

Non-commutative Information Theory

An Overview

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Motivation:

Understand von Neumann algebras.

$\hookrightarrow \mathcal{M} \subseteq \mathcal{B}(\mathcal{H})$ unital $*$ -closed subalgebra

with: (i) $\mathcal{M} = \overline{\mathcal{M}}^{\text{SOT}}$

(ii) $\mathcal{M} = \overline{\mathcal{M}}^{\text{WOT}}$

(iii) $\mathcal{M} = (\mathcal{M}')'$ the bicommutant

(von Neumann: (i) \Leftrightarrow (ii) \Leftrightarrow (iii))

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Examples

- $M_n(\mathbb{C})$
- $\mathcal{B}(\mathcal{H})$
- $\text{Diag}(M_n(\mathbb{C})) \cong \mathbb{C}^n$
- Block diagonal subalgebra of $\mathcal{B}(\mathcal{H})$
(if (p_i) are orthogonal projections with $\sum_i p_i = 1$,
 $\bigoplus_i p_i \mathcal{B}(\mathcal{H}) p_i$ (...))
- Γ countable, discrete group
 $\lambda: \mathbb{C}[\Gamma] \hookrightarrow \mathcal{B}(l^2(\Gamma))$
 $L\Gamma := \overline{\lambda(\mathbb{C}[\Gamma])}^{\text{SOT}}$
- $L^\infty(\Omega, \mathbb{P}) \subseteq L^2(\Omega, \mathbb{P})$

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Problem: It is hard to tell if two vNa's are isomorphic.

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- $L(\mathbb{Z}) \cong L^\infty(\mathbb{T}) \cong L^\infty(\mathbb{T}^2) \cong L(\mathbb{Z}^2)$
- Any two separable amenable discrete ICC groups generate isomorphic vNas
($\mathbb{S}_\infty, (\mathbb{Z}/2\mathbb{Z}) \wr \mathbb{Z}, \dots$)
- $L(\mathbb{F}_2) \not\cong L(\mathbb{S}_\infty)$
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Non-commutative probability theory

If commutative vNa's are $L^\infty(\Omega, \mathbb{P})$,
view arbitrary vNa's as spaces of
"non-commutative random variables".

A *tracial state* on \mathcal{M} is a linear map

$$\tau : \mathcal{M} \longrightarrow \mathbb{C} \quad \text{which is}$$

positive $(\tau(a^*a) \geq 0)$

unital $(\tau(1) = 1)$

tracial $(\tau(ab) = \tau(ba))$

We'll generally also assume

faithful $(\tau(a^*a) = 0 \Rightarrow a = 0)$

normal $(\tau(\sup x_i) = \sup \tau(x_i))$ for (x_i) bounded increasing positive net

Examples

- $\text{tr} : M_n(\mathbb{C}) \rightarrow \mathbb{C}$
 - $\langle \delta_{e_j}, \delta_{e_i} \rangle : L^1 \rightarrow \mathbb{C}$
 - $\mathbb{E} : L^\infty(\Omega, \mathcal{P}) \rightarrow \mathbb{C}$
-

If $x = x^* \in (\mathcal{M}, \tau)$, x generates a commutative vNa.
What is the corresponding probability space?

$L^\infty(\sigma(x), \tau \circ P_x)$ where P_x is the spectral measure:

$$f(x) = \int_{\sigma(x)} f(\lambda) dP_x(\lambda)$$

Free Independence

$(\mathcal{A}_i)_{i \in I}$ \ast -subalgebras of (\mathcal{M}, τ) are

freely independent if whenever

$$i_1 \neq i_2 \neq \dots \neq i_n \quad \text{in } I$$

$$x_k \in \mathcal{A}_{i_k} \cap \ker(\tau)$$

we have $x_1 \dots x_n \in \ker(\tau)$

Specifies all mixed moments by centring and recursion:

$$0 = \tau\left((x_1 - \tau(x_1)1) \cdots (x_n - \tau(x_n)1)\right)$$

$$= \tau(x_1 \cdots x_n) - \tau(x_1)\tau(1 x_2 \cdots x_n) - \dots$$

Example

$$\begin{aligned} 0 &= \tau((x - \tau(x)1)(y - \tau(y)1)) \\ &= \tau(xy) - \tau(x)\tau(1y) - \tau(x1)\tau(y) + \tau(x)\tau(y)\tau(1) \\ \Rightarrow \tau(xy) &= \tau(x)\tau(y) \end{aligned}$$

