Quantum Dynamical Entropies and Complexity

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Coarse-graining and Information Dynamical Entropy and Stability Statistical Randomness Information source compressibility

Classical Dynamical Entropy

Measure theoretic setting

- \mathcal{X} measure space with a σ -algebra Σ of measurable subsets.
- $\mu : \mathcal{X} \mapsto [0,1]$: probability measure or reference state; $\mu(\mathcal{X}) = 1$.
- $T : \mathcal{X} \mapsto \mathcal{X}$: invertible dynamical map such that $\mu \circ T = \mu$: equilibrium state.
- $\Pi = \{P_i\}_{i=1}^d$: finite, measurable partition,

$$P_i \in \Sigma \ , \ P_i \cap P_j = \delta_{ij} P_i \ , \ \bigcup_{i=1}^d P_i = \mathcal{X} \ .$$

- $\Pi_k = T^{-k}(\Pi) = \{T^{-k}(P_i)\}_{i=1}^d$ partition at time t = k;
- $\Pi^{(n)} = \{P_{\mathbf{i}^{(n)}}\}, \mathbf{i}^{(n)} \in \{1, \dots, d\}^n$: refined partition up to t = n 1;

$$P_{\mathbf{i}^{(n)}} = P_{i_0} \cap T^{-1}(P_{i_1}) \cap \dots \cap T^{-n+1}(P_{i_{n-1}})$$

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Dynamical Information Gain

Shannon entropy, information and dynamics: learning from history

- $\Pi_k = \{T^{-k}(P_i)\}_{i=1}^d$: coarse-grained information about $x \in \mathcal{X}$ at t = k
- Uncertainty about $x \in \mathcal{X}$ given μ : Shannon entropy,

$$H_{\mu}(\Pi_k) = H_{\mu}(\Pi) = -\sum_{i=1}^d \mu(P_i) \log \mu(P_i) \ .$$

- $x \in P_{\mathbf{i}^{(n)}}$ implies $T^k x \in P_{i_k}$ for $0 \le k \le n-1$
- Uncertainty about the coarse-grained trajectory $\{T^{-k}(P_{i_k})\}_{k=0}^{n-1}$:

$$H_{\mu}(\Pi^{(n)}) = -\sum_{\mathbf{i}^{(n)}} \mu(P_{\mathbf{i}^{(n)}}) \log \mu(P_{\mathbf{i}^{(n)}}) \; .$$

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Maximal Dynamical Information Gain

Maximal coarse-grained time-averaged information

• Average information gain from one step to the next one:

$$\begin{split} h_{\mu}(T,\Pi) &= \lim_{n \to \infty} \frac{1}{n} \sum_{k=0}^{n-1} \left(H_{\mu}(\Pi^{(k+1)}) - H_{\mu}(\Pi^{(k)}) \right) \\ &= \lim_{n \to \infty} \frac{1}{n} H_{\mu}(\pi^{(n)}) \; . \end{split}$$

- Kolmogorov-Sinai dynamical entropy: $h_{\mu}(T) = \sup_{\Pi} h_{\mu}(T, \Pi)$.
- Kolmogorov Theorem: if Π is a generating partition,

$$\bigvee_{n\in\mathbb{Z}} \mathcal{T}^n(\Pi) = \Sigma ext{ then } \quad h_\mu(\mathcal{T}) = h_\mu(\mathcal{T},\Pi) \;.$$

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Baker's map

• Dynamical triple: $\mathcal{X} = \mathbb{T}^2 = \{(x, y) \text{ mod}1\}, d\mu(x, y) = dsdy$

$$B(x,y) = \begin{cases} (2x,y/2) & 0 \le x < 1/2 \\ 2x - 1, (1+y)/2 & 1/2 \le x < 1 \end{cases}$$
$$B^{-1}(x,y) = \begin{cases} (x/2,2y) & 0 \le y < 1/2 \\ (1+x)/2, 2y - 1 & 1/2 \le y < 1 \end{cases}$$

• Generating partition: $\Pi = \{P_0, P_1\},\$

$$P_0 = \{(x, y) : 0 \le x < 1/2\}, \ P_1 = \{(x, y) : 1/2 \le x < 1\}$$

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Baker's map

• Refined partitions: $\Pi^{(n)} = \{P_{\mathbf{i}^{(n)}}\}$

$$P_{\mathbf{i}^{(n)}} = \left\{ (x, y) : x_1^{(n)} \le x \le x_2^{(n)}, x_2^{(n)} - x_1^{(n)} = 1/2^n \right\}$$

• Entropy rate and KS-entropy:

$$\mu(P_{\mathbf{i}^{(n)}}) = 1/2^n \implies h_{\mu}(B, \Pi) = \lim_{n} \frac{1}{n} H_{\mu}(\Pi^{(n)} = \log 2 = h_{\mu}(B))$$

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Baker's map as a one-dim left-shift

• Write $x = \sum_{k=1}^{\infty} x_k 2^{-k}$, $y = \sum_{j=1}^{\infty} y_j 2^{-j}$ and order (x, y) as a bi-infinite string:

$$(x, y) = \cdots \mathbf{z} = \{z_{\ell}\}_{\ell=-\infty}^{+\infty} , \ z_{-n} = y_n , \ z_n = x_{n+1}$$

• *B* acts as a left-shift:

$$(B\mathbf{z})_n = z_{n+1}$$
, $B^n \mathbf{z} = 2^n \mathbf{z} \mod 1$

• $h_{\mu}(B) = \log 2$ is the Lyapounov exponent of the Baker map

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Dynamical Instability

Lyapounov exponents

• Exponential increase of initial small errors:

$$d(x_1, x_2) = \delta \mapsto d(T^n x_1, T^n x_2) \simeq \mathrm{e}^{\lambda t} d(x_1, x_2) \; .$$

• Positive Lyapounov exponents:

$$\lambda(T,x) = \lim_{n \to +\infty} \frac{1}{n} \lim_{\delta \to 0} \log \frac{d(T^n x, T^n(x+\delta))}{d(x, x+\delta)}$$

- $\lambda(T, x) = \lambda(T)$: independence of x for ergodic systems
- Pesin's Theorem: for ergodic systems

$$h_\mu(T) = \sum_i \lambda_i^+(T) \; .$$

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Degrees of Statistical Randomness

• ergodicity: time-averaged correlation functions factorize,

$$\lim_{T\to+\infty}\frac{1}{T}\sum_{n=0}^{T-1}\mu(fg\circ T^n)=\mu(f)\mu(g)\ ,\quad \mu(f)=\int_{\mathcal{X}}\mathrm{d}\mu(x)\,f(x)\ .$$

- \bullet The only constants of the motion of ergodic are constant on ${\cal X}$
- Convex combinations

$$\mu = \sum_{i} \lambda_{i} \mu_{i} \ , \lambda_{i} \ge 0 \ , \sum_{i} \lambda_{i} = 1$$

into *T*-invariant states, $\mu_i \circ T = \mu_i$, impossible for ergodic states.

