### Information and Context

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#### Information is a fundamental concept in Engineering

- Data science
- Communication theory
- Signal processing
- Compression algorithms

# Why work with information?

Information is also an important concept in

- Physics
  - Statistical mechanics
  - In certain interpretations of quantum mechanics
  - Quantum computation
  - Considered by some physicists as part of the underlying fabric of reality:
    - Wheeler's it-from-bit
    - D'Arianno's derivation of QFT
- Philosophy
  - Language
  - Mind (IIT)
- Social sciences?

# What is information?

- Information is a difficult to define concept.
- In physics and in engineering:
  - Claude Shannon's definition from theory of communications.
  - Requires a probability space  $(\Omega, \mathcal{F}, p)$

### Information and context

- Kolmogorov's axioms for  $(\Omega, \mathcal{F}, p)$ 
  - $p: \mathcal{F} \rightarrow [0, 1]$

• 
$$p(\Omega) = 1$$

- $p(A \cup B) = p(A) + p(B), A \cap B = \emptyset.$
- For some random variables,  $\neg \exists (\Omega, \mathcal{F}, p)$ 
  - Contextuality
- How do we extend Shannon to those situations?





2 Contextuality in Physics and Psychology

#### 3 Measuring Information

- Measuring Contextual Information
- Information and Negative Probabilities

#### 4 Final Remarks





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# From linguistics

- A text or speech is considered contextual if the parts which are related to it are connected to its meaning.
- Consider the statement:
  - "Cheap dates are great" True or False?

# From linguistics

- A text or speech is considered contextual if the parts which are related to it are connected to its meaning.
- Consider the statement:
  - "Cheap dates are great" True or False?
  - Context 1: conversation about social engagements
  - Context 2: discussion about the Phoenix dactylifera fruits.

# Contextuality is about truth values

- A system of propositions is contextual (or exhibit contextuality) if truth values change with context.
- Seems ok in linguistics
  - But what about physics or psychology?
- We need a more measurable definition of contextuality.

# Underpinnings for contextuality

- Start with probability space,  $(\Omega, \mathcal{F}, p)$ 
  - $\bullet \ \Omega \ \text{is a sample space}$
  - $\mathcal{F}$  is a  $\sigma$ -algebra over  $\Omega$
  - p a function  $p:\mathcal{F} \to [0,1]$
- Kolmogorov's Axioms
  - K1.  $p(\Omega) = 1$
  - K2.  $p(A \cup B) = p(A) + p(B)$  for  $A, B \in \mathcal{F}$  and  $A \cap B = \emptyset$ .

# Representations as RV

- Outcomes of measurements can be modeled with RV.
- Random variable  $\mathbf{R}: \Omega \rightarrow O$ 
  - *O* is the set of outcomes (e.g.  $\{1, 2, 3, 4, 5, 6\}$ ,  $\{-1, 1\}$ ,  $\mathbb{R}$ , etc)
  - R is measurable
- Example:  $\Omega = \{(1, 1), (1, 2), \dots, (5, 6), (6, 6)\}, \mathcal{F} = 2^{\Omega}, p(\omega_i) = 1/36$

#### Not all observations can be modeled with RV? Or not all RV have a joint?

Let X, Y, and Z be ±1-valued RV (e.g. *O* = {−1, 1})
 Let

$$E(\mathbf{XY}) = E(\mathbf{XZ}) = E(\mathbf{YZ}) = -1.$$

• If it existed, we would reach a contradiction:

 $\mathbf{X} = 1 \rightarrow \mathbf{Y} = -1 \rightarrow \mathbf{Z} = 1 \rightarrow \mathbf{X} = -1$ 

Contradiction comes context-independency!

