, Nagaya University, Nagaya 464-8601, Japan.

The uncertainty principle formulated by Heisenberg in 1927 describes a trade-off between the error of a measurement of one obscrible and the disturbance caused on another complementary obscrable such that their product should be no less than the limit set by Hanck's constant. However, Ozawa in 1988 showed a model of position measurement that breases Betienberg's relation and in 2003 receipted an alternative Pradict of the state Pradict of the state of the Orawa relation for a state of person proven university valid. Here, we report an experimental test of Orawa relation for a state of person proven university of the state of the state of the state of the state relations of the state relations are stated of the state relations of the state of the sta



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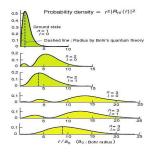
Some history Why do we need ITUR? Rényi's entropy

Why do we need ITUR?

Q: Why do we need information-theoretic UR in the first place?

A: Essence of **VUR** is to put an upper bound to the degree of concentration of two (or more) probability distributions ⇔ impose a lower bound to the associated uncertainties. Usual **VUR** has many limitations*:

• variance as a measure of concentration is a dubious concept when PDF contains more than one peak, e.g., PDF of electron in **H atom**





I. Białynicky-Birula, 1975; D. Deutsch, 1983; H. Maasen, 1988; J. Uffink, 1990

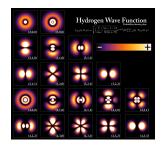
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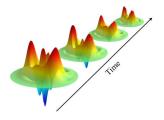
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Why do we need ITUR? Example I

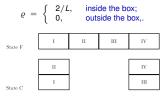
When the distribution is multimodal the variance is often non-intuitive quantifier of uncertainty

Example I: consider two states of a particle in one dimension

First state describes a particle with a *uniform* probability density in a box of total length L, i.e.

 $\varrho = \begin{cases}
1/L, & \text{inside the box;} \\
0, & \text{outside the box,.}
\end{cases}$

Second state describes a particle localized with equal probability densities in two boxes each of length L/4,



states F = flat and C = clustered

Q: In which case, F or C, is the uncertainty in the position greater?

A: Intuition \Rightarrow the uncertainty is greater in the case F. In the case C we know more about the position; the particle is not in the regions II and III. However, $\Delta x_F = L/\sqrt{12}$ while $\Delta x_C = \sqrt{7/4L}/\sqrt{12}$

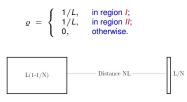


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Why do we need ITUR? Example II

When the distribution is multimodal the variance does not give a sensible measure of uncertainty

Example II: consider a particle in one dimension where the probability density is *constant* in two regions I and II separated by a large distance NL (N is a large number). The region I has the size L(1 - 1/N) and the distant region I has the size L/N. Probability density is:



Example II: $\Delta x \sim L/\sqrt{12}\sqrt{1+12N}$.

NOTE 1: Δx tends to infinity with N even though the probability of finding the particle in the region I tends to 1

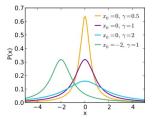
NOTE 2: Problem with the standard deviation It gets very high contributions from distant regions because these enter with a *large weight*: namely, the distance from the mean value.



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Why do we need ITUR?

 variance diverges in many distributions even though such distributions are sharply peaked — heavy-tail distributions, e.g., Lévy, Cauchy, etc.*



Cauchy–Lorentz PDF can be freely concentrated into an arbitrarily small region by changing its scale parameter, while its standard deviation remains very large or even **infinite**.

It is desirable to quantify the inherent quantum unpredictability also in a different way, e.g., in terms of various information measures —entropies

^{*}F. Lillo and R.N. Mantegna, Phys. Rev. Lett. 84 (2000).

