

# Positivity everywhere

## Lecture 1

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TA : Slim Kammoun ( ENS  $\mapsto$  Poitiers )

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Rough format:

Lectures — bird's eye view

Tutorials — work out the details

Many interconnected themes: choose your own adventure!

Good to keep in mind:

- look to understand (familiar) ideas from multiple perspectives
- look for interesting juxtapositions

## Table of Contents

- (1) Positivity — in what sense?
- (2) The moment sequence's toolkit
- (3) Fruits & gifts & many open questions

Part I : Positivity (non-negativity) everywhere

Algebraic / combinatorial object  $\leadsto$  Representing matrix (?)  $\sim$  Positivity of the matrix (?)

$$[a_{ij}] \quad \text{s.t.} \quad a_{ij} \geq 0 \quad \forall i,j$$

(Maybe  $k \times m$ ,  
maybe infinite)

$$A = [a_{ij}] = [\bar{a}_{j,i}] \quad \text{s.t.} \quad \langle Av, v \rangle \geq 0 \quad \forall v$$

$$\det[a_{ij}]_{i,j \in I} \geq 0 \quad \forall I$$

Principal minors

$$\det[a_{ij}]_{1 \leq i,j \leq n} \geq 0 \quad \forall n$$

Principal Leading minors

$$\det[a_{ij}]_{i \in I, j \in J} \geq 0 \quad \forall |I| = |J|$$

All minors

$$\det[a_{ij}]_{i \in I, j \in J} \geq 0 \quad \forall |I| = |J| = 2$$

2x2 minors

$$\det[a_{ij}]_{i \in I, j \in J} \geq 0 \quad \forall |I| = |J| = \text{rank}$$

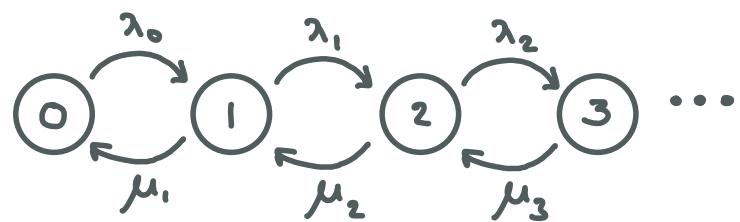
Maximal minors

Notation:  $[n] := \{1, 2, \dots, n\}$

Def A matrix is **totally positive** if all of its minors are non-negative.

### Example 1

Birth-death processes



$$p_{j,k}(t) = P(X(t) = k \mid X(0) = j)$$

$$p_{k,k+1}(\Delta t) = \lambda_k \Delta t + o(\Delta t)$$

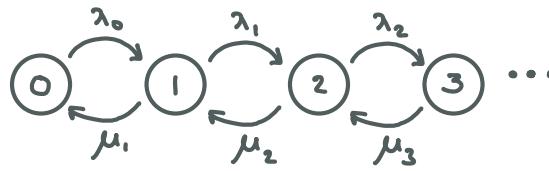
$$p_{k,k-1}(\Delta t) = \mu_k \Delta t + o(\Delta t)$$

$$p_{k,k}(\Delta t) = 1 - (\lambda_k + \mu_k)\Delta t + o(\Delta t)$$

Thm (Karlin & McGregor '57) For any  $t > 0$ ,

$[p_{j,k}(t)]_{j,k \geq 0}$  is totally positive.

Proof sketch #1 (KM'57, §5):



$$P_{k,k+1}(\Delta t) = \lambda_k \Delta t + o(\Delta t)$$

$$P_{k,k-1}(\Delta t) = \mu_k \Delta t + o(\Delta t)$$

$$P_{k,k}(\Delta t) = 1 - (\lambda_k + \mu_k) \Delta t + o(\Delta t)$$

$$P_n(t) = [P_{j,k}(t)]_{0 \leq i,j \leq n}$$

$$P_n(0) = I_n$$

$$\frac{d}{dt} P_n(t) = A_n P_n(t)$$

where  $A_n = \begin{bmatrix} -(\lambda_0 + \mu_0) & \lambda_0 & & & \\ \mu_1 & -(\lambda_1 + \mu_1) & \lambda_1 & & \\ & \mu_2 & -(\lambda_2 + \mu_2) & \dots & \\ & & & \ddots & \\ & & & & -(\lambda_{n-1} + \mu_{n-1}) \end{bmatrix}$

Observe  $I + \frac{t}{n} A_n$  is TP for  $n$  large enough.

