Piotr Śniady IMPAN

Lectures on random matrices and free probability theory

Lecture 7. Free cumulants 2

December 10, 2019

Plan

* free central limit theorem

* functional relation between R-tauform and moment - generaling function.

>[MN], Sed. 2.4.

Funtional Ilation

fix acy

$$M(z) = 1 + \sum_{n \geq 1}^{n} \varphi(a^{n}) z^{n}$$

$$C(z) = 1 + \sum_{n \geq 1} K(a, \dots, a) z^n$$

$$M(z) = C(zM(z))$$

$$[z^n] M(z) = \varphi(a^n) = \sum_{k=1}^{\infty} k_k =$$

$$= \sum_{i_{a},\dots,i_{3} \geqslant 0}$$

$$s \geqslant 1 \qquad s_{i_{1}},\dots,i_{n} = n$$

$$\sqrt{\int_{1}^{\infty} eNC(i, i)} \varphi(a^{i_1})$$

$$\begin{array}{c} \left(\sum_{\overline{u}_{1} \in NC(i_{n})}^{K_{\overline{u}_{1}}} \right) \cdot \cdot \cdot \cdot = \cdot \end{array}$$

S footons

$$= \begin{bmatrix} z^n \end{bmatrix}$$

$$\sum_{i_1,\dots,i_5 \geqslant 0} \sum_{i_4,\dots,i_5 = n} \sum_{i_5 \neq i_5 \neq i_5} \sum_{i_5 \neq i_5 \neq i_5 \neq i_5} \sum_{i_5 \neq i_5 \neq i_5 \neq i_5} \sum_{i_5 \neq i_5 \neq i_5 \neq i_5} \sum_{i_7 \neq i_7 \neq i_7 \neq i_7} \sum_{i_7 \neq i_7$$

$$= [2^h] \quad C(z M(z))$$

- - $K_n(x,\ldots,x) = \begin{cases} 0 \\ 1 \end{cases}$

$$C(2) = 1 + 2$$

 $H(z) = 1 + \left(z \, H(z)\right)^2$

 $z^{2} M(z)^{2} - M(z) + 1 = 0$ $M(z) = \frac{1 \pm \sqrt{1 - 4z^{2}}}{2z^{2}}$

M(z) = C(zM(z)

 $(1+x)^{\alpha} = \sum_{i \neq 0} (x^{\alpha}) x^{i}$

 $\sqrt{1-4z^2} = \sum_{i \geq 0}^{7} \left(\frac{A}{z}\right) (-4)^i z^{i} =$

 $= \int + \left(\frac{1}{4}\right)(-4) z^{2} +$

fix a e &

$$M(z) = 1 + \sum_{n \geq 1}^{n} \varphi(a^{n}) z^{n}$$

$$C(z) = 1 + \sum_{n>1}^{n} K(a_1, \ldots, a_n) z^n$$

$$M(z) = C(zM(z))$$

County transform
$$G(z) = \varphi\left(\frac{1}{z-a}\right) = \sum_{n \geq 0} \frac{\varphi(a^n)}{z^{n+1}} = \frac{1}{z} M(\frac{1}{z})$$
found power series in $\frac{1}{z}$

$$R(z) = \frac{C(z)-1}{2} = \sum_{n \geq 0}^{\infty} K_{n \in A}(a, \dots a) \cdot z$$

$$K(z) = R(z) + \frac{1}{2} = \frac{C(z)}{2}$$
 formal devices teries exact for $\frac{1}{2}$ H is a formal power in z