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Some history Why do we need ITUR? **Rényi's entropy**

Rényi vs. Shannon — discrete case

Rényi entropy:



A. Rényi (1921-1970)

 $\mathcal{I}_q(\mathcal{P}) = \frac{1}{1-q} \log_2\left(\sum_{x} p^q(x)\right), \quad q > 0$



L.P. Kadanoff (1937 - 2015)

- for q = 1 Rényi entropy equals Shannon's entropy
- is additive, i.e., $\mathcal{I}_q(\mathcal{A}_1 \cup \mathcal{A}_2) = \mathcal{I}_q(\mathcal{A}_1) + \mathcal{I}_q(\mathcal{A}_2|\mathcal{A}_1)$
- $\max_{\mathcal{P}} \mathcal{I}_q(\mathcal{P}) \Rightarrow \mathcal{P} = \{1/n, \dots, 1/n\}$
- second law of "thermodynamics": $\mathcal{I}_q(\mathcal{B}|\mathcal{A}) \leq \mathcal{I}_q(\mathcal{B})$
- it has operational meaning via coding theorem (Campbell, 1965)

A. Rényi, 1970, 1976; L.P. Kadanoff et all, Phys. Rev. Let. 55 (1985) 2798



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Entropy power

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Entropy power

Entropy power — Shannon's case

Let \mathcal{X} is a random vector in \mathbb{R}^{D} with PDF \mathcal{F} . The differential (or continuous) entropy $\mathcal{H}(\mathcal{X})$ of \mathcal{X} is defined as

$$\mathcal{H}(\mathcal{X}) = -\int_{\mathbb{R}^{D}} \mathcal{F}(\boldsymbol{x}) \log_{2} \mathcal{F}(\boldsymbol{x}) \, d\boldsymbol{x}$$

NOTE: Discrete version is nothing but Shannon's entropy which represents an average number of binary questions needed to reveal the value of \mathcal{X} .



C.E. Shannon (1916 - 2001)

NOTE: Strictly $\mathcal{H}(\mathcal{X})$ is **not** a proper entropy but rather an **information gain**^{*}.



* C.E. Shannon 1948, A. Rényi, 1970, 1976

Entropy power

Entropy power — Shannon's case

Entropy power $N(\mathcal{X})$ of \mathcal{X} is a unique number such that^{*}

 $\mathcal{H}(\mathcal{X}) = \mathcal{H}(\mathcal{X}_G)$

where \mathcal{X}_G is a Gaussian random vector with zero mean and variance equal to $N(\mathcal{X})$. So, equivalently

$$\mathcal{H}(\mathcal{X}) = \mathcal{H}\left(\sqrt{\mathcal{N}(\mathcal{X})} \cdot \mathcal{Z}_{G}\right)$$

with Z_G representing a Gaussian random vector with **zero mean** and **unit covariance matrix**. The solution is^{*} (for Shannon measured in *nats*)

$$\mathsf{V}(\mathcal{X}) = rac{1}{2\pi e} \exp\left(rac{2}{D}\mathcal{H}(\mathcal{X})
ight)$$



* C. Shannon 1948, M.H.M. Costa 1985

Entropy power

Entropy power — Rényi's case

Differential Rényi entropy $\mathcal{I}_{p}(\mathcal{X})$ of \mathcal{X} has the form ($p \in \mathbb{R}$):

$$\mathcal{I}_{p}(\mathcal{X}) = \frac{1}{(1-p)} \log_{2} \left(\int_{M} d\boldsymbol{x} \, \mathcal{F}^{p}(\boldsymbol{x}) \right)$$

NOTE: One can check that for $p \to 1$ one has $\mathcal{I}_p(\mathcal{X}) \to \mathcal{H}(\mathcal{X})$.

Definition

The *p*-th **Rényi entropy power** $N_p(\mathcal{X})$ is the solution of the equation

$$\mathcal{I}_{p}(\mathcal{X}) = \mathcal{I}_{p}\left(\sqrt{N_{p}(\mathcal{X})} \cdot \mathcal{Z}_{G}\right)$$

With \mathcal{Z}_G being a Gaussian random vector with **zero mean** and **unit covariance matrix**.



Entropy power

Entropy power — Rényi's case

Theorem

Let \mathcal{X} be a random vector in \mathbb{R}^{D} with PDF $\mathcal{F} \in \ell^{p}(\mathbb{R}^{D})$, where p > 1. The *p*-th **Rényi entropy power** of \mathcal{X} of the form

$$\mathsf{N}_{\!p}(\mathcal{X}) \quad = \quad rac{1}{2\pi}
ho^{-
ho'/
ho} \exp\left(rac{2}{D} \mathcal{I}_{\!p}(\mathcal{F})
ight)$$

(with p' and p being Hölder conjugates) is the **only** admissible class of solutions in the former equation.

