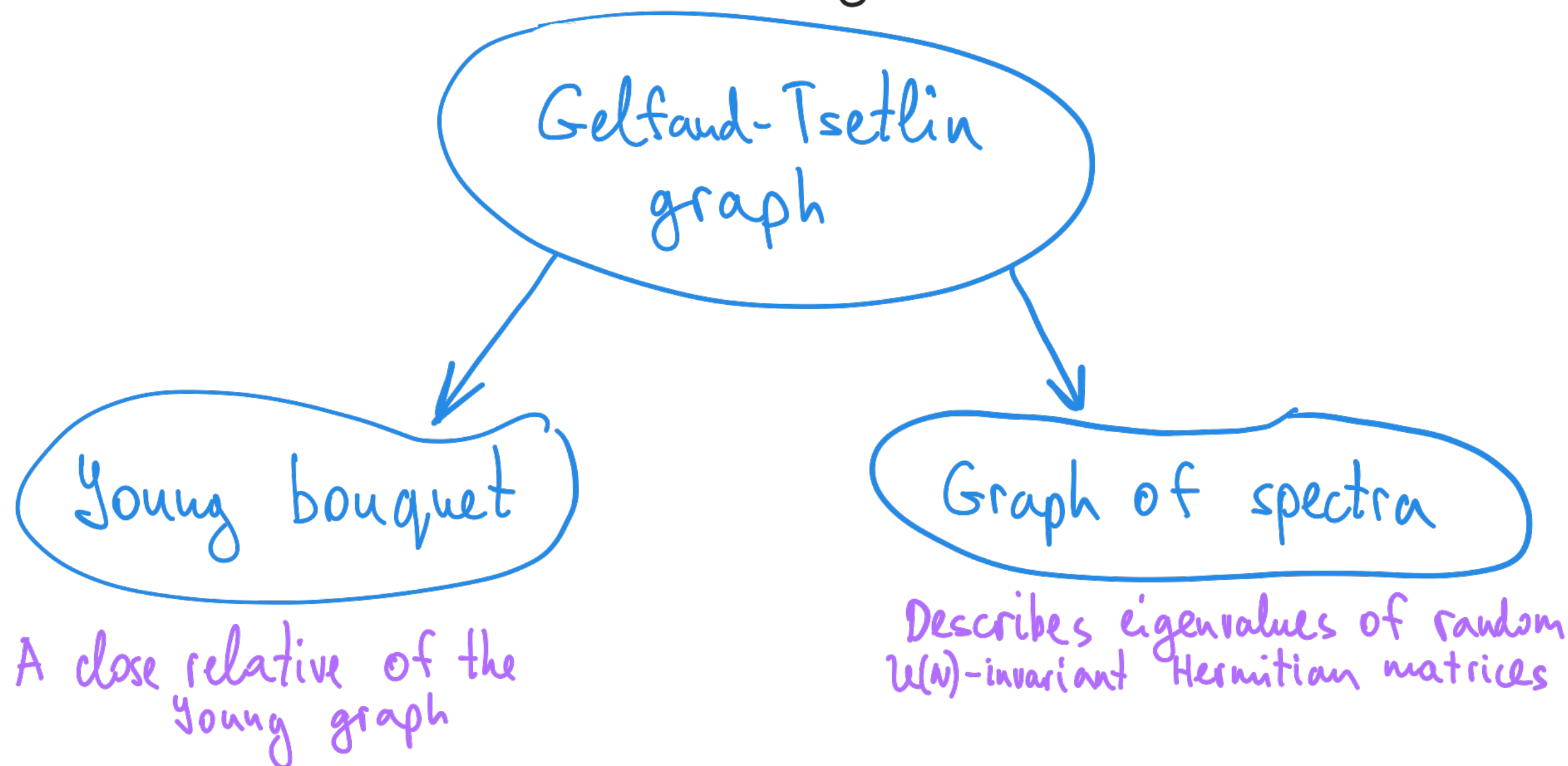


Math 833

Degenerations of the Gelfand-Tsetlin graph

Week 9

Today we connect GT-graph to Young graph and to random matrices through two limit transitions



We start from the Young graph.