Deduce  $e^{tA_n}$  is TP  $\forall t > 0$ .

Some further questions you could ask:

- Which solutions  $P$  satisfying the backward equation

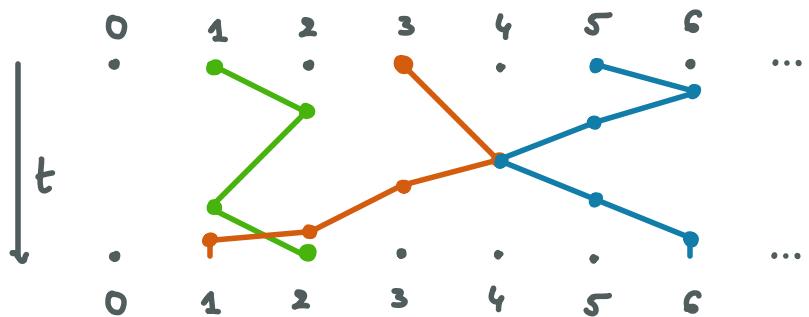
$$\frac{d}{dt} P(t) = A P(t) \text{ and the fwd eg } \frac{d}{dt} P(t) = P(t) A$$

$P(0) = I$  are transition matrices of some Markov process?

- If we start with a time-differentiable matrix of probabilities and if the matrix has "enough positivity", will its time derivative be of the tridiagonal form?

(See KM '57)

Proof sketch #2 (KM '59):  $n$  particles executing the birth-death process



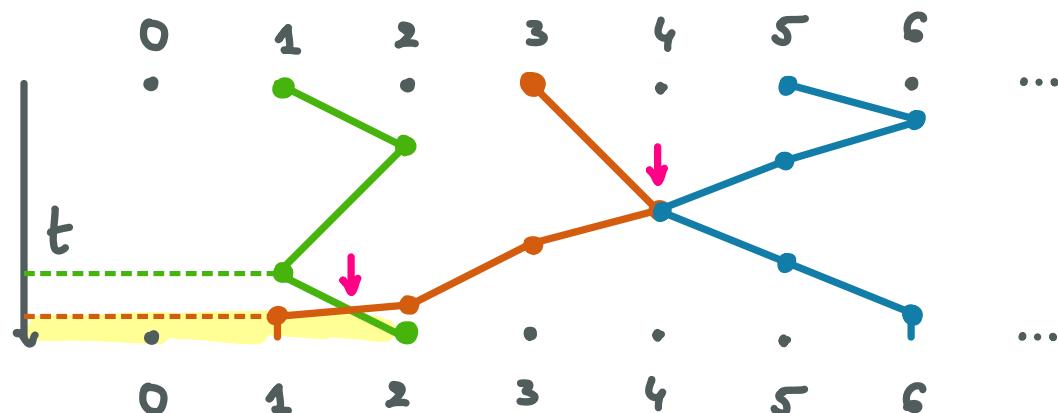
$$\det \left[ p_{ij}(t) ; \begin{matrix} i_1 < i_2 < \dots < i_n \\ j_1 < j_2 < \dots < j_n \end{matrix} \right] = \sum_{\sigma \in S_n} (-1)^{\text{inv}(\sigma)} p_{i_1, j_{\sigma(1)}}(t) p_{i_2, j_{\sigma(2)}}(t) \dots p_{i_n, j_{\sigma(n)}}(t)$$

Claim:

$$\det \left[ p_{ij}(t) ; \begin{matrix} i_1 < i_2 < \dots < i_n \\ j_1 < j_2 < \dots < j_n \end{matrix} \right] = \text{prob} (\text{at time } t, \text{ particles found in } j_1, j_2, \dots, j_n \text{ without having coincided in any state})$$