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Mixing and K-mixing

Mixing and Relaxation

• mixing: correlation functions factorize,

$$\lim_{n\to+\infty}\mu(fg\circ T^n)=\mu(f)\mu(g).$$

• Mixing implies ergodicity and tendency to equilibrium: $\nu(g) = \int_{\mathcal{X}} d\mu(x) f_{\nu}(x) g(x)$ perturbation of μ ,

$$\begin{split} \nu(g \circ T^n) &= \int_{\mathcal{X}} \mathrm{d}\mu(x) \, f_{\nu}(x) \, g(T^n x) \longrightarrow \mu(f_{\nu})\mu(g) \\ &= \nu(\mathcal{X})\mu(g) = \mu(g) \; . \end{split}$$

• K-mixing: total memory loss on long times,

$$\lim_n h_\mu(T^n,\Pi) = H_\mu(\Pi) \Longleftrightarrow h_\mu(T,\Pi) > 0 \,\,\forall \,\,\Pi \,\,.$$

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Classical Information Sources

• Classical Information Sources emit finite length binary strings

$$\mathbf{i}^{(n)} \in \Omega^{(n)} = \{\mathbf{i}^{(n)}\} \ , \ \mathbf{i}^{(n)} = i_1 i_2 \dots i_n \ , \ i_j = 0, 1$$

• with probabilities: $\pi^{(n)} = \{p^{(n)}(\mathbf{i}^{(n)})\}$ such that

$$\sum_{i_n} p^{(n)}(i_1i_2\cdots i_n) = p^{(n-1)}(i_1i_2\cdots i_{n-1})$$
 compatibility

- Set of binary sequences: $\Omega \ni \mathbf{i} = \{i_j\}_{j=1}^{\infty}$
- Emission of letters, one after the other, left shift on Ω : $\sigma(\mathbf{i})_n = \mathbf{i}_{n+1}$
- Stationary information source:

$$\sum_{i_1} p^{(n)}(i_1i_2\cdots i_n) = p^{(n-1)}(i_2i_3\cdots i_n)$$
 stationarity

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Asymptotic Equipartition Theorem

• Kolmogorov-Sinai entropy of a stationary source:

$$h_{\pi}(\sigma) = \lim_{n \to +\infty} rac{1}{n} \left(-\sum_{\mathbf{i}^{(n)}} p^{(n)}(\mathbf{i}^{(n)}) \log p^{(n)}(\mathbf{i}^{(n)})
ight)$$

• Typical subset: given $\epsilon > 0$, $\exists N_{\epsilon}$ such that, for $n \ge N_{\epsilon}$,

$$A_n^{\epsilon} = \left\{ \mathbf{i}^{(n)} : 2^{-n(h_{\pi}(\sigma)+\epsilon)} \le p(\mathbf{i}^{(n)}) \le 2^{-n(h_{\pi}(\sigma)-\epsilon)} \right\}$$
$$\pi(A_n^{\epsilon}) \ge 1 - \epsilon$$
$$(1 - \epsilon)2^{n(h_{\pi}(\sigma)-\epsilon)} \le \operatorname{card}(A_n^{\epsilon}) \le 2^{n(h_{\pi}(\sigma)+\epsilon)}$$

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Asymptotic Equipartition and compression

• Untypical subset: $\pi\left(\left(A_{n}^{\epsilon}\right)^{c}\right) \leq \epsilon$,

$$\left(A_{n}^{\epsilon}\right)^{c} = \left\{\mathbf{i}^{(n)}: \left|-\frac{1}{n}\log p^{(n)}(\mathbf{i}^{(n)}) - h_{\pi}(\sigma)\right| > \epsilon\right\}$$

- Encoding A_n^{ϵ} requires $h_{\mu}(\pi)$ bits per letter
- sending all $(A_n^{\epsilon})^c$ into a same symbol: error with probability $\leq \epsilon$

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Algorithmic Complexity

Individual Randomness versus Statistical Randomness

Stationary information sources (of digit strings)

- Source emitting bit-strings $\mathbf{i}^{(n)} = i_1 i_2 \cdots i_n \in \Omega_2^n = \{0, 1\}^n$
- with probabilities $\pi^{(n)} = \{p(\mathbf{i}^{(n)})\}_{\mathbf{i}^{(n)} \in \Omega_2^{(n)}}$
- Shannon entropy

$$H(\pi^{(n)}) = -\sum_{\mathbf{i}^{(n)} \in \Omega_2^{(n)}} p(\mathbf{i}^{(n)}) \log p(\mathbf{i}^{(n)})$$

a measure of the average randomness of the emitted strings

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Kolmogorov Complexity

Individual randomness: How difficult is to describe a digit string?

- binary program: any string $p \in \Omega_2^{(m)}$
- length of a binary program: $\ell(p) = m$
- Universal, prefix Turing machine \mathcal{U} :

$$\Omega_2^{(m)}
i p \mapsto \mathcal{U}(p) = \mathbf{i}^{(n)} \in \Omega_2^{(n)}$$

Kolmogorov complexity of i⁽ⁿ⁾ ∈ Ω₂⁽ⁿ⁾: length of the shortest binary program that, run by U, outputs i⁽ⁿ⁾ and halts:

$$K(\mathbf{i}^{(n)}) = \min\left\{\ell(p) : \mathcal{U}(p) = \mathbf{i}^{(n)}\right\}$$

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Turing Machines

Classical Turing Machines ${\cal U}$

- (0,1) on Input and Output Cells: configuration space t
- Read-Write Head Position: position space h
- Internal Control Unit: internal state space q
- *U* reversibly operates on input *p* and halts with *U(p)* written on the output tape
- universal TM: capable to act as any other TM
- Prefix UTM: if $\mathcal{U}[p]$ halts, $\mathcal{U}[pq]$ does not halt

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Regular bit-strings

Regular bit strings can be shortly described

•
$$i^{(1000)} = \underbrace{11111111111\cdots111}_{1000 \ bits}$$

• short program:

Write 1 1000 times
$$\implies \mathcal{K}(i^{(1000)}) = \log 1000 + C$$

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Random Bit-Strings

Random bit strings cannot be compressed

•
$$i^{(1000)} = \underbrace{1001010101000\dots100}_{1000 \ bits}$$

• short program:

Write
$$1, 0, 0, 1, 0, 1... \implies K(i^{(1000)}) = 1000 + C$$

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Counting Argument

• there are 2^n binary strings of length n, but only

$$\sum_{j=0}^{\ell-1} 2^j = 2^\ell - 1 < 2^\ell$$

binary programs p of length $\ell(p) < \ell$

• Thus, there are more than

$$2^{n} - 2^{\ell} = 2^{n} \left(1 - \frac{1}{2^{n-\ell}} \right)$$

Binary strings with complexity larger than ℓ .