### $E(\mathbf{XY}) = E(\mathbf{XZ}) = E(\mathbf{YZ}) = -1.$

- $\bullet$  Contradiction:  $\textbf{X}=1\rightarrow \textbf{Y}=-1\rightarrow \textbf{Z}=1\rightarrow \textbf{X}=-1$
- Observations cannot be contradictory: our model is the problem!
  - Assumption: X in experiment measuring E (XY) is the same as E (XZ)
- Consider:
- Context 1: X and Y
- Context 2: X and Z
- Context 3: Y and Z
- If we index variables (e.g. X<sub>1</sub>, X<sub>2</sub>, Y<sub>1</sub>, Y<sub>3</sub>, Z<sub>2</sub>, and Z<sub>3</sub>), no contradiction.

# No need for perfect correlations

- For three X, Y, and Z that are  $\pm 1$ -valued,  $-1 \leq XY + XZ + YZ$ .
- Therefore

$$-1 \leq E(\mathbf{XY}) + E(\mathbf{XZ}) + E(\mathbf{YZ}) \leq 3.$$

- Necessary and sufficient conditions for the non-contextuality.
- For  $E(\mathbf{X}) = E(\mathbf{Y}) = E(\mathbf{Z}) = 0$ ,  $E(\mathbf{XY}) = E(\mathbf{XZ}) = E(\mathbf{YZ}) = \epsilon$ , and  $E(\mathbf{XYZ}) = \beta$ , non-contextual if within the bounds of a polytope on  $(\epsilon, \beta)$ with vertices:
  - (1,0), (0,-1), (0,1), and (-1/3,0).





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### Properties in physics

- Newtonian physics:
  - state of a system is a point in phase space (e.g.  $(\mathbf{p}, \mathbf{r})$  for a single particle)

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- properties are defined by subsets of the phase space
- properties are non-contextual
- Quantum physics:
  - states are vectors (or operators) in a Hilbert space
  - properties are given by Hermitian operators
  - properties may be contextual

- Binary properties are projection operators
  - for a w and a projector P̂, P̂w = w means w is an eigenvector with eigenvalue 1
    - we say **w** has property *P* associated to  $\hat{P}$ .
  - For a **v** orthogonal to **w**,  $\hat{P}$ **v** = 0; **v** is eigenvector with eigenvalue 0
    - **v** does not have property *P*.
  - Linear combinations  $a\mathbf{w} + b\mathbf{v}$  are not an eigenvector of  $\hat{P}$ 
    - We cannot tell it has property *P* (unless we measure it).

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• Once we measure P, state "collapses" to either  $\mathbf{w}$  or  $\mathbf{v}$ .

### Example of quantum contextual properties

Four dimensional vector space  $\mathbb{R}^4$ .

$$\begin{split} \hat{P}_{0,0,0,1} + \hat{P}_{0,0,1,0} + \hat{P}_{1,1,0,0} + \hat{P}_{1,-1,0,0} &= 1, \\ \hat{P}_{0,0,0,1} + \hat{P}_{0,1,0,0} + \hat{P}_{1,0,1,0} + \hat{P}_{1,0,-1,0} &= 1, \\ \hat{P}_{1,-1,1,-1} + \hat{P}_{1,-1,-1,1} + \hat{P}_{1,1,0,0} + \hat{P}_{0,0,1,1} &= 1, \\ \hat{P}_{1,-1,1,-1} + \hat{P}_{1,1,1,1} + \hat{P}_{1,0,-1,0} + \hat{P}_{0,1,0,-1} &= 1, \\ \hat{P}_{0,0,1,0} + \hat{P}_{0,1,0,0} + \hat{P}_{1,0,0,1} + \hat{P}_{1,0,0,-1} &= 1, \\ \hat{P}_{1,-1,-1,1} + \hat{P}_{1,1,1,1} + \hat{P}_{1,0,0,-1} + \hat{P}_{0,1,-1,0} &= 1, \\ \hat{P}_{1,1,-1,1} + \hat{P}_{1,1,1,-1} + \hat{P}_{1,-1,0,0} + \hat{P}_{0,0,1,1} &= 1, \\ \hat{P}_{1,1,-1,1} + \hat{P}_{-1,1,1,1} + \hat{P}_{1,0,0,1} + \hat{P}_{0,1,0,-1} &= 1, \\ \hat{P}_{1,1,1,-1} + \hat{P}_{-1,1,1,1} + \hat{P}_{1,0,0,1} + \hat{P}_{0,1,-1,0} &= 1. \end{split}$$