Comparing Week 5, Theorem 2 with Week 8, Theorem 2 we see a remarkable similarity: the boundaries of GT and Y are described by almost identical formulas.

But why? The branching rules in these graphs look very different!

From representation-theoretic point of view, we discuss unitary groups $U(N)$ and symmetric groups $S(n)$.

Despite their different natures, the theories have similarities

- For both the character theory links to Schur functions
- Construction of bases and matrix realisations of irreducible reps can be done in a similar way (Gelfand-Tsetlin basis vs Young's orthogonal form)
- Both theories can be constructed from the same block: Schur-Weyl duality

Take two integers $N, n > 0$ and look at

This is a linear space of dimension N^n

$U(N)$ acts here: $u(v_1 \otimes v_2 \otimes \dots \otimes v_n) := (uv_1) \otimes (uv_2) \otimes \dots \otimes (uv_n)$

$S(n)$ also acts here: $\sigma(v_1 \otimes \dots \otimes v_n) := v_{\sigma^{-1}(1)} \otimes \dots \otimes v_{\sigma^{-1}(n)}$

These actions **commute**: $u \cdot \sigma(v_1 \otimes \dots \otimes v_n) = \sigma \cdot u(v_1 \otimes \dots \otimes v_n)$

Theorem 1: (Schur-Weyl duality)

$$(\mathbb{C}^N)^{\otimes n} = \bigoplus_{\lambda \in Y_n \cap GT_N} T_{\lambda}^{S(n)} \otimes T_{\lambda}^{U(N)}$$

$\nwarrow$ irred. rep of $S(n)$
 $\nearrow$ irred. rep of $U(N)$

in particular, $N^n = \sum_{\lambda \in Y_n \cap GT_N} \underset{\substack{\uparrow \\ \text{dimension in } Y}}{\dim(\lambda)} \cdot \underset{\substack{\uparrow \\ \text{dimension in } GT}}{\text{Dim}_N(\lambda)}$

We are not proving **Theorem 1**. Its important conceptual consequence is that irreducible representations of either $U(N)$ or $S(n)$ can be all constructed decomposing the same vector space $(\mathbb{C}^n)^{\otimes n}$.

[For $U(N)$ one also needs to be able to shift all coordinates to get negative λ_i , but this can be achieved by tensoring with 1-dimensional representations $u \rightarrow \det(u)^k$]

Hence, there is a link $U(N) \leftrightarrow S(n)$ already on finite level.

At $N=n=\infty$, the link is even tighter

Take $\lambda \in Y$ and large N , identify λ with an element of GT_N $(\lambda_1 \geq \lambda_2 \geq \dots \geq \lambda_n)$ [has many 0's]

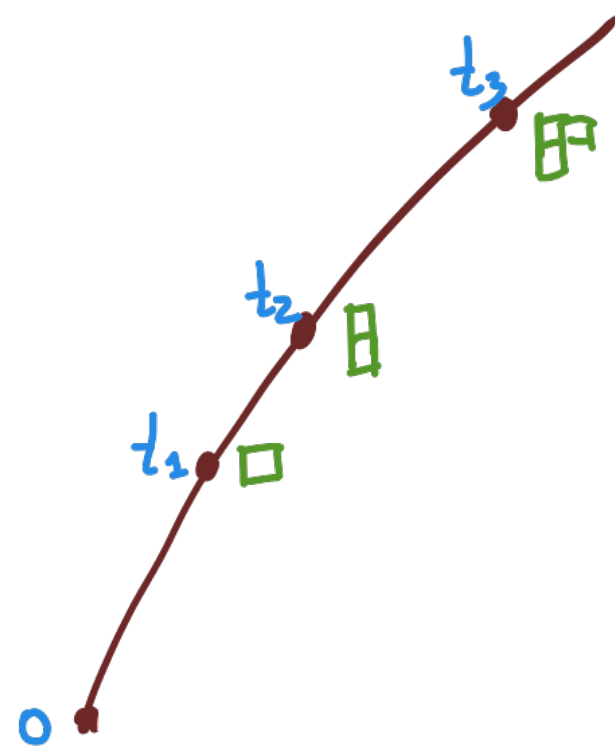
Question: How does a uniformly random path $\phi \rightarrow \lambda$ in GT look like?