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Complexity of a length n binary string with k ones

Trial program p: specify k and the string among $\binom{n}{k}$ strings:

$$\begin{aligned} \mathcal{K}(\mathbf{i}^{(n)}) &\leq \ell(p) = C + \log_2 k + \log_2 \binom{n}{k} \\ &\leq C + \log_2 k + n H_2(k/n, 1 - k/n) \\ \mathcal{H}(k/n) &= -\frac{k}{n} \log_2 \frac{k}{n} - (1 - \frac{k}{n}) \log_2 (1 - \frac{k}{n}) \end{aligned}$$

Indeed,

$$\frac{1}{n+1} 2^{n H_2(k/n)} \le \binom{n}{k} \le 2^{n H_2(k/n)}$$

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Kolmogorov complexity and entropy rate

Brudno's theorem

- ergodic sources: $\mathbf{i}^{(n)} \in \Omega_2^{(n)} \longmapsto \mathbf{i} \in \Omega_2 = \{0, 1\}^{\infty}$, $\pi^{(n)} \longmapsto \pi$,
- entropy rate

$$h(\pi) = \lim_{n \to \infty} \frac{1}{n} H(\pi^{(n)})$$

• complexity per symbol:

$$k(\mathbf{i}) = \lim_{n \to +\infty} \frac{1}{n} \, \mathcal{K}(\mathbf{i}^{(n)})$$

• for ergodic sources $k(\mathbf{i}) = h(\pi)$ $\pi - a.e.$

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Brudno's theorem at its simplest

- Bernoulli source: $p(i_0 i_1 \cdots i_{n-1}) = \prod_{j=0}^{n-1} p(i_j), \ p(0) = p,$ p(1) = 1 - p
- Law of large numbers: $n \ge N_{\epsilon,\delta}$, $N_1(\mathbf{i}^{(n)}) = \frac{1}{n} \sum_{j=0}^{n-1} i_j$

$$\mathsf{Prob}\left\{\left|H_{2}(p)-H_{2}\left(N_{1}(n)\right)\right|\geq\epsilon\right\}\leq\epsilon$$

• Counting argument:

$$\mathsf{Card}\left\{\mathbf{i}^{(n)}:\frac{1}{n}\mathsf{K}(\mathbf{i}^{(n)})\leq \mathsf{H}_2(p)-\delta\right\}\leq 2^{n(\mathsf{H}_2(p)-\delta)}$$

• Typical subset:

$$\operatorname{Card}(A^n_{\epsilon}) \leq 2^{n(H_2(p)+\epsilon)}$$

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Brudno's theorem at its simplest Prob
$$\left\{ \left| \frac{1}{n} \mathcal{K}(\mathbf{i}^{(n)}) - \mathcal{H}_2(p) \right| \ge \delta \right\} \le \delta$$

• From law of large numbers and $\mathcal{K}(\mathbf{i}^{(n)}) \leq C + \log_2 N_1(\mathbf{i}^{(n)}) + n H_2(N_1(\mathbf{i}^{(n)})),$

$$\operatorname{Prob}\left\{\frac{1}{n}K(\mathbf{i}^{(n)}) \geq H_2(p) + \epsilon\right\} \leq \epsilon$$

• From counting argument and typicality with $\delta > \epsilon$:

$$\begin{aligned} &\operatorname{Prob}\left\{\frac{1}{n}\mathcal{K}(\mathbf{i}^{(n)}) \leq H_2(p) - \epsilon\right\} \leq \operatorname{Prob}\left((A_n^{\epsilon})^c\right) \\ &+\operatorname{Prob}\left\{\mathbf{i}^{(n)} \in A_n^{\epsilon} : \frac{1}{n}\mathcal{K}(\mathbf{i}^{(n)}) \leq H_2(p) - \epsilon\right\} \\ &\leq \epsilon + 2^{n(H_2(p) + \epsilon)}2^{-n(H_2(p) - \delta)} \end{aligned}$$

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Algorithmic complexity of point trajectories

- ergodic dynamical system $(\mathcal{X}, \Theta, \mu)$
- coarse-graining of \mathcal{X} by finite partitions $\Pi = \{P_i\}_{i=1}^p$
- Coarse-grained trajectory of $x \in \mathcal{X}$: $\mathbf{i}_{\Pi}^{n}(x) = i_{0}(x)i_{1}(x)\cdots i_{n-1}(x)$:

$$x \in P_{i_0}$$
, $Tx \in P_{i_1}$, ... $T^{n-1}x \in P_{i_{n-1}}$.

• *x*-dynamical string: $\mathbf{i}_{\Pi}(x) = \lim_{n} \mathbf{i}_{\Pi}^{n}(x)$

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Dynamical Brudno's theorem

• Complexity rate of coarse-grained trajectories:

$$k(\Theta, \mathbf{i}_{\Pi}(x)) = \lim_{n \to +\infty} \frac{1}{n} \mathcal{K}(\mathbf{i}_{\Pi}^{n}(x))$$

• maximal complexity rate of coarse-grained trajectories:

$$k_{\Theta}(x) = \sup_{\Pi} k_{\Pi}(\mathbf{i}_{\Pi}(x))$$

Brudno's theorem

ergodic $(\mathcal{X}, \Theta, \mu)$

$$k_{\Theta}(\mathbf{i}) = h_{\mu}(\Theta) \qquad \mu - a.e.$$

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Universal probability

• a-priori probability of $\mathbf{i}^{(n)} \in \Omega_2^{(n)}$:

$$\mathbb{P}(\mathbf{i}^{(n)}) := \sum_{p \ : \ \mathcal{U}[p] = \mathbf{i}^{(n)}} 2^{-\ell(p)} \ , \ \ \mathcal{U} \ \mathsf{prefix} \ \mathsf{UTM}$$

• universality of \mathbb{P} : for all semi-computable semi-measures μ :

$$\exists C_{\mu} > 0 : C_{\mu}\mu(\mathbf{i}^{(n)}) \leq \mathbb{P}(\mathbf{i}^{(n)})$$

• universal probability and complexity:

$$\mathcal{K}(\mathbf{i}^{(n)}) = -\log \mathbb{P}(\mathbf{i}^{(n)}) + C$$

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Why $\sum_{\mathbf{i}^{(n)}} \mathbb{P}(\mathbf{i}^{(n)}) \leq 1$?

- Code for $\mathbf{i}^{(n)} \in \Omega^{(n)}$: $W : \mathbf{i}^{(n)} \mapsto W(\mathbf{i}^{(n)})$
- List the code-words as $\{w_j\}_{j=1}^{2^n}$
- Prefix condition: w_j code words for $\mathbf{i}^{(n)} \in \Omega^{(n)}$ then $w_j w_k$ not code word
- Counting argument: ℓ_j length of w_j ,

 $\begin{array}{rcl} N_j &=& \text{number of code words with length} \quad \ell_j \\ N_1 &\leq& 2^{\ell_1} \ , \ N_2 \leq 2^{\ell_2} \ , \quad N_k \leq 2^{\ell_k} - 2^{\ell_k - \ell_{k-1}} - \cdots - 2^{\ell_k - \ell_1} \end{array}$

• Kraft's inequality:

$$\sum_{j=1}^{M} N_j 2^{-\ell_j} = \sum_{\mathbf{i}^{(n)} \in \Omega^{(n)}} 2^{-\ell(W(\mathbf{i}^{(n)}))} \le 1$$

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Kolmogorov complexity and entropy

- $\pi^{(n)} = \{p(i^{(n)})\}$ a computable distribution on $\Omega_2^n = \{0, 1\}^n$:
- From universality, $2^{-\kappa(\pi)}\pi \leq \mathbb{P}$, and $-\log \mathbb{P}(\mathbf{i}^{(n)}) + C = \kappa(\mathbf{i}^{(n)})$,