(Proof is based on the scaling property $\mathcal{I}_{\rho}(a\mathcal{X}) = \mathcal{I}_{\rho}(\mathcal{X}) + D\log|a|$)

NOTE: In the limit $p \to 1_+$ one has $N_p(\mathcal{X}) \to N(\mathcal{X})$.

NOTE: There are two immediate important observations:

$$N_{
ho}(\sigma \mathcal{X}_G^{\mathbf{1}}) = \sigma^2$$
 and $N_{
ho}(\mathcal{X}_G^{\mathbb{K}}) = |\mathbb{K}|^{1/D}$ $(\mathcal{X}_G^{\mathbb{K}} \sim \mathcal{N}(\mathbf{0}, \mathbb{K}))$



Entropy power

Entropy power uncertainty relations — B-B theorem

Theorem (Beckner–Babenko theorem)

Let
$$f^{(2)}(\mathbf{x}) \equiv \hat{f}^{(1)}(\mathbf{x}) = \int_{\mathbb{R}^D} e^{2\pi i \mathbf{x} \cdot \mathbf{y}} f^{(1)}(\mathbf{y}) d\mathbf{y}$$

then for $p \in [1, 2]$ one has

$$\|\hat{f}\|_{p'} \leq \frac{|p^{D/2}|^{1/p}}{|(p')^{D/2}|^{1/p'}} \|f\|_{p} \quad \text{with} \quad 1/p+1/p'=1$$

NOTE: Inequality is saturated only for Gaussian PDF's.*

Define the square-root density likelihood: $|f(\mathbf{y})| = \sqrt{\mathcal{F}(\mathbf{y})}$ then BBI implies

$$\left(\int_{\mathbb{R}^{D}} [\mathcal{F}^{(2)}(\boldsymbol{y})]^{(1+t)} d\boldsymbol{y}\right)^{1/t} \left(\int_{\mathbb{R}^{D}} [\mathcal{F}^{(1)}(\boldsymbol{y})]^{(1+r)} d\boldsymbol{y}\right)^{1/r} \leq [2(1+t)]^{D} |t/r|^{D/2r}$$

 $(r = p/2 - 1 \text{ and } t = p'/2 - 1 \Rightarrow t = -r/(2r + 1))$



* E.H. Lieb, 1990

Entropy power

Entropy power uncertainty relations

When the negative logarithm is applied on both sides then

$$\mathcal{I}_{1+t}(\mathcal{F}^{(2)}) + \mathcal{I}_{1+r}(\mathcal{F}^{(1)}) \geq \frac{1}{r} \log[2(1+r)] + \frac{1}{t} \log[2(1+t)]$$

This is equivalent to

$$N_{1+t}(\mathcal{F}^{(2)})N_{1+t}(\mathcal{F}^{(1)}) = N_{p/2}(\mathcal{X})N_{q/2}(\mathcal{Y}) \geq \frac{1}{16\pi^2}$$

NOTE 1: When both \mathcal{X} and \mathcal{Y} represent random Gaussian vectors then

$$|\mathbb{K}_{\mathcal{X}}|^{1/D}|\mathbb{K}_{\mathcal{Y}}|^{1/D} = \frac{1}{16\pi^2}$$

NOTE 2: When \mathcal{X} ia random vector with the covariant matrix $(\mathbb{K}_{\mathcal{X}})_{ij}$ then

$$N(\mathcal{X}) \leq |\mathbb{K}_{\mathcal{X}}|^{1/D} \leq \sigma_{\mathcal{X}}^2$$



Entropy power

Enters QM

Consider state vectors that are Fourier transform duals, i.e.

$$egin{aligned} \psi(\mathbf{x}) &= \int_{\mathbb{R}^D} e^{i\mathbf{p}\cdot\mathbf{x}/\hbar} \hat{\psi}(\mathbf{p}) \, rac{d\mathbf{p}}{(2\pi\hbar)^{D/2}} \,, \ & \hat{\psi}(\mathbf{p}) \,= \, \int_{\mathbb{R}^D} e^{-i\mathbf{p}\cdot\mathbf{x}/\hbar} \, \psi(\mathbf{x}) \, rac{d\mathbf{x}}{(2\pi\hbar)^{D/2}} \end{aligned}$$

