



Operator Algebras
$$\longrightarrow$$
 Function Algebras
Unital C*-Algebras \longrightarrow $G(X)$ for X compact, Hausdorff
Von Neumann Algebras \longrightarrow $L^{\infty}(X, \mu)$

A concrete example:

$$A = \mathcal{F}([0,1]) \subset \mathcal{B}(L^{2}([0,1],d\lambda))$$

$$A'' = \overline{A}^{00T} = \overline{A}^{00T} = L^{\infty}([0,1],d\lambda) \subset \mathcal{B}(L^{2}([0,1],d\lambda))$$
So we should be thinking measure theory or probability.

Probability Crash Course
A probability space is a triple
$$(\Omega, \mathcal{F}, \mathbb{P})$$
 where:
 Ω is a set;
 \mathcal{F} is a \mathcal{G} -algebra; and
 \mathcal{P} is a gositive measure on (Ω, \mathcal{F}) with $\mathbb{P}(\Omega) = 1$
These give rise to an expectation
 $\mathbb{E}: L'(\Omega, \mathbb{P}) \longrightarrow \Omega$
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 \mathbb{E}
We have the space of essentially bounded random variables $L^{\infty}(\Omega, \mathbb{P})$ which give
bounded operators on $L^{1}(\Omega, \mathbb{P})$ by multiplication.

Probability Crash Course
let's record some properties:

$$\cdot [\mathbb{P}(\Omega, P)] = L^{\infty}(\Omega, P)'$$
 is a von Neumann algebra
 $\cdot \mathbb{E}[1] = 1$
 $\cdot [\mathbb{F} | X = 0 \text{ in } L^{\infty}(\Omega, P) \text{ then } \mathbb{E}[X] = 0$
 $\cdot [\mathbb{F} | 0 \leq X_1 \leq X_2 \leq \dots \leq \infty, \text{ then } \lim_{n \to \infty} \mathbb{E}[X_n] = \mathbb{E}[\lim_{n \to \infty} X_n]$
(Mondene convergence theorem)
 $\cdot \mathbb{E}[XT] = \mathbb{E}[XX]$
A key property is independence:
(unital) subalgebras $(A_1)_{i \in I}$ of $L^{\infty}(X, P)$ are independent if
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whenever nelly, $i_{1,2-y}$ in $\mathbb{E}[X \text{ ore } distint,$
and $X_i \in A_{i_1,2-y_1} = \mathbb{E}[X_{i_1} - \mathbb{E}[X_{i_1}] = \dots = \mathbb{E}[X_{i_n}] = 0$
We have $\mathbb{E}[X_{i_1} - \mathbb{E}[X_{i_n}] = 0$.
Variables are independent if the van Neumann algebras they generate are

Wait, what? That isn't the usual definition!
Wait, what? That isn't the usual definition!
Well, suppose
$$T_1, T_2$$
 are independent variables
and f_1, f_2 are measurable functions.
Write $\alpha_1 = \mathbb{E}[\mathcal{F}_1(\mathbb{F}_1)], \ \alpha_2 = \mathbb{E}[\mathcal{F}_2(\mathbb{F}_2)].$
Then

$$O = \mathbb{E}\left[\left(\mathcal{Y}_{1}(\mathcal{I}_{1}) - \alpha_{1}\right)\left(\mathcal{Y}_{2}(\mathcal{I}_{2}) - \alpha_{2}\right)\right]$$

$$= \mathbb{E}\left[\mathcal{Y}_{1}(\mathcal{Y}_{1})\mathcal{Y}_{2}(\mathcal{I}_{2})\right] - \alpha_{1}\mathbb{E}\left[\mathcal{Y}_{2}(\mathcal{I}_{2})\right] - \alpha_{2}\mathbb{E}\left[\mathcal{Y}_{1}(\mathcal{I}_{2})\right] + \alpha_{1}\alpha_{2}$$

$$\mathbb{E}\left[\mathcal{Y}_{1}(\mathcal{I}_{1})\mathcal{Y}_{2}(\mathcal{I}_{2})\right] = \alpha_{1}\alpha_{2}$$

Using similar tricks, independence prescribes all mixed moments in terms of pure moments.