 $\langle K \rangle_{\pi} \leq H(\pi) + C + K(\pi)$

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Shannon entropy=

Kolmogorov complexity

• Convexity: $x(\log x - \log y) \ge x - y$,

$$0 \leq \sum_{\mathbf{i}^{(n)}} p(\mathbf{i}^{(n)}) \left(\log p(\mathbf{i}^{(n)}) - \log \frac{2^{-K(\mathbf{i}^{(n)})}}{\sum_{\mathbf{i}^{(n)}} 2^{-K(\mathbf{i}^{(n)})}} \right)$$
$$= -H(\pi) + \langle K \rangle_{\pi} + \sum_{\mathbf{i}^{(n)}} p(\mathbf{i}^{(n)}) \log \sum_{\mathbf{i}^{(n)}} 2^{-K(\mathbf{i}^{(n)})}$$
$$\leq -H(\pi) + \langle K \rangle_{\pi}$$

• Shannon entropy=average Kolmogorov complexity:

$${\mathcal H}(\pi^{(n)})\simeq \sum_{{\mathbf i}^{(n)}\in\Omega_2^n}\,{\mathcal p}({\mathbf i}^{(n)})\,{\mathcal K}({\mathbf i}^{(n)}):=\langle{\mathcal K}
angle_\pi$$

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Quantum Sources

Quantum Sources

Quantum Spin Chains

- Local algebras: $\mathcal{M}_2^{(n)} = (M_2)^{\otimes n}$
- local states: $M_2^{(n)} \ni \rho^{(n)}$, $\operatorname{Tr}_1 \rho^{(n)} = \operatorname{Tr}_n \rho^{(n)} = \rho^{(n-1)}$
- translation-invariant state on $\mathcal{M} := \mathcal{M}_2^{\otimes \infty}: \ \rho^{(n)} \longmapsto \omega$
- von Neumann Entropy Rate

$$egin{array}{rcl} S(
ho^{(n)}) &=& -\operatorname{Tr}
ho^{(n)} \log
ho^{(n)} \ s(\omega) &=& \lim_{n o +\infty} rac{1}{n} S(
ho^{(n)}) \end{array}$$

Connes-Narnhofer-Thirring Entropy Alicki-Fannes entropy

Quantum-Classical matching

● Classical Partitions→Quantum Subalgebras:

$$\Pi = \{P_i\}_{i=1}^d \longrightarrow A_{\Pi} = \{\hat{P}_i\}_{i=1}^d , \ \hat{P}_i\hat{P}_j = \delta_{ij}\hat{P}_i$$

● Shannon entropy of a coarse grained trajectory → von Neumann entropy of the generated algebra?

$$H_{\mu}(\bigcap_{j=0}^{n-1}\Theta^{j}(\Pi))\longrightarrow S(\rho \upharpoonright \bigvee_{j=0}^{n-1}\Theta^{j}(A_{\Pi})) ?$$

NO

• Subadditivity does not hold:

$$S(\rho \upharpoonright A \lor \Theta(A)) \nleq S(\rho \upharpoonright A) + S(\rho \upharpoonright \Theta(A))$$

• Already $A_{\pi} \vee \Theta(A)$ can become infinite dimensional

Connes-Narnhofer-Thirring Entropy Alicki-Fannes entropy

Relative entropy: $S(\rho_1; \rho_2) = \text{Tr}\left(\rho_1(\log \rho_1 - \log \rho_2)\right)$

- Positivity: $S(\rho_1; \rho_2) \ge 0$, $S(\rho_1; \rho_2) = 0 \iff \rho_1 = \rho_2$.
- Joint convexity:

$$\mathcal{S}\left(\sum_{j}\lambda_{j}
ho_{1j};\sum_{j}\lambda_{j}
ho_{2j}
ight)\leq\sum_{j}\lambda_{j}\,\mathcal{S}(
ho_{1j};
ho_{2j})\,,$$

• Monotonicity under completely positive unital maps $\gamma : \mathcal{M} \mapsto \mathcal{M}$:

$$\gamma^{\mathsf{T}}[\rho] = \rho \circ \gamma \;, \qquad \mathcal{S}(\gamma^{\mathsf{T}}[\rho_1]; \gamma^{\mathsf{T}}[\rho_2]) \leq \mathcal{S}(\rho_1; \rho_2) \;.$$

Connes-Narnhofer-Thirring Entropy Alicki-Fannes entropy

Convex Decompositions

Entropy of a subalgebra

• Decompositions of ω :

$$\omega = \sum_j \lambda_j \hat{\omega}_j \;,\; \lambda_j \geq 0\;,\; \sum_j \lambda_j = 1\;, \hat{\omega}_j(1) = 1\;.$$

• Restriction of a state to a subalgebra: $\omega \upharpoonright A \subset \mathcal{M}$

• Entropy of A:

$$egin{array}{rcl} \mathcal{H}_{\omega}(\mathcal{A}) &=& \sup_{\omega=\sum\lambda_{i}\hat{\omega}_{i}}\sum_{i}\lambda_{i}\,S(\hat{\omega}_{i}\,\!\!\mid\!\mathcal{A};\omega\!\mid\!\!\mathcal{A}) \ &=& S(\omega\!\mid\!\!\mathcal{A})-\sum_{i}\lambda_{i}\,S(\hat{\omega}_{i}\,\!\mid\!\!\mathcal{A})\;. \end{array}$$

Connes-Narnhofer-Thirring Entropy Alicki-Fannes entropy

n-subalgebra Entropy

n subalgebras A_1, A_2, \ldots, A_n

• Decompositions of ω , $\mathbf{i}^{(n)} = i_1 i_2 \cdots i_n \in I^n$, I discrete index set:

$$\omega = \sum_{\mathbf{i}^{(n)}} \lambda_{\mathbf{i}^{(n)}} \hat{\omega}_{\mathbf{i}^{(n)}} \;,\; \lambda_{\mathbf{i}^{(n)}} \geq 0 \;,\; \sum_{\mathbf{i}^{(n)}} \lambda_j = 1 \;, \hat{\omega}_{\mathbf{i}^{(n)}}(1) = 1 \;.$$

• sub-decompositions: sums over all indices but one,

$$egin{array}{rcl} \omega_{i_j}^j &=& \displaystyle\sum_{\mathbf{i}^{(n)}, \hat{l}_j} \lambda_{\mathbf{i}^{(n)}} \hat{\omega}_{\mathbf{i}^{(n)}} \;,\; \hat{\omega}_{i_j}^j = rac{\omega_{i_j}^J}{\omega_{i_j}^j(1)} \ \lambda_{i_j}^j &=& \displaystyle\sum_{\mathbf{i}^{(n)}, \hat{l}_j} \lambda_{\mathbf{i}^{(n)}} \;. \end{array}$$