Then

$$K(G(z)) = \frac{1}{G(z)} C(G(z)) = \frac{1}{G(z)} C(\frac{1}{z} M(\frac{1}{z})) =$$

$$= \frac{1}{G(z)} M(\frac{1}{z}) = z$$

$$= \frac{1}{G(z)} M(\frac{1}{z}) =$$

Quich check if this all makes some
$$G(z) = \frac{1}{2} + \frac{?}{z^2} + \frac{?}{z^3} + \cdots$$

$$\frac{1}{G(z)} = 2 + ? + \frac{?}{z} + \frac{?}{z^2} + \dots =$$

2. (found series in
$$\frac{1}{2}$$
)

$$K(z) = \frac{1}{2} + \frac{9}{2} + \frac{9}{2} + \frac{2}{2} + \cdots$$

$$\frac{1}{K(x)} = Z + ? 2^2 + ? 2^3 = Z \cdot (pose xie) in Z$$

$$R(G(z)) = ? G(z) + ? G(z) + ? G(z) + ...$$

$$\omega_{a}(x) = x - R(6(0) = 2)$$

$$= x + ? + ? \frac{1}{x} + ? \frac{1}{x} \cdot \cdots$$

$$K(G(z)) = \frac{1}{G(z)} + ? + ? G(z) + ? G(z) + ...$$

=
$$Z + \left(\text{formal series in } \frac{1}{Z} \right)$$
.

$$G(K(2)) = \frac{1}{K(2)} + \frac{?}{K(2)^{2}} + \frac{?}{K(2)^{3}} + \dots =$$

$$= 2 + ? 2^{2} ? 2^{3} ! \dots$$

$$G(K(2)) = 2$$

$$G\left(R(z)+\frac{1}{z}\right)=2.$$

poor
$$S$$
 that
$$K_n(a+b) = K_n(a) + K_n(b)$$

 $\omega = \omega(z)$ $\omega : z \mapsto \omega(z)$

$$Z = G_{a+6}\left(R_{a+6}(z) + \frac{1}{2}\right) =$$

$$= G_{a+b} \left(R_a(z) + R_b(z) + \frac{1}{z} \right) =$$

$$= G_{\alpha} \left(R_{\alpha}(z) + \frac{1}{z} \right) =$$

$$= G_{\alpha} \left(\omega - R_{6}(z) \right) =$$

$$G_{a+b}(z) = G_a(\omega_a(z))$$

 $\omega_a(x) =$

= x- R6 (G..6(x))

Carb is SUBORDINATE to G. CONCEPTUAL EXPLANATION -> AHEAD

3.1 The Cauchy transform

Definition 1. Let $\mathbb{C}^+ = \{z \in \mathbb{C} \mid \operatorname{Im}(z) > 0\}$ denote the complex upper half-plane, and $\mathbb{C}^- = \{z \mid \operatorname{Im}(z) < 0\}$ denote the lower half-plane. Let v be a probability measure on \mathbb{R} and for $z \notin \mathbb{R}$ let

$$G(z) = \int_{\mathbb{R}} \frac{1}{z - t} \, d\mathbf{v}(t);$$

G is the Cauchy transform of the measure v.

Let us briefly check that the integral converges to an analytic function on \mathbb{C}^+ .

Lemma 2. *G* is an analytic function on \mathbb{C}^+ with range contained in \mathbb{C}^- .

Lemma 3. *Let G be the Cauchy transform of a probability measure* v. *Then:*

$$\lim_{y \to \infty} iy G(iy) = 1 \quad and \quad \sup_{y > 0, x \in \mathbb{R}} y |G(x + iy)| = 1.$$

Theorem 6. Suppose V is a probability measure on \mathbb{R} and G is its Cauchy transform. For a < b we have

$$-\lim_{y\to 0^+} \frac{1}{\pi} \int_a^b \operatorname{Im}(G(x+iy)) \, dx = v((a,b)) + \frac{1}{2}v(\{a,b\}).$$

If v_1 and v_2 are probability measures with $G_{v_1} = G_{v_2}$, then $v_1 = v_2$.

$$\operatorname{Im}(G(x+iy)) = \int_{\mathbb{R}} \operatorname{Im}\left(\frac{1}{x-t+iy}\right) dv(t) = \int_{\mathbb{R}} \frac{-y}{(x-t)^2 + y^2} dv(t).$$