Theorem : The (Weekened) Central Limit Theorem
Suppose
$$(X_n)_{n \in \mathbb{N}}$$
 are independent random variables in $L^{\infty}(\Omega, \mathbb{P})$,
so that:
 $E[X_n] = E[X_n] \quad \forall k$ (they are identically distributed)
 $E[X_n] = 0$
 $E[X_n] = 0$
 $E[X_n]^2 = 1$.
Let $S_N = \int_N \sum_{i=1}^N X_i$.
Then for all k , $\lim_{N \to \infty} E[S_N^k] = \{(k-i)!\}$ if k is even
 0 else

Note: The Gaussian distribution N(0, i) is the unique distribution with these moments, but it is not bounded and so doesn't quite fit in this simplified framework.

Proof: Write
$$M_{L} = \mathbb{E}[X_{1}^{L}]$$
.

$$\mathbb{E}\left[S_{N}^{L}\right] = N^{\frac{1}{2}} \mathbb{E}\left[\left(\sum_{i=1}^{N} X_{i}\right)^{k}\right]$$

$$= N^{\frac{1}{2}} \mathbb{E}\left[\left(\sum_{i=1}^{N} X_{i}\right)^{k}\right]$$

$$= N^{\frac{1}{2}} \mathbb{E}\left[X_{alo} - X_{all}\right]$$

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$$= N^{\frac{1}{2}} \mathbb{E}\left[X_{alo} - X_{alo}$$

Frequencies Course
. Suppose M is a von Neumann algebra with trace
$$\tau$$
.
. $\tau[1] = 1$
. If $X \ge 0$ in M then $\tau[X] \ge 0$
. $|f \quad X \nearrow X$. then $\lim_{X \to X} \mathbb{E}[X_{n}] = \mathbb{E}[\lim_{X \to X} X_{n}]$
. $\tau[XY] = \tau[XX]$
A key property is Frequence integradence:
(unital) subalgebras $(M_{1})_{ieT}$ of M are independent if
whenever neW, $i_{1}, \dots, i_{n} \in T$ with $\tau[X_{1}] = \dots = \tau[X_{n}] = 0$
We have $\tau[X_{1}, \dots, X_{n} \in A_{1}]$ with $\tau[X_{1}] = \dots = \tau[X_{n}] = 0$
We have $\tau[X_{1}, \dots, X_{n} \in A_{1}]$ for Neumann algebras they generate are

Why do we need a new definition of independence?
The commutative one doesn't tell us how to evaluate products with
repeated terms, since they can always be reduced in that settings
e.g.
$$\Sigma_1 \Sigma_2 \Sigma_1 \Sigma_2 = \Sigma_1^2 \Sigma_2^2$$
,
Why not try a simpler rule, like factor τ across independent dyebres or just group
variables by algebra? E.g., ask for $\tau(\Sigma_1 \Sigma_2 \Sigma_3 \Sigma_3) = c(\Sigma_1) \tau(\Sigma_1) \tau(\Sigma_2)$ or
 $= \tau(\Sigma_1^2) \tau(\Sigma_2^2)?$.
In the first case, the constants work be independent from most algebras; in the second,
things are only independent if they commute under τ , which isn't great.
Using the same centring trick, we can compute for Σ_1, Σ_2 free, that
 $\tau[\Sigma_1 \Sigma_2 \Sigma, \Sigma_2] = \tau[\Sigma_1^2] \tau[\Sigma_2]^2 + \tau[\Sigma_2]^2 \tau[\Sigma_2^2] - \tau[\Sigma_1]^2 \tau[\Sigma_2]^2$

Theorem : The (Weekened) Central Limit Theorem
Suppose
$$(X_n)_{n \in \mathbb{N}}$$
 are subsciented random variables in M
so that : $\tau[X_n^k] = \tau[X_n^k] \quad \forall k \quad (they are identically distributed)$
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 $\cdot \tau[X_n^k] = 0$
 $\cdot \tau[X_n^k] = 1$.
Let $S_N = \frac{1}{N} \sum_{i=1}^{N} X_i$.
Then for all k , $\lim_{N \to \infty} \tau[S_N^k] = \begin{cases} C_{k/2} & \text{if } k \text{ is even} \\ 0 & \text{else} \end{cases}$
where $C_b = \frac{1}{b+1} {2b \choose b}$ is the b-th Catalan number.