Connes-Narnhofer-Thirring Entropy Alicki-Fannes entropy

n-subalgebra Entropy: $A = \bigvee_{j=1}^{n} A_j$ algebra generated by $\{A_j\}_{j=1}^{n}$

$$\begin{split} H_{\omega}(A_{1},A_{2},\ldots,A_{n}) &= \sup_{\omega=\sum\lambda_{i(n)}\hat{\omega}_{i(n)}} \left\{ H(\{\lambda_{i(n)}\}) - \sum_{j=1}^{n} H(\{\lambda_{i_{j}}^{j}\}) \right. \\ &+ \sum_{j=1}^{n} \sum_{i_{j}} \lambda_{i_{j}}^{j} S(\hat{\omega}_{i_{j}}^{j} \upharpoonright A_{j}; \omega \upharpoonright A_{j}) \right\} \\ &= \sup_{\omega=\sum\lambda_{i(n)}\hat{\omega}_{i(n)}} \left\{ H(\{\lambda_{i(n)}\}) - \sum_{j=1}^{n} \left(H(\{\lambda_{i_{j}}^{j}\}) \right. \\ &+ \left. \sum_{i_{j}} \lambda_{i_{j}}^{j} S(\hat{\omega}_{i_{j}}^{j} \upharpoonright A_{j}) - S(\omega \upharpoonright A_{j}) \right) \right\}. \end{split}$$

 $\leq S(\omega \restriction A_1) + S(\omega \restriction A_2)$

Connes-Narnhofer-Thirring Entropy Alicki-Fannes entropy

2-subalgebra Entropy

• Quantum upper bound:

• Concavity of von Neumann entropy:

$$S(\omega \upharpoonright A_j) \le H(\{\lambda_{i_j}^j\}) + \sum_{i_j} \lambda_{i_j}^j S(\hat{\omega}_{i_j}^j \upharpoonright A_j)$$

• Classical upper bound: $H_{\omega}(A_1, A_2) \leq \sup_{\omega = \sum \lambda_{i_1 i_2} \hat{\omega}_{i_1 i_2}} H(\{\lambda_{i_1 i_2}\})$

Connes-Narnhofer-Thirring Entropy Alicki-Fannes entropy

Commutative case: explicit computation

- $A_{1,2}$ subalgebras of A Abelian: $H_{\omega}(A_1, A_2) = S(\omega \upharpoonright A_1 \lor A_2)$
- Minimal projections $\{a_{i_j}^j\}$, j = 1, 2, $\sum_{i_j} a_{i_j}^j = 1$, $a_{i_j}^j a_{i_k}^j = \delta_{i_j i_k}$:

$$\begin{split} \omega(a) &= \sum_{i_1 i_2} \omega(a_{i_1}^1 a_{i_2}^2 a) = \sum_{i_1} \omega(a_{i_1}^1) \sum_{i_2} \frac{\omega(a_{i_1}^1 a_{i_2}^2 a)}{\omega(a_{i_1}^1)} \\ \hat{\omega}_{i_j}^j(a) &= \frac{\omega(a_{i_j}^j a)}{\omega(a_{i_j}^j)} , \ \lambda_{i_1 i_2} = \omega(a_{i_1}^1 a_{i_2}^2 a) , \ \lambda_{i_j}^j = \omega(a_{i_j}^j) \end{split}$$

•
$$\omega_{i_j}^j \upharpoonright A_j = \{\omega_{i_j}^j (a_{k_j}^j) = \{\delta_{i_j k_j}\}, \ \omega \upharpoonright A_j = \{\omega(a_{i_j}^j) = \lambda_{i_j}^j\} \text{ imply}$$

 $S(\omega \upharpoonright A_1 \lor A_2) \le H_\omega(A_1, A_2) \le H(\{\lambda_{i_1 i_2}\}) = S(\omega \upharpoonright A_1 \lor A_2).$

Connes-Narnhofer-Thirring Entropy Alicki-Fannes entropy

Non Commutative case: factor state on a quantum spin chain

• Decomposition of a density matrix:

$$x_j \ge 0 \;,\; \sum_j x_j = 1 \Longrightarrow
ho = \sum_j \sqrt{
ho} x_j \sqrt{
ho} \;.$$

- $M_{1,2}$ commuting spin algebras at sites 1, 2, $\omega = \bigotimes_{n \in \mathbb{Z}} (\rho)_n$
- a_{ij}^{j} eigenprojections of $(\rho)_{j}$: $\sqrt{(\rho)_{j}}a_{i_{j}}^{j}\sqrt{(\rho)_{j}} = (\rho)_{j}a^{j}i_{j}$.
- $A_j \subset M_j$: maximally Abelian subslagebra generated by $\{a_{i_j}^j\}$:

$$\begin{array}{rcl} S(\omega \upharpoonright M_1 \otimes M_2) & \geq & H_{\omega}(M_1, M_2) \geq H_{\omega}(A_1, A_2) = H(\{\omega(a_{i_1}^1 a_{i_2}^2)\}) \\ & = & S(\omega \upharpoonright M_1 \otimes M_2) = 2S(\rho) \ . \end{array}$$

• $H_{\omega}(M_1, M_2) = H_{\omega}(A_1, A_2) = S(\omega \upharpoonright M_1 \otimes M_2) = 2S(\rho)$

Connes-Narnhofer-Thirring Entropy Alicki-Fannes entropy

Properties of *n*-subalgebra entropies

• Positivity and boundedness:

$$0 \leq H_\omega(A_1,A_2,\ldots,A_n) \leq \sum_{j=1}^n S(\omega {\upharpoonright} A_j)$$

• Invariance under repetition:

$$H_{\omega}(A_1, A_1, A_2, \ldots, A_n) = H_{\omega}(A_1, A_2, \ldots, A_n)$$
.

• Invariance under permutations:

$$H_{\omega}(A_1,\ldots,A_n)=H_{\omega}(A_{\pi(1)},\ldots,A_{\pi(n)})$$
.

• Invariance under automorphisms $\Theta : \mathcal{M} \mapsto \mathcal{M}$:

$$H_{\omega}(\Theta(A_1),\Theta(A_2),\ldots,\Theta(A_n))=H_{\omega}(A_1,A_2,\ldots,A_n)\;.$$

Connes-Narnhofer-Thirring Entropy Alicki-Fannes entropy

Properties of *n*-subalgebra entropies

• Sub-additivity:

$$H_{\omega}(A_1,\ldots,A_n) \leq H_{\omega}(A_1,\ldots,A_p) + H_{\omega}(A_{p+1},\ldots,A_n)$$

Mononicity under embeddings

$$egin{aligned} &A_i\subseteq B_i\Longrightarrow H_\omega(A_1,\ldots,A_n)\leq H_\omega(B_1,\ldots,B_n)\ &A=\bigvee_{j=1}^nA_j\Longrightarrow H_\omega(A_1,A_2,\ldots,A_n)\leq H_\omega(A)\ . \end{aligned}$$

Connes-Narnhofer-Thirring Entropy Alicki-Fannes entropy

CNT-Entropy

- Quantum triple: $(\mathcal{M}, \Theta, \omega)$, $\omega \circ \Theta = \omega$
- Dynamical information rate about $A \subseteq \mathcal{M}$:

$$h_{\omega}^{\mathsf{CNT}}(\Theta, A) = \lim_{n \to +\infty} \frac{1}{n} H_{\omega}(A, \Theta(A), \dots, \Theta^{n-1}(A))$$