More calculations

$$\Rightarrow \tau(xyxy) = \tau(x^2)\tau(y)^2 + \tau(x)^2\tau(y^2) - \tau(x)^2\tau(y)^2$$

In particular,

$$\begin{aligned} \tau((xy - yx)^*(xy - yx)) &= 2(\tau(x^2) - \tau(x)^2)(\tau(y^2) - \tau(y)^2) \\ &= 2 \operatorname{Var}(x) \operatorname{Var}(y) \end{aligned}$$

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The Mandatory Dictionary Slide

$$(L^\infty(\Omega, \mathbb{P}), \mathbb{E}) \quad \text{—————} \quad (\mathcal{M}, \tau)$$

$$([n], \frac{1}{n} \sum \delta_i) \quad \text{—————} \quad (M_n(\mathbb{C}), \text{tr})$$

$$\text{Independence} / \otimes \quad \text{—————} \quad \text{Free independence} / *$$

$(x_i)_i$ i.i.d.

$$\frac{1}{\sqrt{n}} \sum x_i \rightarrow \text{bell curve}$$

⋮

$(x_i)_i$ f.i.i.d.

$$\frac{1}{\sqrt{n}} \sum x_i \rightarrow \text{semi-circle}$$

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Free Information Theory

Characterize $W^*(x_1, \dots, x_n)$ based on the mixed moments of x_1, \dots, x_n .

$$\left(\text{E.g. } W^*(x_1, \dots, x_n) \text{ abelian} \Leftrightarrow \tau \left(\sum_{i,j} | [x_i, x_j] | \right) = 0 \right)$$

Entropic quantities give a measure of "how close to free" x_1, \dots, x_n are.

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Microstates:

$$\chi(\pi_1, \dots, \pi_n) = \sup_{R > 0} \inf_{\mathcal{Q} \ni \mu_n} \limsup_{k \rightarrow \infty} \left(\frac{n \log k}{2} + \frac{1}{k^2} \log \lambda \left\{ X \in M_k^{\text{sc}}(\mathbb{C})^n \mid \begin{array}{l} \|X\| < R \\ \mu_X \in \mathcal{Q} \end{array} \right\} \right)$$

= asymptotic log volume of matricial approximations

If $U \sim \text{Haar}(\mathcal{U}(N))$, $X \in M_N(\mathbb{C})$ deterministic,
 X and UXU^* are almost free

If $W^*(\pi_1, \dots, \pi_n)$ has a central projection,
matricial approximations are almost simultaneously
block diagonalizable.

[Voiculescu, many others]

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de Bruijn's formula, oversimplified:

$$\text{if } X(t) = X + \sqrt{t}N,$$

$$\frac{dS(X(t))}{dt} = \frac{1}{2} \left\| \left(\frac{d}{dt} \right)^* 1 \right\|_{L^2(\mathbb{R}, \mu_{X(t)})}^2$$

Consequence:

to understand entropy, we can look at the derivative as an unbounded operator on L^2 .

$L^2(\mathcal{M}, \tau)$ the completion of \mathcal{M} wrt $\langle x, y \rangle = \tau(x^*y)$.

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Free Stein Dimension

[C, Nekou]

$$A = \mathbb{C}\langle x_1, \dots, x_n \rangle \subseteq W^*(x_1, \dots, x_n) = \mathcal{M}$$

$$\text{Der}_{\tau, \sigma_1}(A) = \left\{ \delta: A \rightarrow L^2(\mathcal{M}, \tau) \otimes L^2(\mathcal{M}, \tau) \mid \begin{array}{l} \delta(xy) = x\delta(y) + \delta(x)y \\ \mathbb{1} \in \text{dom}(\delta^*) \end{array} \right\}$$

$$= \left\{ \text{"nice" derivations } A \rightarrow L^2(\mathcal{M}, \tau) \otimes L^2(\mathcal{M}, \tau) \right\}$$

can be equipped with the structure of

an \mathcal{M} - \mathcal{M} bimodule, and has a corresponding dimension in $[0, n]$, denoted $\sigma(A)$.

The bimodule structure depends on x_1, \dots, x_n , but $\sigma(A)$ does not!

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Free Stein Dimension

Can bootstrap many results: e.g.,

$\sigma(x_1, \dots, x_n) = n \Rightarrow$ no n.c. rational relations

(heavily using work of [Connes-Shlyakhtenko]
and [Mai-Speicher-Yin])