Notice
$$\tau \left[X_1 X_2 X_1 X_2 \right] = 0$$

 $\tau \left[X_1^2 X_2^2 \right] = \tau \left[X_1^2 \right] \tau \left[X_2^2 \right] = 1$.
If $\pi \in P_2(k)$, we can compute $m_{\pi \tau}$ recursively.
If $k = 0$, $\pi = \emptyset$, then $m_{\pi} = 1$.
If τ has a block $\{d, d+1\}$, then
 $m_{\pi} = \mathbb{E} \left[\dots X_{\pi(d)}^2 \dots \right] = \mathbb{E} \left[X_{\pi(d)}^2 \right] M_{\pi \tau} \{\{d, d+1\}\}$.
If π has no such block, $m_{\pi} = 0$.
So $m_{\pi} = 1$ precisely when π can be reduced to the empty partition by removing blocks of consecutive ekments.

Combinatorial fact: these are precisely the non-crossing partitions.

A partition
$$\pi \in \mathbb{P}(k)$$
 is non-crossing if whenever
 $1 \notin \mathbb{W} < \mathfrak{X} < \mathfrak{Y} < \mathfrak{Z} \notin \mathbb{W}$ with $\mathbb{W} \sim \mathfrak{Y}, \ \mathfrak{X} \sim \mathfrak{Z}, \ \mathbb{W}$ have $\mathbb{W} \sim \mathfrak{Z}$.
The set of such is denoted NC(M).



Proof: Write
$$M_{k} = \mathbb{E}[X_{1}^{k}]$$
.

$$\mathbb{E}[S_{N}^{k}] = N^{k/2} \mathbb{E}[\left(\sum_{i=1}^{k} X_{1}\right)^{k}\right]$$

$$= N^{k/2} \mathbb{E}\left[\left(\sum_{i=1}^{k} X_{1}\right)^{k}\right]$$

$$= N^{k/2} \sum_{\alpha \in [1 \to [N]} \mathbb{E}[X_{\alpha(\alpha)} - X_{\alpha(\alpha)}]$$

$$= N^{k/2} \sum_{\alpha \in [1 \to [N]} \mathbb{E}[X_{\alpha(\alpha)} - X_{\alpha(\alpha)}]$$

$$= N^{-k/2} \sum_{\alpha \in [n]} \mathbb{E}\left[X_{\alpha(\alpha)} - X_{\alpha(\alpha)}\right]$$

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Is this the distribution of some random variable?
Consider
$$\mathcal{M} = \mathbb{P}^{2}(\mathbb{IN})$$
 and let $a \in \mathbb{B}(\mathcal{M})$ be the unilateral
shift, so $a \delta_{j} = \delta_{j+1}$. Note that $a^{*}\delta_{j} = \begin{cases} 0 & \text{if } j=1 \\ \delta_{j-1} & \text{else} \end{cases}$
Then $a^{*}a = 1$ and $aa^{*} = 1 - \operatorname{Proj}_{\delta_{1}}$.
Set $S = a + a^{*}$.
What is $\langle \delta_{1}, S^{k} \delta_{1} \rangle$?
A term in the expansion of S^{k} corresponds to a path
stepping up at each a^{*} and down at each a_{2}
 c_{3} . $a^{*}a^{*}aa^{*}aa$
To contribute, it must begin and end at the same level,
and never cross below where it started. These are the Dyck
paths, which are counted by C_{kh} .



Fact: The von Neumann algebra generated by n independent Gaussians is
isomosphic to the algebra generated by
$$1: L^{\infty}(\mathbb{R}^n, d\mathbb{X}^n) \cong L^{\infty}(\mathbb{R}, d\mathbb{X})$$

Ruestion: What about for free semicircular variables ?

 $\left(L^{\infty}(\Omega, d\mathbb{P}), \mathbb{E}\right)$ < >> Semicircular distribution Gaussian distribution Ind ependence Free independence Partitions < Non-crossing partitions Log Fourier transform 2 -> R transform Conditional expectation <>> Conditional expectation Entropy / information theory Free entropy Brownian motion Free Brownian motion (Lévy process with ind. stationary Gaussian increments) Free Brownian motion (Lévy process with find. stationary (increments)