• CNT-Entropy:

$$h_{\omega}^{\mathsf{CNT}}(\Theta) = \sup_{A \subseteq \mathcal{M}} h_{\omega}^{\mathsf{CNT}}(\Theta, A)$$

• Nested sequence of finite dim. subalgebras: $\bigvee_n A_n = \mathcal{M}$,

$$h^{\mathsf{CNT}}(\theta) = \lim_{n \to +\infty} h^{\mathsf{CNT}}_{\omega}(\Theta, A_n) .$$

Connes-Narnhofer-Thirring Entropy Alicki-Fannes entropy

Quantum spin chains: CNT-entropy for $\omega = \bigotimes_{n \in \mathbb{Z}} (\rho)_n$

- Quantum spin chain: $(M_d^{\otimes \infty}, \Theta, \omega), \Theta[(M_d)_n] = (M_d)_{n+1}$.
- Nested generating sequence: $A_n = M_{[0,n-1]}$:

$$\begin{aligned} h_{\omega}^{\mathsf{CNT}}(\Theta) &= \lim_{n} h_{\omega}^{\mathsf{CNT}}(\Theta, M_{[0,n-1]}) \\ &\leq \lim_{n} \lim_{k} \frac{1}{k} H_{\omega}(M_{0,n-1]}, M_{[1,n]}, \dots, M_{[k,n+k-1]}) \\ &\leq \lim_{n} \lim_{k} \frac{1}{k} H_{\omega}(M_{0,n+k-1]}) \\ &\leq \lim_{n} \lim_{k} \frac{1}{k} S(\omega(M_{0,n+k-1]}) = s(\omega) = S(\rho) \;. \end{aligned}$$

Connes-Narnhofer-Thirring Entropy Alicki-Fannes entropy

Quantum spin chains: CNT-entropy for $\omega = \bigotimes_{n \in \mathbb{Z}} (\rho)_n$

• Choose k = np + q, $0 \le q < n$

• h_w^C

 A_[jn,(j+1)n-1] ⊂ M_[jn,(j+1)n-1] maximally abelian subalgebras generated by the eigenprojections of ⊗^{(j+1)n-1}_{ℓ=in}(ρ)_ℓ:

$$\begin{split} H_{\omega}(M_{0,n-1}], M_{[1,n]}, \dots, M_{[k,n+k-1]}) &\geq \\ &\geq H_{\omega}(M_{0,n-1}], M_{[n,2n-1]}, \dots, M_{[(\rho-1)n,\rho n-1]}) \\ H_{\omega}(A_{0,n-1}], A_{[n,2n-1]}, \dots, A_{[(\rho-1)n,\rho n-1]}) &\geq \\ &\geq S(\omega \upharpoonright M_{[0,kn-1]}) = knS(\rho) \\ \end{split}$$

Connes-Narnhofer-Thirring Entropy Alicki-Fannes entropy

Alicki-Fannes Entropy: Quantum Partitions

- Quantum triple: $(\mathcal{M}, \Theta, \omega)$
- Partition of unity:

$$\mathcal{X} = \{X_i\}_{i=1}^p , \quad \sum_{i=1}^p X_i^{\dagger} X_i = 1 , \quad X_i \in \mathcal{M}_0$$

• Partition of unit time-evolved up to t = n - 1:

$$\mathcal{X}^{(n)} = \{X_{\mathbf{i}^{(n)}}\}, \quad X_{\mathbf{i}^{(n)}} = \Theta^{(n-1)}[X_{i_{n-1}}] \cdots \Theta[X_{i_1}]X_{i_0}$$

• partition-density matrix:

$$\rho[\mathcal{X}^{(n)}] = [\omega(X_{\mathbf{i}^{(n)}}^{\dagger}X_{\mathbf{j}^{(n)}}^{\dagger}]_{p^n \times p'}$$

Connes-Narnhofer-Thirring Entropy Alicki-Fannes entropy

Alicki-Fannes entropy

• Entropy of a partition:

$$S[\mathcal{X}^{(n)}] = -\mathrm{Tr}\Big(
ho[\mathcal{X}^{(n)}]\log
ho[\mathcal{X}^{(n)}]\Big)$$

• Partition entropy rate:

$$h_{\omega}(\Theta,\mathcal{X}) = \limsup_{n o \infty} rac{1}{n} S[\mathcal{X}^{(n)}]$$

• Alicki-Fannes Entropy:

$$h^{\sf AF}_\omega(\Theta) = \sup_{\mathcal{X}} h_\omega(\Theta, \mathcal{X})$$

Connes-Narnhofer-Thirring Entropy Alicki-Fannes entropy

AF Entropy: another interpretation

• GNS representation: $\rho = \sum_{j=1}^{n} r_j |r_j\rangle \langle r_j|, \ \mathcal{X} = \{X_i\}_{i \in I},$

$$|\sqrt{\rho}\rangle = \sum_{j=1}^{n} \sqrt{r_j} |r_j\rangle \otimes |r_j\rangle$$

• Reference vector state: $\{|i\rangle\}_{i\in I}$ ONB,

$$|\Psi_{\mathcal{X}}\rangle = \sum_{i\in I}\sum_{j=1}^{n}\sqrt{r_{j}}X_{i}|r_{j}\rangle\otimes|r_{j}\rangle\otimes|i\rangle$$

• Marginal density matrices: partial traces,

$$\begin{array}{lll} \mathrm{Tr}_{12}|\Psi_{\mathcal{X}}\rangle\langle\Psi_{\mathcal{X}}| &=& \rho[\mathcal{X}]\\ \mathrm{Tr}_{3}|\Psi_{\mathcal{X}}\rangle\langle\Psi_{\mathcal{X}}| &=& \mathcal{R}[\mathcal{X}] = \sum_{i\in I}X_{i}\otimes 1|\sqrt{\rho}\rangle\langle\sqrt{\rho}|X_{i}^{\dagger}\otimes 1\end{array}$$

Connes-Narnhofer-Thirring Entropy Alicki-Fannes entropy

Successive measurement processes

• Marginal density matrices of pure states: same von Neumann entropy:

$$S(\rho[\mathcal{X}]) = S(\mathcal{R}[\mathcal{X}])$$

• Partition: Measurement processes

$$\mathcal{X} \Longrightarrow \rho \mapsto \mathbb{E}_{\mathcal{X}}[\rho] = \sum_{i \in I} X_i \, \rho \, X_i^{\dagger}$$

• Repeated measurements at successive times upon the GNS projector:

$$ho[\mathcal{X}^{(n)}] = \sum_{\mathbf{i}^{(n)}} X_{\mathbf{i}^{(n)}} |\sqrt{
ho} \rangle \langle \sqrt{
ho} | X_{\mathbf{i}^{(n)}}^{\dagger}$$

Connes-Narnhofer-Thirring Entropy Alicki-Fannes entropy

Quantum spin chains: AF-entropy

- Quantum spin chain: $(M_d^{\otimes \infty}, \Theta, \omega), \Theta[(M_d)_n] = (M_d)_{n+1}$
- Partition: $\mathcal{X} = \{X_{ij}\}_{i,j=1}^d$

$$X_{ij} = rac{|i\rangle\langle j|}{\sqrt{d}} \;,\; \{|i
angle\}_{i=1}^d \; {\it ONB} \; {\it in} \mathbb{C}^d \;.$$