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## Another contextual example in $\mathbb{C}^4$

- Alice: A, A'
- Bob: *B*, *B*'
- For non-contextuality

$$-2 \le E(\mathbf{AB}) + E(\mathbf{AB}') + E(\mathbf{A'B}) - E(\mathbf{A'B'}) \le 2$$
$$-2 \le E(\mathbf{AB}) + E(\mathbf{AB'}) - E(\mathbf{A'B}) + E(\mathbf{A'B'}) \le 2$$
$$-2 \le E(\mathbf{AB}) - E(\mathbf{AB'}) + E(\mathbf{A'B}) + E(\mathbf{A'B'}) \le 2$$
$$-2 \le -E(\mathbf{AB}) + E(\mathbf{AB'}) + E(\mathbf{A'B}) + E(\mathbf{A'B'}) \le 2$$

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- Quantum theory predicts  $E(\mathbf{AB}) + E(\mathbf{AB'}) + E(\mathbf{A'B}) - E(\mathbf{A'B'}) < -2$
- Experimentally verified.

 Presents challenges to interpretations of theory and concept of property

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- It is purely quantum
  - Cannot be reproduced classically (not without cost)
- It is a resource in quantum computation

# What about psychology?

- Order effect (explicit contextuality)
  - Clinton is honest and trustworthy? (non-comparative: 50%; comparative 57%)
  - Gore is honest and trustworthy? (non-comparative: 68%; comparative 60%)
- Cervantes and Dzhafarov (hidden contextuality)
  - Gerda, Troll; Beautiful, Unattractive
  - Gerda, Troll; Kind, Evil
  - Snow Queen, Old Finn Woman; Beautiful, Unattractive

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- Snow Queen, Old Finn Woman; Kind, Evil
- But it is not like quantum contextuality!

Measuring Contextual Information nformation and Negative Probabilities

# Outline



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# How to define and measure information?

Shannon's mathematical theory of communication

- Information is related to how surprising a source *s* is.
  - Quantified by  $-\log p(s)$
  - E.g. for four equally surprising outcomes:

• 
$$p(s_i) = 1/4$$

• 
$$I = -\log_2 p(s_1) = 2$$

- Interpret as two bits of information.
- For a source represented by r.v. **X** with output  $\{x_1, x_2, \ldots, x_N\}$ , average amount of information is

$$H = -\sum_{i=1}^{N} p(x_i) \log_2 p(x_i).$$
 (1)

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• *H* is known as Shannon's *entropy*.

# Classical Entropy

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$$H = -\sum_{i=1}^{N} p(x_i) \log_2 p(x_i), \qquad (2)$$

base 2 gives unit of information

- if 2, measure in bits.
- Shannon's coding theorem:
  - it is impossible to compress data such that the average number of bits per symbol is less than *H* without data loss
  - it is possible to create a code whose rate is arbitrarily close to  ${\cal H}$
- Example:
  - ASCII 8 bits
  - 26 english letters require 4.7 binary bits in simple coding (no probability)
  - Morse code allows for 4.14 bits (average)

- A property *O* of a system *S* is a Hermitian operator  $\hat{O}$  on a Hilbert space  $\mathcal{H}$ .
  - Hermitian operators are called *observables*.
  - State is a vector in  $\mathcal{H}$  or by a density operator (a positive semidefinite observable with trace one).
    - vectors are pure states
    - density operator are mixed states.
  - E.g. the normalized vector w ∈ H is a pure state; the density operator is p̂<sub>w</sub> = wω, where ω is the dual to w.

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• Density operators are more general.