Comparing with entropy power UR's we have

$$f^{(2)}(\mathbf{x}/\sqrt{2\pi\hbar}) = (2\pi\hbar)^{D/4}\psi(\mathbf{x}),$$

$$f^{(1)}(\mathbf{p}/\sqrt{2\pi\hbar}) = (2\pi\hbar)^{D/4}\hat{\psi}(\mathbf{p})$$

Consequently we can write the associated RE-based UR's as

$$N_{1+t}(|\psi|^2)N_{1+r}(|\hat{\psi}|^2) \geq \frac{\hbar^2}{4}$$



Entropy power

Reconstruction theorem

NOTE: In the case that the PDFs are Gaussian, the whole family of **REPURs** reduces to the single familiar **VUR**

$$\sigma_x^2 \sigma_p^2 = \frac{\hbar^2}{4}$$

Q: In what sense is the entire **tower** of REPURs more general than a single Robertson–Schrödinger **VUR**?

A: _ _ _

Theorem

In order to uniquely reconstruct the underlying PDF for observed QM system one needs to know **all** associated entropy powers *.

NOTE: In cases when the underlying distribution has **all cumulants** finite ⇔ Hamburger–Stiltjes moment problem



* PJ., J. Dunningham and J. Joo, AOP 2014; PRE 2016

Entropy power

Reconstruction theorem

RE can be written as

$$\mathcal{I}_{p}(\mathcal{X}) = \frac{1}{(1-p)} \log_{2} \mathbb{E} \left[2^{(1-p)i_{\mathcal{X}}} \right]$$

Here $i_{\mathcal{X}}(\mathbf{x}) \equiv -\log_2 \mathcal{F}(\mathbf{x})$ is the information in \mathbf{x} .

 \Rightarrow RE can be viewed as a reparametrized version of the cumulant generating function of the variable $i_{\mathcal{X}}(\mathcal{X}) \Rightarrow$ cumulant expansion

$$\rho \mathcal{I}_{1-\rho}(\mathcal{X}) = \log_2 e \sum_{n=1}^{\infty} \frac{\kappa_n(\mathcal{X})}{n!} \left(\frac{p}{\log_2 e}\right)^n$$

 $\kappa_n(\mathcal{X}) \equiv \kappa_n(i_{\mathcal{X}})$ is the *n*-th cumulant of $i_{\mathcal{X}}(\mathcal{X}) \Rightarrow$ reconstruction theorem \Rightarrow

- Gaussian PDF is the only PDF that saturates all REPURs.
- When $N_{1/2}(\mathcal{F}^{(1)})N_{\infty}(\mathcal{F}^{(2)}) = \hbar^2/4$ or $N_{1/2}(\mathcal{F}^{(2)})N_{\infty}(\mathcal{F}^{(1)}) = \hbar^2/4$ then the respective peak-tail parts are Gaussian.
- The closer is \mathcal{F} to Gaussian the smaller neighbourhood of p = 1 is needed in N_p .

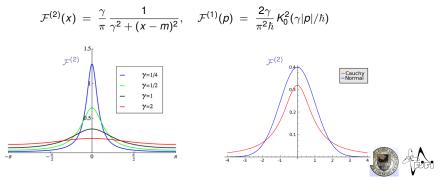


Simple examples I — heavy tailed distributions

• Consider the wave function

$$\psi(x) = \sqrt{\frac{\gamma}{\pi}} \sqrt{\frac{1}{\gamma^2 + (x - m)^2}} \quad \Rightarrow \quad \hat{\psi}(p) = e^{-i\gamma p/\hbar} \sqrt{\frac{2\gamma}{\pi^2 \hbar}} \mathcal{K}_0(\gamma |p|/\hbar) \quad (\text{both} \in \ell^2(\mathbb{R}))$$

The corresponding PDFs read



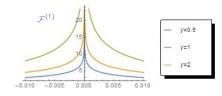
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$$\mathcal{F}^{(2)}(x) = \frac{\gamma}{\pi} \frac{1}{\gamma^2 + (x-m)^2}, \quad \mathcal{F}^{(1)}(p) = \frac{2\gamma}{\pi^2 \hbar} K_0^2(\gamma |p|/\hbar)$$





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Particularly interesting REPURs are