Thus

$$\begin{split} \int_{a}^{b} \operatorname{Im}(G(x+iy)) \, dx &= \int_{\mathbb{R}} \int_{a}^{b} \frac{-y}{(x-t)^{2} + y^{2}} \, dx \, dv(t) \\ &= -\int_{\mathbb{R}} \int_{(a-t)/y}^{(b-t)/y} \frac{1}{1+\tilde{x}^{2}} \, d\tilde{x} \, dv(t) \\ &= -\int_{\mathbb{R}} \left[\tan^{-1} \left(\frac{b-t}{y} \right) - \tan^{-1} \left(\frac{a-t}{y} \right) \right] \, dv(t), \end{split}$$

where we have let $\tilde{x} = (x - t)/y$.

In the next two exercises we need to choose a branch of $\sqrt{z^2-4}$ for z in the upper half-plane, \mathbb{C}^+ . We write $z^2-4=(z-2)(z+2)$ and define each of $\sqrt{z-2}$ and $\sqrt{z+2}$ on \mathbb{C}^+ . For $z\in\mathbb{C}^+$, let θ_1 be the angle between the x-axis and the line joining z to 2; and θ_2 the angle between the x-axis and the line joining z to -2. See Fig. 3.1. Then $z-2=|z-2|e^{i\theta_1}$ and $z+2=|z+2|e^{i\theta_2}$ and so we define $\sqrt{z^2-4}$ to be $|z^2-4|^{1/2}e^{i(\theta_1+\theta_2)/2}$.

Exercise 2. For $z = u + iv \in \mathbb{C}$ let $\sqrt{z} = \sqrt{|z|}e^{i\theta/2}$ where $0 < \theta < \pi$ is the argument of z. Show that

$$\operatorname{Re}(\sqrt{z}) = \sqrt{\frac{\sqrt{u^2 + v^2} + u}{2}} \quad \text{and} \quad \operatorname{Im}(\sqrt{z}) = \sqrt{\frac{\sqrt{u^2 + v^2} - u}{2}}.$$

 $\varepsilon_n \rightarrow 0^+$ $\varepsilon_n \rightarrow 0^-$

This shows that v_1 and v_2 agree on all open intervals and thus are equal.

Example 7 (The semi-circle distribution).

As an example of Stieltjes inversion let us take a familiar example and calculate its Cauchy transform using a generating function and then using only the Cauchy transform find the density by using Stieltjes inversion. The density of the semi-circle law $v := \mu_s$ is given by

$$dv(t) = \frac{\sqrt{4-t^2}}{2\pi} dt$$
 on [-2,2];

and the moments are given by

$$m_n = \int_{-2}^2 t^n dv(t) = \begin{cases} 0, & n \text{ odd} \\ C_{n/2}, & n \text{ even} \end{cases}$$

where the C_n 's are the Catalan numbers:

$$C_n = \frac{1}{n+1} \binom{2n}{n}.$$

Now let M(z) be the moment-generating function

$$M(z) = 1 + C_1 z^2 + C_2 z^4 + \cdots$$

then

$$M(z)^2 = \sum_{m,n\geq 0} C_m C_n z^{2(m+n)} = \sum_{k\geq 0} \left(\sum_{m+n=k} C_m C_n\right) z^{2k}.$$

Now we saw in equation (2.5) that $\sum_{m+n=k} C_m C_n = C_{k+1}$, so

$$M(z)^{2} = \sum_{k \geq 0} C_{k+1} z^{2k} = \frac{1}{z^{2}} \sum_{k \geq 0} C_{k+1} z^{2(k+1)}$$

and therefore

$$z^{2}M(z)^{2} = M(z) - 1$$
 or $M(z) = 1 + z^{2}M(z)^{2}$.