#### Expectations from $\hat{\rho}$

- Three dimensional Hilbert space, basis  $\mathbf{e}_i$ , i = 1, 2, 3.
- In vector formalism, the expectation  $\left\langle \hat{P}_i \right\rangle = \left| \hat{P}_i \mathbf{w} \right|^2$ .
- For  $\hat{\rho} = \mathbf{w}\underline{\omega}, \left\langle \hat{P}_i \right\rangle = \left| \hat{P}_i \mathbf{w} \right|^2 = \operatorname{Tr} \left( \hat{\rho} \hat{P}_i \right)$
- If we write  $\hat{\rho}_M = c_1 \hat{\rho}_1 + c_2 \hat{\rho}_2$ , the linearity of the trace gives us that  $\langle \hat{P}_i \rangle = \text{Tr} \left( \hat{\rho}_M \hat{P}_i \right) = c_1 \text{Tr} \left( \hat{\rho}_1 \hat{P}_i \right) + c_2 \text{Tr} \left( \hat{\rho}_2 \hat{P}_i \right)$ , or

$$\left\langle \hat{P}_{i} \right\rangle = c_{1} \left\langle \hat{P}_{i} \right\rangle_{1} + c_{2} \left\langle \hat{P}_{i} \right\rangle_{2}$$

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• For any observable  $\hat{O}$  it follows that  $\left\langle \hat{O} \right\rangle = \mathsf{Tr}\left( \hat{
ho} \hat{O} 
ight).$ 

- The quantum system itself is a source, equivalent to **X** in classical communications theory.
- For a state  $\hat{\rho}$ , its von Neumann entropy [?] is defined as

$$S = -\mathrm{Tr}\left(\hat{\rho}\log\hat{\rho}\right). \tag{3}$$

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- E.g. for three binary properties, represented by basis  $\mathbf{e}_i$ ,  $i = 1, \dots, N$ .
  - Observables  $\hat{P}_i = \mathbf{e}_i \underline{\epsilon}_i$  have binary outcomes 0 or 1.
  - If the state is  $\hat{\rho} = \sum_{i} c_i \mathbf{w}_i \underline{\omega}_i$ , it follows that  $S = -\sum c_i \log c_i$ .

- Despite similarities, Shannon is not the same as von Neumann
  - If  $\hat{\rho}$  was not a proper mixture of non-orthogonal projectors, Shannon would not follow.
- However, there exists a quantum coding theorem.
- For a general orthomodular lattice, natural measure of informational content is von Neumann's entropy
  - Shannon's entropy emerges as the measure for classical-like situations.

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#### Two sources

- Consider **X** and **Y**, valued +1 or -1 with zero expectations.
- Shannon's entropy yields for each, separately,

$$H(\mathbf{X}) = -p_{X} \log p_{X} - p_{\overline{X}} \log p_{\overline{X}}$$
(4)  
$$= -\frac{1}{2} \log \frac{1}{2} - \frac{1}{2} \log \frac{1}{2} =$$
  
$$H(\mathbf{Y}) = 1.$$

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• X and Y seem to have have, combined, two bits of information.

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$$= -\frac{1}{2} \log \frac{1}{2} - \frac{1}{2} \log \frac{1}{2} =$$
  
$$H(\mathbf{Y}) = 1.$$

- X and Y seem to have have, combined, two bits of information.
- True only if uncorrelated.
  - E.g. E(XY) = 1 or E(XY) = -1 would reduce information.

#### How to compute information for two sources?

- Consider **X** and **Y** as a pair: outcomes are all pairs, namely xy,  $x\overline{y}$ ,  $\overline{x}y$ , and  $\overline{xy}$ .
- Setting  $E(XY) = \alpha$ , it follows that

$$p_{xy} + p_{x\overline{y}} + p_{\overline{x}y} + p_{\overline{x}\overline{y}} = 1, \qquad (5)$$

$$p_{xy} + p_{x\overline{y}} - p_{\overline{x}y} - p_{\overline{x}\overline{y}} = 0, \qquad (6)$$

$$p_{xy} - p_{x\overline{y}} + p_{\overline{x}y} - p_{\overline{x}\overline{y}} = 0, \qquad (7)$$

$$p_{xy} - p_{x\overline{y}} - p_{\overline{x}y} + p_{\overline{x}\overline{y}} = \alpha.$$
(8)