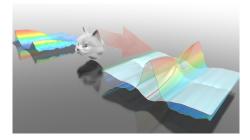
$$N_1(\mathcal{F}^{(1)})N_1(\mathcal{F}^{(2)}) = 0.0052\hbar^2\pi^4 > \frac{\hbar^2}{4}, \quad N_{1/2}(\mathcal{F}^{(1)})N_{\infty}(\mathcal{F}^{(2)}) \stackrel{!}{=} \frac{\hbar^2}{4}$$

cf. with $\langle (\Delta x)^2 \rangle_{\psi} = \infty$ and $\langle (\Delta p)^2 \rangle_{\psi} = \hbar^2 \pi / 16c^2$

 \Rightarrow Schrödinger–Robertson's VUR is completely uninformative

Simple examples II — cat states

Consider a superposition of a vacuum $|0\rangle$ and a squeezed vacuum $|z\rangle$ – cat state



Spectroscopy with cat states of laser light is used in material science (The Cundiff group and Brad Baxley, 2014)



Simple examples II — cat states

Consider a superposition of a vacuum $|0\rangle$ and a squeezed vacuum $|z\rangle$, i.e.

$$|\psi
angle \ = \ \mathcal{N}\left(|0
angle + |z_{\zeta}
angle
ight)$$

where

$$|z_{\zeta}\rangle = \sum_{m=0}^{\infty} (-1)^m \frac{\sqrt{(2m)!}}{2^m m!} \left[\frac{(\tanh \zeta)^m}{\sqrt{\cosh \zeta}} \right] |2m\rangle$$

is a superposition of even number states $|2m\rangle$ with the **squeezing** parameter ζ . \Rightarrow

$$\mathcal{F}^{(2)}(x) = \mathcal{N}^2 \sqrt{\frac{\omega}{\pi\hbar}} \left| \exp\left(-\frac{\omega x^2}{2\hbar}\right) + e^{\zeta/2} \exp\left(-\frac{\omega e^{2\zeta} x^2}{2\hbar}\right) \right|^2$$
$$\mathcal{F}^{(1)}(p) = \mathcal{N}^2 \frac{1}{\sqrt{\pi\hbar\omega}} \left| \exp\left(-\frac{p^2}{2\hbar\omega}\right) + e^{-\zeta/2} \exp\left(-\frac{e^{-2\zeta} p^2}{2\hbar\omega}\right) \right|^2$$

$$\Rightarrow N_{1/2}(\mathcal{F}^{(2)})N_{\infty}(\mathcal{F}^{(1)}) = N_{\infty}(\mathcal{F}^{(2)})N_{1/2}(\mathcal{F}^{(1)}) = \frac{\hbar^2}{4}$$

Simple examples II — cat states

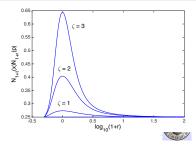
Q: What is so special about the extremal values $\mathcal{I}_{1/2}$ and \mathcal{I}_∞

A: _ _ _

Theorem

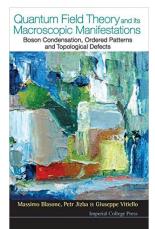
Non-linear nature of RE emphasizes the **more probable** parts of the PDF (typically the **middle parts**) for p > 1 while for p < 1 the **less probable** parts of the PDF (typically the **tails**) are accentuated. In other words, $\mathcal{I}_{1/2}$ mainly carries information on the rare events while \mathcal{I}_{∞} on the common events.

REPUR is saturated at extremal *p*'s because PDFs are **Gaussian** both at wings and at peaks p = x = 0. **REPURs** with different indices do not saturate bound.



Simple examples II — cat states

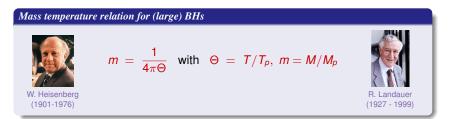
For, some further reading on (Schrödinger) cat states in condensed matter physics, see, e.g.





Little speculation at the end ...

NOTE: There are **information-theoretic** derivations of **black-hole** evap. formula *. The idea is to combine Landauer principle + Heisenberg's UR



Q: What happens with the BH evaporation formula when Heisenberg's **UR** is augmented with other (higher order) **REPURs**?

A: Either IT derivations are **hoax**, or the BH radiation spectrum gets more texture than the simple Planck's BB formula suggests.