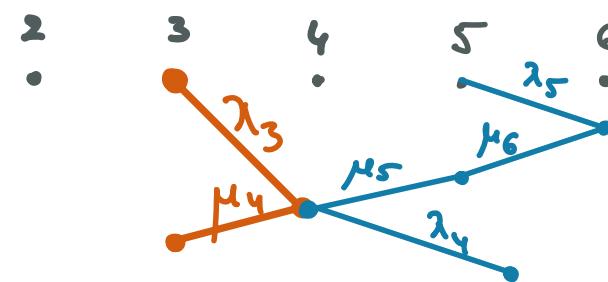
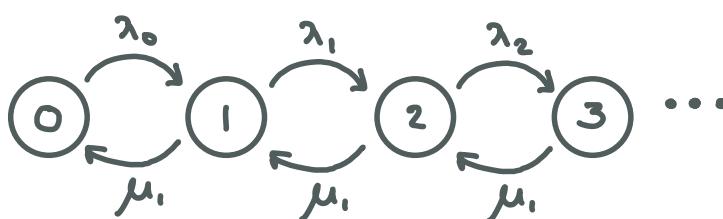
$$\det \left[ p_{ij}(t) : i_1 < i_2 < \dots < i_m \atop j_1 < j_2 < \dots < j_n \right] = \sum_{\sigma \in S_m} (-1)^{\text{inv}(\sigma)} p_{i_1, j_{\sigma(1)}}(t) \dots p_{i_m, j_{\sigma(n)}}(t) =$$

prob (at time  $t$ , particles found in  $j_1, j_2, \dots, j_n$  resp. without having coincided in any state)

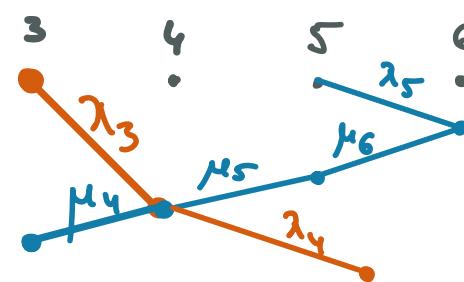


Proof by example:

$p_{ij}(t)$  computed from



vs



same overall probability but opposite signs

Some further questions you could ask:

- Where did we use the fact that the process is birth-death?
- What about a general Markov process on  $\mathbb{N}_0$ .

Exercise : Write down formula for the determinant

- Did we need to have probabilities, or even positive weights?

Independently: Gessel-Viennot '85 based on Lindström '73

$G = (V, E)$  locally finite edge-weighted directed acyclic graph,

$A = \{a_1, \dots, a_n\}, B = \{b_1, \dots, b_n\} \subseteq V$  (need not be disjoint)

weight of a path  $p = \text{wt}(p) =$  product of edge weights

$W = [w_{ij}]_{1 \leq i, j \leq n}$  where  $w_{ij} = \sum_{\substack{\text{paths } p \\ a_i \mapsto b_j}} \text{wt}(p)$

Lemma (LGV)

$$\det W = \sum_{\sigma \in S_n} (-1)^{\text{inv}(\sigma)} \sum_{\substack{\text{vertex-disjoint paths} \\ p_1 : a_i \mapsto b_{\sigma(i)} \\ \vdots \\ p_n : a_n \mapsto b_{\sigma(n)}}} \text{wt}(p_1) \cdots \text{wt}(p_n)$$

Proof :  
same idea (exercise)



## Example 2

Matroids: unify several notions of independence

Def Matroid  $M = (E, \mathcal{B})$ , where  $E$  is a finite set ("ground set")

and  $\mathcal{B} \subseteq 2^E$  ("bases of  $M$ "),  $\mathcal{B} \neq \emptyset$ , s.t.  $\forall B_1, B_2 \in \mathcal{B}$

and  $b_1 \in B_1 - B_2$ ,  $\exists b_2 \in B_2 - B_1$  w/  $(B_1 - \{b_1\}) \cup \{b_2\} \in \mathcal{B}$ .

("basis exchange axiom")

E.g.  $A \in \text{Mat}_{d \times n}(K)$ ,  $\text{rank}(A) = d$ ,  $A = (a_1, a_2, \dots, a_n)$ .

let  $\mathcal{B} = \{B \subseteq [n] \mid \{a_i\}_{i \in B} \text{ form a linear basis for } K^d\}$ .

Check:  $M(A) = ([n], \mathcal{B})$  is a matroid.

Such a matroid is **representable**.