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Solution is

$$p_{xy} = p_{\overline{xy}} = \frac{1}{4} (1 + \alpha), \qquad (9)$$
$$p_{\overline{x}y} = p_{\overline{x}y} = \frac{1}{4} (1 - \alpha), \qquad (10)$$

• Entropy:

$$\begin{split} H\left(\mathbf{X},\mathbf{Y}\right) &= -\frac{1}{4}\left(1+\alpha\right)\log\left(\frac{1}{4}\left(1+\alpha\right)\right) \\ &- \frac{1}{4}\left(1-\alpha\right)\log\left(\frac{1}{4}\left(1-\alpha\right)\right). \end{split}$$

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### Information for two sources



Figure: Joint entropy of **X** and **Y** as a function of the correlation  $\alpha$ .

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#### Three variables

Consider ±1-valued random variables X, Y, and Z, zero expectation, correlations E (XY) = E (XZ) = E (YZ) = ε, E (XYZ) = β.

$$p_{xyz} + p_{xy\overline{z}} + p_{x\overline{y}z} + p_{\overline{x}yz} + p_{\overline{x}yz} + p_{\overline{x}y\overline{z}} + p_{\overline{x}y\overline{z}} + p_{\overline{x}y\overline{z}} = 1, \quad (11)$$

$$p_{xyz} + p_{xy\overline{z}} + p_{x\overline{y}z} - p_{\overline{x}yz} - p_{\overline{x}yz} - p_{\overline{x}y\overline{z}} + p_{x\overline{y}\overline{z}} - p_{\overline{x}y\overline{z}} = 0, \quad (12)$$

$$p_{xyz} + p_{xy\overline{z}} - p_{x\overline{y}z} + p_{\overline{x}yz} - p_{\overline{x}yz} + p_{\overline{x}y\overline{z}} - p_{\overline{x}y\overline{z}} - p_{\overline{x}y\overline{z}} = 0, \quad (13)$$

$$p_{xyz} - p_{xy\overline{z}} + p_{x\overline{y}z} + p_{\overline{x}yz} + p_{\overline{x}yz} - p_{\overline{x}y\overline{z}} - p_{\overline{x}y\overline{z}} - p_{\overline{x}y\overline{z}} = 0, \quad (14)$$

$$p_{xyz} + p_{xy\overline{z}} - p_{x\overline{y}z} - p_{\overline{x}yz} + p_{\overline{x}yz} - p_{\overline{x}y\overline{z}} - p_{\overline{x}y\overline{z}} + p_{\overline{x}y\overline{z}} = \epsilon, \quad (15)$$

$$p_{xyz} - p_{xy\overline{z}} + p_{\overline{x}\overline{y}z} - p_{\overline{x}yz} - p_{\overline{x}y\overline{z}} - p_{\overline{x}\overline{y}\overline{z}} + p_{\overline{x}\overline{y}\overline{z}} = \epsilon, \quad (16)$$

$$p_{xyz} - p_{xy\overline{z}} - p_{x\overline{y}\overline{z}} + p_{\overline{x}yz} - p_{\overline{x}y\overline{z}} + p_{\overline{x}\overline{y}\overline{z}} + p_{\overline{x}\overline{y}\overline{z}} = \epsilon, \quad (17)$$

$$p_{xyz} - p_{xy\overline{z}} - p_{x\overline{y}\overline{z}} - p_{\overline{x}yz} + p_{\overline{x}\overline{y}\overline{z}} + p_{\overline{x}\overline{y}\overline{z}} - p_{\overline{x}\overline{y}\overline{z}} = \epsilon, \quad (17)$$

# Solution

Solution is

$$p_{xyz} = \frac{1}{8} (1 + \beta + 3\epsilon), \quad p_{xy\overline{z}} = p_{\overline{x}\overline{y}\overline{z}} = p_{\overline{x}yz} = \frac{1}{8} (1 - \beta - 3\epsilon)$$
(19)  
$$p_{\overline{x}\overline{y}\overline{z}} = p_{\overline{x}\overline{y}\overline{z}} = p_{\overline{x}\overline{y}\overline{z}} = \frac{1}{8} (1 + \beta - 3\epsilon), \quad p_{\overline{x}\overline{y}\overline{z}} = \frac{1}{8} (1 - \beta + 3\epsilon).$$
(20)

- Some p's may be negative for certain values of  $\epsilon$  and  $\beta$ .
  - Non-negative solutions correspond to non-contextual cases

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• Negative solutions are contextual.