^{*} L. Susskind, JHEP 2005; R.J. Adler, GRG 2001; PJ., H. Kleinert and F. Scardigli, PRD 2008 · · ·

Summary

- We have generalized Shannon's ITUR to account for generalized information measures of Rényi. We have seen that in QM systems REPUR's provide more structural information on quantum states (related PDF) than conventional VUR's.
- Our method holds future promise precisely because a large part of the structure of QM is concerned with information.





Summary

- We have generalized Shannon's ITUR to account for generalized information measures of Rényi. We have seen that in QM systems REPUR's provide more structural information on quantum states (related PDF) than conventional VUR's.
- Our method holds future promise precisely because a large part of the structure of QM is concerned with information.
- Entropy-power inequality is instrumental in treatments of QM systems with heavy tailed or multi-peak distributions (Bright–Wigner systems, Schrödinger cat states, etc.)*



^{*} PJ, J.Dunningham and J.Joo, AOP 2015; PRE 2016



"Its all quite elementary, my dear Watson"



- Holmes, *A Study in Scarlet* Arthur C. Doyle



Existent ITUR's and QM

Landau–Pollack ineq. ⇒ Shannon's ITUR for discrete PDF's

 $\mathcal{S}(\mathcal{P}^{(2)}) + \mathcal{S}(\mathcal{P}^{(1)}) \geq -2\log_2 c$

Riesz–Thorin ineq. ⇒ Rényi's ITUR for discrete PDF's

 $\mathcal{I}_{1+t}(\mathcal{P}^{(2)}) + \mathcal{I}_{1+r}(\mathcal{P}^{(1)}) \geq -2\log_2 c^* \quad [-]$

Beckner–Babenko ineq. ⇒ Rényi's ITUR for continuous PDF's

$$\mathcal{N}_{p/2}(\mathcal{X})\mathcal{N}_{q/2}(\mathcal{Y}) \geq rac{1}{16\pi^2} * ~ \car{2}$$

- right-hand sides are independent of the sate $|\psi
 angle$
- often more stringent bound on concentrations of PDF's than VUR's



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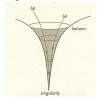
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Bekenstein and Landauer

 $m - \Theta$ relation for (micro) BH sensitively depends on the form of E - x UP \Rightarrow one can easily arrive at non-trivial phenom. consequences

Consider ensemble of unpolarized photons that deliver to MB one single bit of information per particle



In order to ensure that each photon delivers only one bit of information its position uncertainty must be of order $R_S \Rightarrow \Delta X_{\epsilon} \simeq \mu 2R_S$. An extra bit of information added to the micro black hole will increase its energy *at least* by amount ΔE_{ϵ} so that

$$\Delta X_{\epsilon} \Delta E_{\epsilon} \simeq \frac{\hbar c}{2} \left[1 - \frac{\epsilon^2}{2\hbar^2 c^2} (\Delta E_{\epsilon})^2 \right]$$



Bekenstein and Landauer

With Planck's energy

$$\mathcal{E}_{p} = rac{\hbar c}{2\ell_{p}} pprox 0.61 \cdot 10^{19} \, \mathrm{GeV}$$

GUP can be cast to

$$\Delta X_{\epsilon} \simeq \frac{\hbar c}{2E_{\epsilon}} - \frac{a^2 \ell_p E_{\epsilon}}{8\mathcal{E}_p} \qquad (\epsilon = a\ell_p)$$

Using the fact that, $R_S = \ell_p m$, where $m = M/M_p$, we can write

$$2m\mu \simeq rac{\mathcal{E}_p}{\mathcal{E}_\epsilon} - rac{a^2}{8}rac{\mathcal{E}_\epsilon}{\mathcal{E}_p}$$

Landauer principle

Landauer principle:

When a single bit of information is erased the amount of energy dissipated into environment is at least $k_B T \ln 2$ where T is the temperature of erasing environment.

Liberated energy per bit of lost information cannot be grater than E_{ϵ} of the carrier photon

$$\Rightarrow E_{\epsilon} \simeq k_B T$$
.

Defining $T_p = 2\mathcal{E}_p/k_B \approx 10^{32}$ K and $\Theta = T/T_p$, we can rewrite $m - \Theta$ formula as

$$2m = \frac{1}{2\pi\Theta} - 2\pi\zeta^2\Theta$$

where $\zeta = a/(2\sqrt{2}\pi)$ and $\mu = \pi$, in order to agree with Hawking's formula in continuum limit.