By replacing M(z) by $z^{-1}G(1/z)$ we get that G satisfies the quadratic equation $zG(z) = 1 + G(z)^2$. Solving this we find that

$$G(z) = \frac{z \pm \sqrt{z^2 - 4}}{2}.$$

We use the branch of $\sqrt{z^2-4}$ defined before Exercise 2, however we must choose the sign in front of the square root. By Lemma 3, we require that $\lim_{y\to\infty}iyG(iy)=1$. Note that for y>0 we have that, using our definition, $\sqrt{(iy)^2-4}=i\sqrt{y^2+4}$. Thus

$$\lim_{y \to \infty} (iy) \frac{iy - \sqrt{(iy)^2 - 4}}{2} = 1$$

and

$$\lim_{y \to \infty} (iy) \frac{iy + \sqrt{(iy)^2 - 4}}{2} = \infty.$$

Hence

$$G(z) = \frac{z - \sqrt{z^2 - 4}}{2}.$$

Of course, this agrees with the result in Exercise 5.

Returning to the equation $zG(z) = 1 + G(z)^2$ we see that z = G(z) + 1/G(z), so K(z) = z + 1/z and thus R(z) = z i.e. all cumulants of the semi-circle law are 0 except κ_2 , which equals 1, something we observed already in Exercise 2.9.

Now let us apply Stieltjes inversion to G(z). We have

$$\operatorname{Im}\left(\sqrt{(x+iy)^2 - 4}\right) = \left|(x+iy)^2 - 4\right|^{1/2} \sin((\theta_1 + \theta_2)/2)$$

$$\lim_{y\to 0^+} \mathrm{Im}\left(\sqrt{(x+iy)^2-4}\right) = \begin{cases} |x^2-4|^{1/2}\cdot 0 = 0, & |x|>2\\ |x^2-4|^{1/2}\cdot 1 = \sqrt{4-x^2}, & |x|\leq 2 \end{cases}$$
 uniform over x in a corport subset of R .

and thus

$$\begin{split} \lim_{y \to 0^+} & \operatorname{Im}(G(x+iy)) = \lim_{y \to 0^+} \operatorname{Im}\left(\frac{x+iy-\sqrt{(x+iy)^2-4}}{2}\right) \\ &= \begin{cases} 0, & |x| > 2 \\ \frac{-\sqrt{4-x^2}}{2}, & |x| \leq 2 \end{cases}. \end{split}$$

Therefore

$$-\lim_{y\to 0^+} \frac{1}{\pi} \text{Im}(G(x+iy)) = \begin{cases} 0, & |x| > 2\\ \frac{\sqrt{4-x^2}}{2\pi}, & |x| \le 2 \end{cases}$$

Hence we recover our original density.

If G is the Cauchy transform of a probability measure we cannot in general expect G(z) to converge as z converges to $a \in \mathbb{R}$. It might be that $|G(z)| \to \infty$ as $z \to a$ or that G behaves as if it has an essential singularity at a. However (z-a)G(z) always has a limit as $z \rightarrow a$ if we take a *non-tangential* limit. Let us recall the definition. Suppose $f: \mathbb{C}^+ \to \mathbb{C}$ and $a \in \mathbb{R}$, we say $\lim_{\langle z \to a} f(z) = b$ if for every $\theta > 0$, $\lim_{z \to a} f(z) = b$ when we restrict z to be in the cone $\{x + iy \mid y > 0 \text{ and } |x - a| < \theta y\} \subset \mathbb{C}^+$.

Proposition 8. Suppose v is a probability measure on \mathbb{R} with Cauchy transform G. For all $a \in \mathbb{R}$ we have $\lim_{\leq z \to a} (z - a) G(z) = v(\{a\}).$

Proof: Let $\theta > 0$ be given. If z = x + iy and $|x - a| < \theta y$, then for $t \in \mathbb{R}$ we have

$$\left|\frac{z-a}{z-t}\right|^2 = \frac{(x-a)^2 + y^2}{(x-t)^2 + y^2} = \frac{1 + (\frac{x-a}{y})^2}{1 + (\frac{x-t}{y})^2} \le 1 + \left(\frac{x-a}{y}\right)^2 < 1 + \theta^2.$$

$$G(z) = \frac{z - \sqrt{z^2 - 4}}{2}.$$

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Therefore

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