Answer: This is a sequence of Young diagrams
 $\emptyset = \lambda^{(0)} \subset \lambda^{(1)} \subset \lambda^{(2)} \subset \lambda^{(3)} \subset \dots \subset \lambda^{(N)} = \lambda$, in which for most
 indices k we have $\lambda^{(k)} = \lambda^{(k+1)}$ and only for
 $n = |\lambda|$ of those we have $\lambda^{(k+1)} = \lambda^{(k)} + \square$
 $\left[\begin{array}{l} \lambda^{(k+1)} \supset \lambda^{(k)} \text{ guarantees } \lambda^{(k)} \subset \lambda^{(k+1)}. \text{ In principle, we could have} \\ |\lambda^{(k+1)} / \lambda^{(k)}| \geq 2, \text{ but probability of this event } \rightarrow 0 \text{ as } N \rightarrow \infty \end{array} \right]$

Conclusion: Positive part of GT (signatures with $\lambda_n \geq 0$) has
 a limit "small diagrams, large N ", in which we get
 the Young graph with an upgrade: boxes are added to
 a Young diagram not at every step, but at random times

The upgraded Young graph is called the Young bouquet

Paths in YB have the form



t_k = time when the k -th box is added

Stem with leaves (= Young diagrams).
Hence, the name Young bouquet

Definition: YB is a branching graph with $\mathbb{R}_{\geq 0}$ grading.
The r -th level of YB for $r \geq 0$ [real number!] is
the countable set of pairs (λ, r) [r is fixed in YB_r]
 λ - arbitrary Young diagram

The cotransition probabilities are defined for $r' < r$ by

$$\left(\lambda, r \right) \rightarrow \left(\mu, r' \right) \quad \begin{matrix} |\lambda| = n \\ |\mu| = m, \mu \subset \lambda \end{matrix}$$

$$\left(\frac{r'}{r} \right)^m \left(1 - \frac{r'}{r} \right)^{n-m} \frac{n!}{m! (n-m)!} \cdot \frac{\dim \mu \cdot \dim \lambda / \mu}{\dim \lambda}$$

Bimomial distribution Links in Young graph

In words, we transition from (r, λ) to label $r' < r$ in 2 steps

- Choose m from Binomial distribution on $\{0, 1, \dots, n\}$ with param. $\frac{r'}{r}$
- Choose μ with $|\mu| = m$ according to links $L_{n \rightarrow m}(\lambda \rightarrow \mu)$ from the Young graph.

Although the grading is now not by integers, but by real numbers, we can still define coherent systems and boundary

Definition: $\mu_r, r \geq 0$ is a coherent system on $\mathcal{Y}$ if each μ_r is a probability measure on Young diagrams λ and

$$\mu_{r'}(\mu) = \sum_{\lambda \in \mathcal{Y}} \left(\frac{r'}{r}\right)^{|\mu|} \left(1 - \frac{r'}{r}\right)^{n-|\mu|} \frac{n!}{|\mu|!(n-|\mu|)!} \cdot \frac{\dim \mu \cdot \dim \lambda / \mu}{\dim \lambda} \cdot \mu_r(\lambda)$$

for any $r' < r$ and $\mu \in \mathcal{Y}$

Theorem 2: The extreme coherent systems on YB are parameterized by $\alpha_1 \geq \alpha_2 \geq \dots \geq 0, \beta_1 \geq \dots \geq 0$ with $\sum_i (\alpha_i + \beta_i) \leq 1$ and $x \geq 0$ [when $x=0$, α_i and β_i parameters are not needed].

