# Role of information theoretic uncertainty relations in quantum theory

# Petr Jizba<sup>1,2</sup>

<sup>1</sup>ITP, Freie Universität Berlin

<sup>2</sup>FNSPI, Czech Technical University in Prague

in collaboration with J.A. Dunningham

Groningen, 8th October, 2015



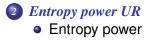




# Outline



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

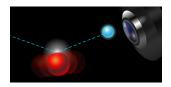






#### Introduction

Entropy power UR Applications in QM Summary Some history Why do we need ITUR? Rényi's entropy



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

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.



W. Heisenberg, Physics and Beyond, 1971



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# 1 Introduction

# Some history

- Why do we need ITUR?
- 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  $\delta s$  as a standard deviation of s
- **1928 Dirac** uses Hausdorff-Young's inequality to prove HUR.  $\delta x$  and  $\delta p_x$  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  $\delta p$  and  $\delta x$  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 Erhart<sup>1</sup>, Stephan Sponar<sup>1</sup>, Georg Sulyok<sup>1</sup>, Gerald Badurek<sup>1</sup>, Masanao Ozawa<sup>2</sup> and Yuji Hasegawa<sup>1</sup>\*

The uncertainty principle generally prohibits simultaneous measurements of certain pairs of observables and forms the basis of indeterminacy in quantum mechanics<sup>1</sup>. Heisenberg's original formulation, illustrated by the famous Y-ray microscope, sets a lower bound for the product of the measurement error and the disturbance<sup>2</sup>. Later, the uncertainty relation was reformulated in terms of standard deviations<sup>3-6</sup>, where the focus was exclusively on the indeterminacy of predictions, whereas the unavoidable recoil in measuring devices has been ignored<sup>6</sup>. 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<sup>n-w</sup>. 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  $\sigma(A)^2 = (p(A^2|\psi) - (p(A|\psi)^2)$ . Note that a positive definite covariance term can be added to the right-hand side of coquation (2), if squarde, as discussed by Schrödinger<sup>2</sup>. 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 out<sup>20</sup>.

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 operavation or limitations on prediction from the pars. On the other hand, the proof of the reciprocal relation for the error e(A) of an A measurement and the distathmene (pR) 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 proof<sup>2</sup> used an unsupported



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PHYSICAL REVIEW A 88, 022110 (2013)

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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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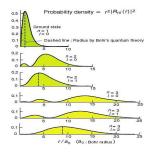
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# Why do we need ITUR?

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

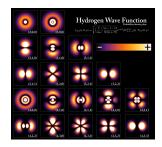
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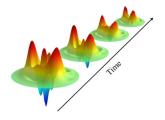
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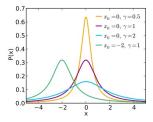
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# Why do we need ITUR?

 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.



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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)

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



- 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)



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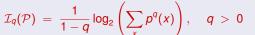
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A. Rényi, 1970, 1976; L.P. Kadanoff et all, Phys. Rev. Let. 55 (1985) 2798

Entropy power

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- *Entropy power UR*Entropy power





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**<sup>\*</sup>.



\* 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<sup>\*</sup>

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

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<sup>\*</sup> (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

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

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

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

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

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



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

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

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** 

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

**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



<sup>\*</sup> PJ., J. Dunningham and J. Joo, AOP 2014

# Simple examples I — heavy tailed distributions

• Consider the wave function

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

• The corresponding PDFs read

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

Particularly interesting REPURs are

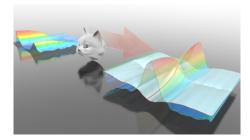
$$N_1(\mathcal{F}^{(1)})N_1(\mathcal{F}^{(2)}) = 0.0052\hbar^2\pi^4 > \frac{\hbar^2}{4}, \quad N_{1/2}(\mathcal{F}^{(1)})N_{\infty}(\mathcal{F}^{(2)}) \stackrel{!}{=} \frac{\hbar^2}{4}$$

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





# Simple examples II — cat states

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

$$|\psi\rangle = \mathcal{N}(|0\rangle + |z\rangle)$$

where

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

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

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

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

# Simple examples II — cat states

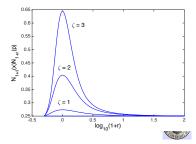
**Q:** What is so special about the extremal values  $\mathcal{I}_{1/2}$  and  $\mathcal{I}_\infty$ 

A: \\_ \\_ \\_

#### Theorem

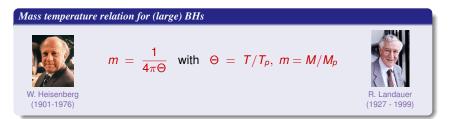
**Non-linear** nature of RE emphasizes the **more probable** parts of the PDF (typically the **middle parts**) for p > 1 while for p < 1 the **less probable** parts of the PDF (typically the **tails**) are accentuated. In other words,  $\mathcal{I}_{1/2}$  mainly carries information on the rare events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



# 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.



<sup>\*</sup> 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 certain QM systems our ITUR's provide more stringent bound on concentrations of involved probability distributions than conventional VUR's.
- Our method holds future promise precisely because a large part of the structure of QM has always concerned with information.





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- We have generalized Shannon's ITUR to account for generalized information measures of Rényi. We have seen that in certain QM systems our ITUR's provide more stringent bound on concentrations of involved probability distributions than conventional VUR's.
- Our method holds future promise precisely because a large part of the structure of QM has always 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





# "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

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

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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