• Refined partition: $\mathcal{X}^{(n)} = \{X_{\mathbf{i}^{(n)}\mathbf{j}^{(n)}}\}$

$$X_{\mathbf{i}^{(n)}\mathbf{j}^{(n)}} = \frac{1}{\sqrt{d^n}} |i_{n-1}\langle j_{n-1} \otimes \cdots |i_{n-1} \langle j_{n-1} \rangle.$$

• Partition density matrix: $\rho[\mathcal{X}^{(n)}] = \frac{1}{d^n} \otimes \rho_{[0,n-1]}$,

$$\rho[\mathcal{X}^{(n)}]_{(\mathbf{i}^{(n)}\mathbf{j}^{(n)}):(\mathbf{k}^{(n)}\mathbf{p}^{(n)})} = \frac{1}{d^n} \omega \left(\bigotimes_{\ell=0}^n |p_\ell\rangle \langle k_\ell | i_\ell\rangle \langle j_\ell | \right) = \prod_{\ell=0}^n \frac{\delta_{k_\ell i_\ell}}{d} \rho_{\mathbf{j}^{(n)}\mathbf{p}^{(n)}} \,.$$

Connes-Narnhofer-Thirring Entropy Alicki-Fannes entropy

Comparison with mean entropy

• Partition entropy:

$$S(\rho[\mathcal{X}^{(n)}]) = S(\rho_{0,n-1}) + n \log d$$

• AF-Entropy:

$$h_{\omega}^{\mathsf{AF}}(\Theta) = s(\omega) + \log d = h_{\omega}^{\mathsf{CNT}}(\Theta) + \log d$$

Has log *d* any interpretation in terms of quantum algorithmic complexity?

Connes-Narnhofer-Thirring Entropy Alicki-Fannes entropy

AF-entropy: successive mesurements

- partitions: $\mathcal{X} = \{X_i\}_{i=1}^p$, $\mathcal{X}^{(n)} = \{X_{\mathbf{i}^{(n)}} = \Theta^{n-1}[X_{i_{n-1}}] \cdots \Theta[X_{i_1}]X_{i_0}\}$
- GNS representation: $(\pi_{\omega}(\mathcal{M}), U_{\omega}^{\Theta}, \Omega_{\omega})$
- purification:

$$|\Psi_{\mathcal{X}}^{(n)}
angle = \sum_{\mathbf{i}^{(n)}} \pi_{\omega}(X_{\mathbf{i}^{(n)}})|\Omega_{\omega}
angle \otimes |\mathbf{i}^{(n)}
angle$$

• partial traces: $\operatorname{Tr}_{I}\left(|\Psi_{\mathcal{X}}^{(n)}\rangle\langle\Psi_{\mathcal{X}}^{(n)}|\right) = \rho[\mathcal{X}^{(n)}]$

$$R[\mathcal{X}^{(n)}] = \operatorname{Tr}_{II}\left(|\Psi_{\mathcal{X}}^{(n)}\rangle\langle\Psi_{\mathcal{X}}^{(n)}|\right) = \sum_{\mathbf{i}^{(n)}} \pi_{\omega}(X_{\mathbf{i}^{(n)}})|\Omega_{\omega}\rangle\langle\Omega_{\omega}|\pi_{\omega}(X_{\mathbf{i}^{(n)}})^{\dagger}$$

• same entropies: $S(\rho[\mathcal{X}^{(n)}]) = S(R[\mathcal{X}^{(n)}])$

Connes-Narnhofer-Thirring Entropy Alicki-Fannes entropy

AF-entropy: successive mesurements

• Density matrix at time n-1:

$$R[\mathcal{X}^{(n)}] = \operatorname{Tr}_{II}\left(|\Psi_{\mathcal{X}}^{(n)}\rangle\langle\Psi_{\mathcal{X}}^{(n)}|\right) = \sum_{\mathbf{i}^{(n)}} \pi_{\omega}(X_{\mathbf{i}^{(n)}})|\Omega_{\omega}\rangle\langle\Omega_{\omega}|\pi_{\omega}(X_{\mathbf{i}^{(n)}})^{\dagger}$$

• Subsequent time-evolutions and mesurements:

$$\pi_{\omega}(X_{i_0i_11_2}^{(2)})|\Omega_{\omega}\rangle = (U_{\omega}^{\dagger})^2 \underbrace{\pi_{\omega}(X_{i_2})}_{meas} \underbrace{U_{\omega}}_{time-ev} \underbrace{\pi_{\omega}(X_{i_1})}_{meas} \underbrace{U_{\omega}^{\dagger}}_{time-ev} \underbrace{\pi_{\omega}(X_{i_0})}_{meas} |\Omega_{\omega}\rangle$$

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Qubit



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Qubit



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CTMs vs QTMs

Classical TM ${\cal U}$

- (0,1) on Input and Output Cells: configuration space t
- Read-Write Head Position: position space h
- Internal Control Unit: internal state space q
- \mathcal{U} reversibly operates on input p and halts with $\mathcal{U}(p)$ written on the output tape

Quantum TM U

- Tape Hilbert Space: $|t\rangle \in \mathbb{H}_{T_{in}} \otimes \mathbb{H}_{T_{out}}$
- Head Hilbert Space: $|h\rangle \in \mathbb{H}_H$
- Control Unit Hilbert Space: $|q
 angle \in \mathbb{H}_Q$
- QTM Hilbert Space: $\mathbb{H}_{THQ} = \mathbb{H}_{T_{in}} \otimes \mathbb{H}_{T_{out}} \otimes \mathbb{H}_{H} \otimes \mathbb{H}_{Q}$

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QTM: CP map \mathbb{R}

• unitary time evolution U:

$$|\Psi_{in}
angle\mapsto U|\Psi_{in}
angle=|\Psi^{*}
angle\in\mathbb{H}_{THQ}$$

- QTM halts halts when Control Unit in halting state $|q^*
 angle$
- output obtained by an CP-map:

$$\mathbb{R}[|\Psi_{in}\rangle\langle\Psi_{in}|] = \mathrm{Tr}_{\mathcal{T}_{in}HQ}[|\Psi^*\rangle\langle\Psi^*|]$$

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Objects to Describe and Descriptions

UTM: *U*

- Objects to describe : bit-strings $\mathbf{i}^{(n)} \in \Omega_2^{(n)}$
- Described by binary programs: $p \in \Omega_2^{(m)}$: $\mathcal{U}(p) = \mathbf{i}^{(n)}$

UQTM: irreversible completely positive map \mathbb{R}

- Object to describe: *n* qubit vectors $|\Psi
 angle\in\mathbb{C}^{2^n}$
- Described by:

Q quantum states: Quantum Qubit Complexity $QC_{q}(\Psi)$

$$\pmb{\sigma} \in \pmb{M}_{2^m} \Longrightarrow \mathbb{R}[\pmb{\sigma}] \simeq |\Psi\rangle\langle\Psi|$$