Explicitly they are given by:

$$\mathcal{M}_r^{(\alpha, \beta; x)}(\lambda) = e^{-rx} \frac{(rx)^{|\lambda|}}{|\lambda|!} \cdot \mathcal{M}_{|\lambda|}^{(\alpha, \beta)}(\lambda)$$

↑
extreme coherent system on Young graph

In words coherent systems on YB are **Poissonizations** of ones on Y

Theorem 3: $\mathcal{M}_r^{(\alpha, \beta; x)}(\lambda) = \lim_{N \rightarrow \infty} \mathcal{M}_N^{(\alpha^+ = \frac{x}{N}\alpha, \beta^+ = \frac{x}{N}\beta, \delta^+ = \frac{x}{N}(1 - \sum_i (\alpha_i + \beta_i)))(\lambda)$

↑
extreme coherent system on Gelfand-Tsetlin graph

and

$$\left(\frac{r'}{r}\right)^{|\mu|} \left(1 - \frac{r'}{r}\right)^{|\lambda| - |\mu|} \frac{|\lambda|!}{|\mu|!(|\lambda| - |\mu|)!} \cdot \frac{\dim \mu \cdot \dim \lambda / \mu}{\dim \lambda} = \lim_{N \rightarrow \infty} L_{Nr \rightarrow Nr'}^{(\text{link in GT})}(\lambda \rightarrow \mu)$$

We will not give proofs of Theorems 2 and 3
[They are not hard, feel free to figure out the proofs yourself]

They were established in:

Moscow Mathematical Journal

Volume 13, Issue 2, April–June 2013 pp. 193–232.

The Young Bouquet and its Boundary

Authors: Alexei Borodin (1) and Grigori Olshanski (2)

Example: Through Theorem 3 the measures $\mu_N^{(t^+ = t/N)}$ on GT
[Connected to TASEP, see slides 14–16 of Week 8]
converge as $N \rightarrow \infty$ to Poissonized Plancherel measure
on Young diagrams with $\text{Prob}(\lambda) = e^{-t} \frac{t^{|\lambda|}}{(|N|!)^2} \cdot [\dim(\lambda)]^2$
dimension in the Young graph

For the second continuous limit of GT we keep N fixed and make $\lambda_1 \gg \dots \gg \lambda_N$ very large

Proposition 1: Take real numbers $x_1 > y_1 > x_2 > \dots > x_N > y_N > x_{N+1}$

Then $\lim_{\varepsilon \rightarrow 0} \varepsilon^{-N} L_{(N+1) \rightarrow N} \left(L^{\varepsilon^{-1} x_1} \downarrow, \dots, L^{\varepsilon^{-1} x_{N+1}} \downarrow \rightarrow L^{\varepsilon^{-1} y_1} \downarrow, \dots, L^{\varepsilon^{-1} y_N} \downarrow \right)$

Cotransitional probability
(link) in Gelfand-Tsetlin
graph

$$N! \cdot \frac{\prod_{i < j \leq N} (y_i - y_j)}{\prod_{i < j \leq N+1} (x_i - x_j)}$$

Proof. By Week 8, Slide 3

$$L_{(N+1) \rightarrow N}(\lambda, \mu) = N! \cdot \frac{\prod_{i < j \leq N} (\mu_i - i - (\mu_j - j))}{\prod_{i < j \leq N+1} (\lambda_i - i - (\lambda_j - j))}$$

Remains to
take direct
limit



Definition: The graph of spectra Sp has levels

$$Sp_N = \{x_1 \geq x_2 \geq \dots \geq x_N \mid x_i \in \mathbb{R}\} \text{ and}$$

cotransition probabilities $L_{(N+1) \rightarrow N}((x_1, \dots, x_{N+1}) \rightarrow (y_1, \dots, y_N))$

given by

$$\begin{cases} N! \cdot \frac{\prod (x_i - y_j)}{\prod (x_i - x_j)} dy_1 dy_2 \dots dy_N, & \text{if } x_1 \geq y_1 \geq x_2 \geq \dots \geq y_N \geq x_N \\ 0, & \text{otherwise.} \end{cases}$$

[This is for the case $x_1 \geq \dots \geq x_{N+1}$. If some coordinates coincide, then we extend the definition by continuity in the space of measure. For instance, if $x_1 = \dots = x_{N+1} = 0$, then we get the unit mass (δ -measure) at $y_1 = \dots = y_N = 0$.]

