

## Cumulants associated to n.c. independence

There are several n.c. independences; free.

One associate cumulants to them, cf. Voiculescu, Speicher.

Nonstochastic independence is more difficult and has not been defined before.

### 1 Independence on a n.c. probability space $(\mathcal{A}, \varphi)$

① Tensor independence: let  $\{\mathcal{A}_i\}_{i=1}^{\infty}$  be a seq of  $\ast$ -subalgebras of  $\mathcal{A}$ . It is tensor independent if  $\varphi(a_1 \dots a_n) = \prod_{i=1}^n \varphi(\prod_{j=1}^m a_{ij})$  " " means that order should be preserved.  
where  $\varphi(\emptyset) = 1$ .

② Free independence: here  $\mathcal{A}$  should be unital and  $\mathcal{A}_i \ni 1_{\mathcal{A}}$ . It is free [independent]

if  $\varphi(a_i) = 0$  and  $a_j \in \mathcal{A}_{i_j}$  with  $i_1 \neq i_2 \neq \dots \neq i_n$  implies  $\varphi(a_1 \dots a_n) = 0$

Application: If  $\{\mathcal{A}_1, \mathcal{A}_2\}$  are free, then  $\varphi(ab) = \varphi(a)\varphi(b)$  for  $a \in \mathcal{A}_1, b \in \mathcal{A}_2$

In fact, let  $\tilde{a} = a - \varphi(a)1, \tilde{b} = b - \varphi(b)1$ , so that  $\varphi(\tilde{a}) = \varphi(\tilde{b}) = 0$  and

$\varphi(\tilde{a}\tilde{b}) = 0$ , i.e.,  $\varphi(ab) = \varphi(a)\varphi(b) - \varphi(a)\varphi(b) + \varphi(a)\varphi(b) = 0$

In fact, you can calculate any mixed moments. When computations get involved, cumulants help!

③ Boolean independence: here  $\mathcal{A}_i \not\ni 1_{\mathcal{A}}$ . It is Boolean independent if

$\varphi(a_1 \dots a_n) = \prod_{j=1}^n \varphi(a_j)$  as soon as  $a_j \in \mathcal{A}_{i_j}, i_1 \neq i_2 \neq i_3 \neq \dots \neq i_n$ .  
in general (all are technical)

④ Monotone independence: here  $\mathcal{A}_i \not\ni 1_{\mathcal{A}}$ . It is monotone independent if

$\varphi(a_1 \dots a_n) = \varphi(a_i) \varphi(a_1 \dots a_{i-1} a_{i+1} \dots a_n)$  if  $a_i \in \mathcal{A}_{i_i}$  and  $\begin{cases} h_{j-1} < h_j < h_{j+1} \\ \text{or} \\ h_1 > h_2 \\ \vdots \\ h_{n-1} < h_n \end{cases} \quad (i \neq n+1)$

These independences appeared in some models.

They are basic in the sense that there are none others! [the crucial condition is associativity of independence]



Idea:  $\varphi((a+b)^n)$  is difficult to compute! Especially in the free case. Cumulants help. In classical probability, these are the coefficients of the log of the Fourier transform. Fourier  $\circ$  behave multiplicatively, log makes it additive.

Voiculescu's def<sup>n</sup> is based on Toeplitz operators. I don't understand well V's def.

For each  $a$ , define  $N.a := a^{(1)} + \dots + a^{(N)}$  where the  $a^{(i)}$  are copies of  $a$ , i.e.,  $\varphi((a^{(i)})^n) = \varphi(a^n)$ , that are independent in the considered sense in the corresponding extended p.u.c. take the free product of the probability spaces.

Lemma:  $\varphi((N.a)^n)$  is a polynomial in  $N$

Example: ( $n=1$ )  $\varphi(N.a) = \sum_{i=1}^N \varphi(a^{(i)}) = N\varphi(a)$ .

$$\begin{aligned} (n=2) \quad \varphi((N.a)^2) &= \sum_{i=1}^N \varphi((a^{(i)})^2) + \sum_{i \neq j} \varphi(a^{(i)} a^{(j)}) \\ &\quad \varphi(a^{(i)}) \varphi(a^{(j)}) \text{ for any independence} \\ &= N\varphi(a^2) + \frac{1}{2}N(N+1)\varphi(a)^2 \text{ that is, a polynomial in } N \end{aligned}$$

for  $n \geq 3$ , the result is independent of the type of independence.

The proof is done by induction on  $n$ .

Def:  $K_n(a) = \text{"coefficient of } N^1 \text{ in } \varphi((N.a)^n)"$  (i.e., the linear term)  
 $\rightarrow$  this linearity of  $K_n$  is expected.

Theorem ① For tensor cumulants  $K_n$ ,  $\varphi(a^n) = \sum_{\pi \in P(n)} K_{\pi}(a)$  where  $K_{\pi}(a) = \prod_{V \in \pi} K_{|V|}(a)$ .

② For free cumulants  $K_n$ ,  $\varphi(a^n) = \sum_{\pi \in \text{all partitions}} K_{\pi}(a)$

③ For Boolean cumulants  $K_n$ ,  $\varphi(a^n) = \sum_{\pi \in \mathcal{I}(n)} K_{\pi}(a)$

④ For monotone cumulants,  $\varphi(a^n) = \sum_{\pi \in \mathcal{M}(n)} \frac{1}{|\pi|!} K_{\pi}(a)$

$\rightarrow$  we have to think of some additional structure on  $\mathcal{M}(n)$ , which is very large!

Given  $\{V_1, V_2\}$ ,  $(V_1, V_2)$  and  $(V_2, V_1)$  are distinguished...

$\rightarrow$  Set of ordered partitions = faces of a Cayley graph (of the symmetric group).

