#### Distribution of Polynomials in Free Variables

#### Roland Speicher

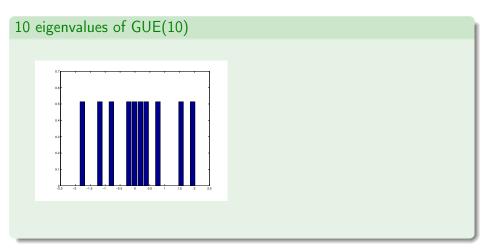
Saarland University Saarbrücken, Germany

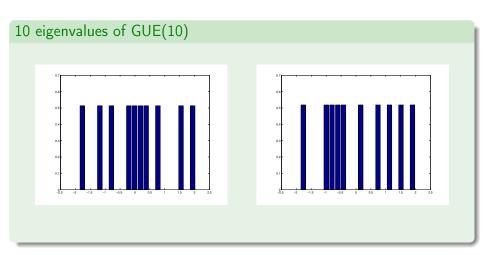
July 12, 2019

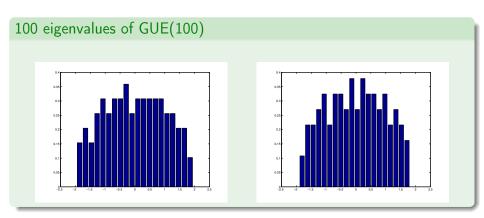
#### Wigner and Voiculescu

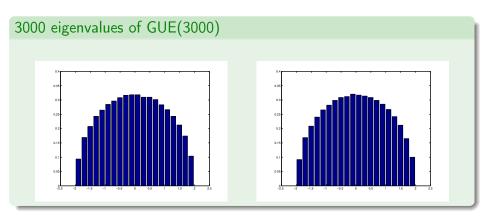












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- "randomly rotated" matrices are asymptotically free:
  - ▶ let  $D_1^{(N)}$ ,  $D_2^{(N)}$  be deterministic (e.g., diagonal) matrices
  - lacktriangle and  $U_N$  a Haar unitary random matrix,

then

$$X_N = U_N D_1^{(N)} U_N^*$$
 and  $Y_N = D_2^{(N)}$ 

are asymptotically free



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- The distribution of p(X,Y) is given by the Cauchy transform of a linearization

$$\hat{p}(X,Y) = b_0 \otimes 1 + b_1 \otimes X + b_2 \otimes Y$$

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 and the use of the subordination calculation of the latter as the sum of two free random variables

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- for this we have analytic theory of operator-valued free convolution
- this allows to calculate the operator-valued Cauchy transform of  $\hat{p}$ , and thus the Cauchy transform of p

#### How to calculate operator-valued free convolution

Theorem (Belinschi, Mai, Speicher 2013)

Consider

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#### How to calculate operator-valued free convolution

#### Theorem (Belinschi, Mai, Speicher 2013)

Consider  $\hat{p} = X + Y$  Then X and Y are free in the operator-valued sense and there exists a unique pair of (Fréchet-)holomorphic maps  $\omega_1, \omega_2: \mathbb{H}^+(\mathcal{B}) \to \mathbb{H}^+(\mathcal{B})$ , such that

$$G_X(\omega_1(b)) = G_Y(\omega_2(b)) = G_{X+Y}(b), \quad b \in \mathbb{H}^+(\mathcal{B}).$$

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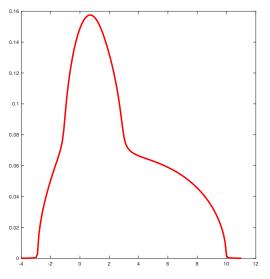
$$G_X(\omega_1(b)) = G_Y(\omega_2(b)) = G_{X+Y}(b), \quad b \in \mathbb{H}^+(\mathcal{B}).$$

Moreover,  $\omega_1$  and  $\omega_2$  can easily be calculated via the following fixed point iterations on  $\mathbb{H}^+(\mathcal{B})$ 

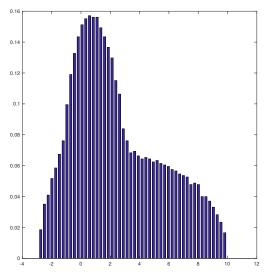
$$w \mapsto h_Y(b + h_X(w)) + b$$
 for  $\omega_1(b)$   
 $w \mapsto h_X(b + h_Y(w)) + b$  for  $\omega_2(b)$ 

where we put  $h_X(b) := G_X(b)^{-1} - b$  and  $h_Y(b) := G_Y(b)^{-1} - b$ ;

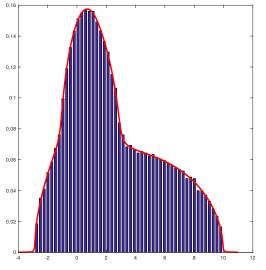
$$\mathcal{B} = M_3(\mathbb{C}), \qquad \mathbb{H}(\mathcal{B})^+ := \{ b \in \mathcal{B} \mid \Im b = \frac{b - b^*}{2i} > 0 \}$$



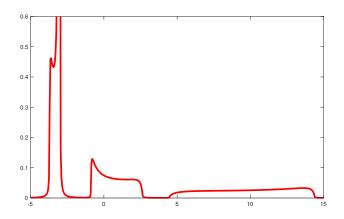
 $\bullet$  p(X,S) $\mu_X = \frac{1}{4}(2\delta_{-2} + \delta_{-1} + \delta_{+1})$  $\mu_S = \bar{\mathsf{semicircle}}$ 



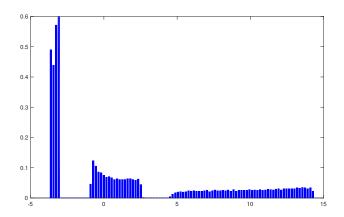
 $\bullet$   $p(X_N, A_N)$  $X_N = diag(-2, -2, -1, 1)$  $A_N$  GUE(N) N = 4000



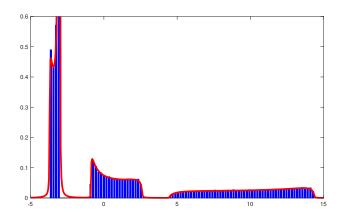
- $\begin{array}{l} \bullet \ p(X,S) \\ \mu_X = \frac{1}{4}(2\delta_{-2} + \delta_{-1} + \delta_{+1}) \\ \mu_S = \text{semicircle} \end{array}$
- $\begin{aligned} & \quad p(X_N,A_N) \\ & \quad X_N = \mathsf{diag}(-2,-2,-1,1) \\ & \quad A_N \; \mathsf{GUE}(\mathsf{N}) \\ & \quad \mathsf{N}{=}4000 \end{aligned}$



• 
$$p(X,Y)$$
  
 $\mu_X = \frac{1}{2}(\delta_1 + \delta_3)$   
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$$\begin{aligned} & \quad p(X_N,Y_N) \\ & \quad X_N = U \mathrm{diag}(1,3) U^* \\ & \quad Y_N = \mathrm{diag}(-2,-2,-1,1) \end{aligned}$$



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