Same as GT , but with integers replaced by reals.

Paths in the graph of spectra
are interlacing arrays
of reals



By taking the limit from Gelfand-Tsetlin graph, the centrality property (= coherency) says that:

- Given $(x_1^N \geq x_2^N \geq \dots \geq x_N^N) \in \text{Sp}_N$ the conditional law of subarray $\{x_i^j\}_{1 \leq i \leq j \leq N-1}$ is **uniform** on the polytope defined by interlacing inequalities

- The dimension in Sp is the **volume** of this polytope

$$\text{Dim}_N(x_1, \dots, x_N) = \text{Volume} = \prod_{1 \leq i < j \leq N} \frac{x_i - x_j}{j - i}$$

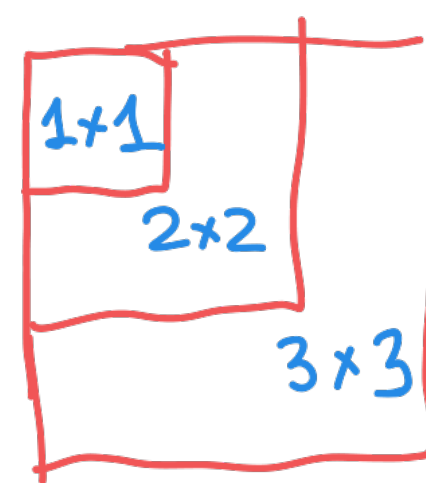
[This is the limit of Theorem 1 on Slide 3 of week 8]

Why is this called the graph of spectra?

Theorem 4: Take $N \times N$ random Hermitian matrix, whose law is invariant under conjugations $A \rightarrow U A U^*$, $U \in U(N)$.

[Conjugations preserve being Hermitian and eigenvalues. It only changes eigenvectors. Hence, in conjugation-invariant matrix, the eigenvectors are uniformly chosen (conditional on eigenvalues)]

Let x_i^k be the i th largest eigenvalue of top-left $k \times k$ corner



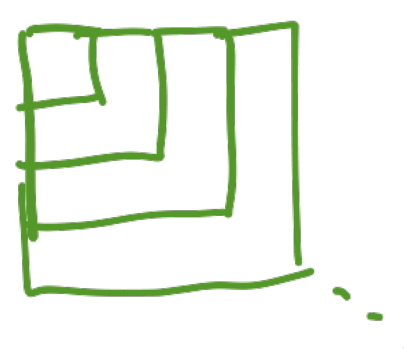
Then:

- Eigenvalues **interlace**: $x_i^{k+1} \geq x_i^k \geq x_{i+1}^{k+1}$ for all $1 \leq i \leq k < N$
- Given $(x_1^N, x_2^N, \dots, x_N^N)$ the conditional law of $\{x_i^k\}_{1 \leq i \leq k \leq N}$ is **uniform** on the polytope of interlacing inequalities

Proof of Theorem 4 —
— next week

Using **Theorem 4** the coherent systems on S_p are random **infinite** Hermitian matrices invariant under conjugations by $u \in \mathcal{U}(\infty) = \bigcup_{N \geq 1} \mathcal{U}(N)$

The bijection is as follows:

infinite Hermitian matrix A
 $\updownarrow$
laws of all its corners A_N 

$\updownarrow$
laws of all eigenvalues $(x_1^N \geq x_2^N \geq \dots \geq x_N^N)$ of A_N

Hence, boundary of S_p is in bijection with laws of **ergodic $\mathcal{U}(\infty)$ -invariant random $\infty \times \infty$ Hermitian matrices.**

Theorem 5: Extreme coherent systems on S_p are parameterized by $\gamma_1 \in \mathbb{R}$, $\gamma_2 \geq 0$, (d_j) with $\sum (d_j)^2 < \infty$