2 bit-strings: Quantum Bit Complexity (Vitanyi 2001) $QC_c(\Psi)$

$$\mathbf{p} \in \Omega_2^m \Longrightarrow \mathbb{R}[\mathbf{p}] \simeq |\Psi\rangle\langle\Psi|$$

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Approximate Descriptions

bit-strings vs qubit-strings

- there are 2ⁿ bit-strings of length n
- the *n*-qubit state-vectors $|\Psi\rangle \in \mathbb{C}^{2^n}$ cover the surface of the unit sphere in $\mathbb{R}^{2^{n+1}}$
- trace distance:

$$D\Big(\mathbb{R}[\sigma], |\Psi\rangle\langle\Psi|\Big) := \mathrm{Tr}\Big|\mathbb{R}[\sigma] - |\Psi\rangle\langle\Psi|\Big|$$

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Quantum Qubit Complexity (Berthiaumme et al. 2001)

Quantum Qubit Complexity: $QC_q(\Psi)$

• Length of a qubit string:

$$\ell(
ho):=\min\left\{n\in\mathbb{N}\,:\,
ho\in M_{2^n}(\mathbb{C})
ight\}$$

• Quantum Qubit Complexity:

$$\operatorname{QC}_{q}(\Psi) = \min \left\{ \ell(\sigma) : D(|\Psi\rangle \langle \Psi|, \mathbb{R}[k, \sigma]) \leq \frac{1}{k} \ \forall \ k \in \mathbb{N} \right\}$$

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Quantum Qubit Complexity: upper bound

$QC_{q}(\Psi)$: upper bound

• Simple copying:

$$\operatorname{QC}_{q}(\Psi) \leq n + C$$

• quantum counting argument: $P = |\Psi\rangle\langle\Psi|$

$$\mathsf{N} \;\;=\;\; \mathsf{card} \left\{ \mathsf{P}_i \perp \mathsf{P}_j : \; D(\mathbb{R}[\sigma_i], \mathsf{P}_i) \leq \delta \;, \; \ell(\sigma_i) \leq n \, \mathsf{s}(\omega)
ight\}$$

$$N \leq 2^{ns(\omega)-\eta(\delta)}$$

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Quantum Qubit Complexity: Brudno's relation (F.B. et al. 2006)

Classical Brudno: entropy per symbol = complexity per symbol

ergodic sources $k(i) = h(\pi)$ $\pi - a.e.$

$QC_{q}(\Psi)$: Brudno's relation

- Ergodic quantum sources with entropy rate $s(\omega)$
- for every $\delta > 0$, there exist typical projections $Q_n \in M_{2^n}$ such that, for *n* large enough, $Tr(\rho^n Q_n) \simeq 1$ and

$$oldsymbol{s}(\omega) - \delta \leq rac{1}{n} \mathrm{QC}_q\left(\Psi
ight) \leq oldsymbol{s}(\omega) + \delta \qquad orall |\Psi
angle \langle \Psi| \leq Q_r$$

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Universal Density Matrix (Gacs 2001)

Universal probability

$$\mathbb{P}(\mathbf{i}^{(n)}) := \sum_{p \ : \ \mathcal{U}[p] = \mathbf{i}^{(n)}} 2^{-\ell(p)} \ , \ \ \mathcal{U} \ \mathsf{prefix} \ \mathsf{UTM}$$

Universal density matrix

- $i \in \Omega$: set of all binary strings
- Ω : standard basis of vectors $|\mathbf{i}\rangle$ spanning the Hilbert space $L^2(\Omega)$
- Elementary vectors $|\Psi_{el}\rangle = \sum_{i \in \Omega} C_i |i\rangle \in L^2(\Omega)$: only finitely many algebraic numbers $C_i \neq 0$: $\Psi_{el} = i_{\Psi_{el}} \in \Omega$
- a-priori semi-density matrix:

$$\mathbb{D} = \sum_{|\Psi_{el}
angle} \mathbb{P}(\mathbf{i}_{\Psi_{el}}) \ket{\Psi_{el}} ra{\Psi_{el}}$$

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Universality of \mathbb{D} and Operator Complexity

- $\bullet~\mathbb{P}$ computed with a prefix UTM implies $\mathrm{Tr}(\mathbb{D}) \leq 1$
- Universality of $\mathbb{D}:$ for all semi-computable, semi-density matrices, ${\rm Tr}\rho\leq 1,$

$$\exists C_{\rho} > 0 : C_{\rho} \rho \leq \mathbb{D}.$$

• Operator Complexity:

$$\kappa_q = -\log \mathbb{D}$$

• Gacs state complexity:

$$\overline{H}(\rho) = \operatorname{Tr}(\rho \,\kappa_q)$$

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$S(\rho)\simeq \overline{H}(\rho)$

• Universality, $\rho \leq C_{\rho}^{-1}\mathbb{D}$ implies

$$-\log
ho\geq -\log\mathbb{D}+\log\mathcal{C}_
ho\Longrightarrow\mathcal{S}(
ho)\geq \overline{H}(
ho)+\log\mathcal{C}_
ho$$

• Positivity of relative entropy,

$$\operatorname{Tr}\left(
ho\left(\log
ho - \lograc{\mathbb{D}}{\operatorname{Tr}\mathbb{D}}
ight)
ight) \geq 0$$

and $\mathrm{Tr}\mathbb{D} \leq 1$ imply

$$S(\rho) \leq \operatorname{Tr}(\rho \kappa_q) + \log \operatorname{Tr}(\mathbb{D}) \leq \overline{H}(\rho)$$
.

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Gacs complexity and Alicki-Fannes entropy

Gacs operator complexity and entropy density (CNT-entropy)

- local partitions: $X_i \in M_{[-p,p]}, X_{i^{(n)}} \in M_{[-p,p+n-1]}$
- computable X_i and $\rho^{(n)}$ states on $M_d^{\otimes n}$:
- Maximal Gacs Complexity rate:

$$\sup_{\mathcal{X}} \lim_{n \to +\infty} \frac{1}{n} \operatorname{Tr} \left(\rho^{(n)} \kappa_q^{(n)} \right) = s(\omega)$$

Quantum Algorithmic Complexities Quantum Qubit Complexity Universal Density Matrix Gacs complexity and Quantum Spin Chains Bibiography

Gacs complexity and AF entropy

- GNS representation: $(\pi_{\omega}(\mathcal{M}), U^{\sigma}_{\omega}, \Omega_{\omega});$
- local partitions: $X_i \in M_{[-p,p]}$, $X_{i^{(n)}} \in M_{[-p,p+n-1]}$
- computable X_i and $R[\mathcal{X}^{(n)}]$ states on $M_d^{\otimes n} \otimes M_d^{\otimes n}$:

$$\operatorname{Tr}\left(R[\mathcal{X}^{(n)}]\,\kappa_q^{(n)}\right)\simeq S(R[\mathcal{X}^{(n)}])=S(\rho[\mathcal{X}^{(n)}])$$

• Maximal Gacs Complexity rate:

$$\sup_{\mathcal{X}} \lim_{n \to +\infty} \frac{1}{n} \operatorname{Tr} \left(R[\mathcal{X}^{(n)}] \, \kappa_q^{(n)} \right) = s(\omega) + \log d$$

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