#### Non-contextual entropy



Figure: Entropy  $H(\mathbf{X}, \mathbf{Y}, \mathbf{Z})$  for  $E(\mathbf{XY}) = E(\mathbf{XZ}) = E(\mathbf{YZ}) = \epsilon$  and  $E(\mathbf{XYZ}) = 0$  as a function of  $\epsilon$ . The maximum of 3 bits occurs when  $\epsilon = 0$ .

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# What about contextual sources?

- Outside of  $-1/3 \le \epsilon \le 1$ , Shannon's entropy is not defined.
- But **X**, **Y**, and **Z** have informational content outside of the probability polytope.

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• How do we measure information outside of the polytope?

#### What about contextual sources?

- Outside of  $-1/3 \le \epsilon \le 1$ , Shannon's entropy is not defined.
- But X, Y, and Z have informational content outside of the probability polytope.
- How do we measure information outside of the polytope?
- Going back to  $\epsilon = -1$ .
  - If we know X we know Y and Z, so 1 bit?
  - No. X in the context of Y cannot be the same as in the context of Z, so information content is more than one bit.

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• Variables are  $X_Y$ ,  $X_Z$ ,  $Y_X$ ,  $Y_Z$ ,  $Z_X$ , and  $Z_Y$ 

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- "Negative probabilities" were introduced by Dirac.
- Name *negative probabilities* may be misleading:
  - some probabilities of atomic events may be negative, but not all

- observable probabilities may not be negative.
- Perhaps better term is "signed probability".

### **Extending Shannon**

- Let  $(\Omega, \mathcal{F}, p^*)$  be a negative probability space.
- The extended entropy  $S_{NP}$  for  $p^*$  is

$$S_{NP} = -\sum_{\omega_i \in \Omega} |p^*(\omega_i)| \log_2 |p^*(\omega_i)|.$$
(21)

- The absolute value comes from using the expression  $A = \sum_{\omega_i \in \Omega} |p^*(\omega_i)|$  as a measure of how contextual a probabilistic system is [?].
  - If A = 1, no contextuality, since  $A = \sum_{\omega_i \in \Omega} p^*(\omega_i)$ , which implies that  $p^*$ 's are non-negative.
  - If A > 1, then the system is contextual.
- S<sub>NP</sub> is Shannon when p<sup>\*</sup> is non-negative.

# $S_{NP}$ for symmetric case



Figure: Surface plot of  $S_{NP}$  as a function of  $\epsilon$  (horizontal axis) and  $\beta$  (vertical axis). Lighter regions correspond to less entropy, whereas darker regions to more entropy.

# Cross section of $S_{NP}$



Figure: Cross section of  $S_{NP}$  as a function of  $\epsilon$  and for  $\beta = 0$ . We see a local maxima at  $\epsilon = 0$  and a global maxima at  $\epsilon = -1$ , where  $S_{NP} = 4$ .

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#### • Contextual informational in key in

- physics
- engineering
- Shannon and von Neumann provide measures
  - $\bullet\,$  von Neumann deals with contextual information, but requires  ${\cal H}$

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- $S_{NP}$  generalizes Shannon for some non-quantum contextual states
  - consistent with von Neumann's quantum inequalities

"I not only use all the brains I have, but all I can borrow" - Woodrow Wilson

 Patrick Suppes (Stanford); Gary Oas (Stanford); Pawel Kurzynski (Adam Mickiewicz); Federico Holik (La Plata); Carlos Montemayor (SFSU); John Perry (Stanford); Stephan Hartmann (LMU); Francisco Doria (UFRJ); Gregory Chaitin (UFRJ)

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# Thank you!!

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