The projection on the first level, $\mu_1^{(d_j; \gamma_1, \gamma_2)}(x)$ has characteristic function

$$E_{\mu_1^{(d_j; \gamma_1, \gamma_2)}} e^{itx} = \Phi^{(d_j; \gamma_1, \gamma_2)}(t) = e^{i\gamma_1 t} e^{-\gamma_2 \frac{t^2}{2}} \prod_{j \geq 1} \frac{e^{-itd_j}}{1 - itd_j}$$

shift
Gaussian random variable
Shifted exponential random variable

More generally, the random matrix A_N has the law such that

$$E \exp(i \text{Trace}(T A_N)) = \Phi^{(d_j; \gamma_1, \gamma_2)}(t_1) \Phi^{(d_j; \gamma_1, \gamma_2)}(t_2) \dots \Phi^{(d_j; \gamma_1, \gamma_2)}(t_n)$$

arbitrary Hermitian matrix
eigenvalues of T

Further, denoting $(x_1^N \geq x_2^N \geq \dots \geq x_n^N)$ the eigenvalues of A_N , the LLN

holds: $\lim_{N \rightarrow \infty} \frac{x_j^N}{N} = d_j^+$; $\lim_{N \rightarrow \infty} \frac{x_{N-j}^N}{N} = d_j^-$; $\lim_{N \rightarrow \infty} \frac{x_1^N + x_2^N + \dots + x_N^N}{N} = \gamma_1$; $\lim_{N \rightarrow \infty} \sum_{j=1}^N \left(\frac{x_j^N}{N}\right)^2 = \gamma_2 + \sum_{j=1}^{\infty} (d_j)^2$

$\{d_j\} = \{d_j^+\} \cup \{d_j^-\}$

Example 1: Fix a (deterministic) number $q \in \mathbb{R}$ and consider a matrix

$$\begin{pmatrix} q & & & \\ & q & & \\ & & 0 & \dots \\ & & & \ddots \\ 0 & & & & 0 \\ \vdots & & & & & \ddots \end{pmatrix}$$

It is invariant under $U(\infty)$ -conjugations and corresponds to measure with $\gamma_1 = q$

Example 2: Let X be infinite matrix with i.i.d. $N(0,1) + i N(0,1)$ matrix elements. Consider $\frac{c}{2} (X + X^*)$

Invariance of i.i.d. Gaussian sequences under rotations implies invariance under $U(\infty)$ -conjugations. Corresponds to the measure with $\gamma_2 = c^2$

This is the **Gaussian Unitary Ensemble** of random matrices

It represents a wider class of **Wigner matrices** with i.i.d. (modulo symmetry) matrix elements.

Example 3: Take a vector V with i.i.d. $N(0,1) + iN(0,1)$ coordinates and consider the matrix $\frac{a}{2} V^* V$. Due to invariance of the i.i.d. Gaussian sequence under rotations this matrix is invariant under $U(\infty)$ -conjugations. It corresponds to the parameters $\delta_1 = \gamma_1 = a$.

This is the simplest example of **Wishart** or **sample covariance** random matrix. Also sometimes called **Laguerre Unitary Ensemble**

Exercise: Check that the LLN claimed in **Theorem 5** holds for the random matrices of **Example 3**

Proposition 2: All measures of **Theorem 5** are sums of **independent** matrices from **Examples 1, 2, and 3**.

Proof: When we add independent random variables, their characteristic functions are multiplied. This is precisely the multiplicative form of the function $\Phi^{(\pm j; \gamma_1, \gamma_2)}(t)$.

We prove **Theorem 5** next week. For now, some connections:

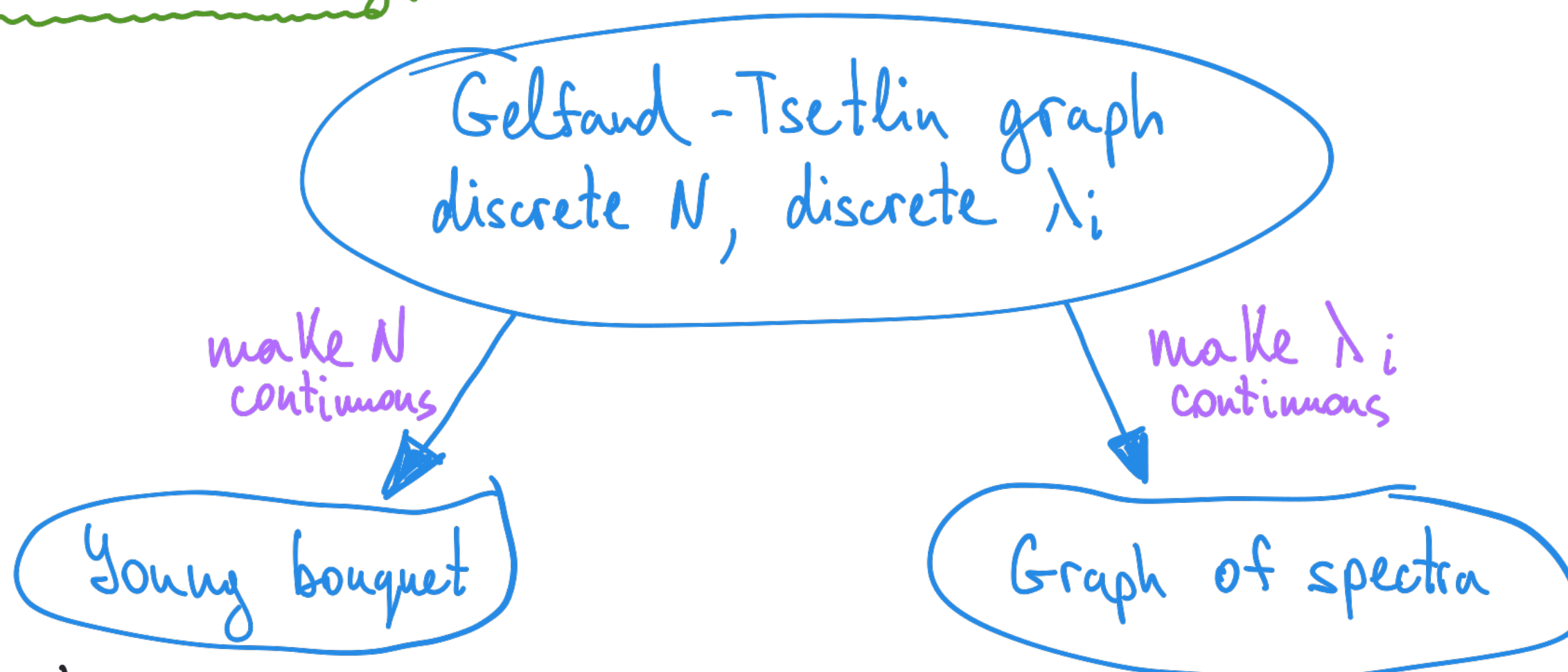
- γ_1 is the limit of $\gamma^\pm$ in the Gelfand-Tsetlin graph.
 δ_i are limits of $\delta_i^\pm$ in GT. However, $\beta^\pm$ disappeared and new parameter γ_2 appeared in the limit transition

- $\left(\Phi^{(\delta_i, \gamma_1, \gamma_2)} \left(\frac{z}{i} \right) \right)^{-1} = \Phi_{\delta, \gamma}(z)$
Theorem 5 Week 2, Slide 1, Theorem 1

Also compare Proposition 2 with Week 2, Slide 19, Theorem 5
An explanation is coming in two weeks!

-
- The connection to total positivity, which we saw for Y and GT, still exists for Sp . This time we need to deal with totally positive functions of real argument instead of sequences. They are called **Pólya frequency functions**.

Conceptual summary:



Reduction of representation theory of $U(N)$ to the one of $S(n)$

"Semiclassical limit"

Reduction of Lie group $U(N)$ to its Lie algebra (tangent space at identity), which can be identified with $(s\text{ken})$ -Hermitian matrices.