Def. (Postnikov) **Positroid** : a matroid on  $[n]$  representable  
by columns of a real matrix,  
whose maximal minors are non-negative.

Positroids  $\longleftrightarrow$  Decorated permutations e.g. 01536427

$\longleftrightarrow$  Grassmann necklaces

$\longleftrightarrow$  J-diagrams

$\longleftrightarrow$  equiv. class. of plabic graphs

(Postnikov '06)  
(Oh '11)

Many matroidal properties: closure properties, duality

### Example 3 (Subtler)

From now on :  $(a_n)_{n \geq 0}$  denotes a real sequence

Def  $(a_n)_n$  is **unimodal** if  $a_0 \leq a_1 \leq \dots \leq a_k \geq a_{k+1} \geq \dots$   
for some  $0 \leq k \leq n$ .

Def  $(a_n)$  is **log-concave** if  $a_k^2 \geq a_{k-1} a_{k+1} \forall k$

Def  $(a_n)$  is **log-convex** if  $a_k^2 \leq a_{k-1} a_{k+1} \forall k$

Example:  $\binom{n}{0}, \binom{n}{1}, \binom{n}{2}, \dots, \binom{n}{n-1}, \binom{n}{n}$

- unimodal

- log-concave: 
$$\frac{\binom{n}{k}^2}{\binom{n}{k-1} \binom{n}{k+1}} = \frac{(n-k+1)(k+1)}{(n-k)k} > 1$$

Exercise: if  $a_n > 0 \forall n$ , log-concavity  $\Rightarrow$  Unimodality.

Notation:  $[n]_g := \frac{1-g^n}{1-g} = 1+g+\cdots+g^{n-1}$  with  $[0]_g := 0$

$$[n]_g ! := [n]_g [n-1]_g \cdots [2]_g [1]_g \text{ with } [0]_g ! := 1$$

Exercise : Show that

$$\begin{bmatrix} n \\ 0 \end{bmatrix}_g, \begin{bmatrix} n \\ 1 \end{bmatrix}_g, \dots, \begin{bmatrix} n \\ k \end{bmatrix}_g := \frac{[n]_g !}{[n-k]_g ! [k]_g !}, \dots, \begin{bmatrix} n \\ n \end{bmatrix}_g$$

is log-concave for  $g \geq 0$ .

Thm (Huh '09) Consider a matroid  $M$  representable over a field of characteristic 0 with characteristic polynomial

$$X_M(q) = \mu_0 q^{r+1} - \mu_1 q^r + \cdots + (-1)^{r+1} \mu_{r+1},$$

The sequence  $\mu_0, \dots, \mu_{r+1}$  is **log-concave**.

Proves a conjecture of Read ('68) that chromatic polynomials of graphs are unimodal. More generally:

(ex.) Conj (Rota, Heron, Walsh ~'70)

Thm Adiprasito, Huh, Katz '15

Coefficients of the characteristic polynomial of any finite matroid form a log-concave sequence.

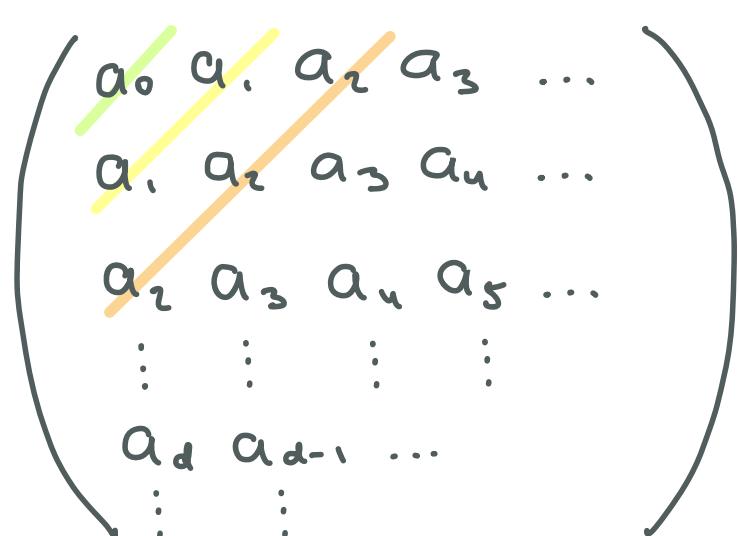
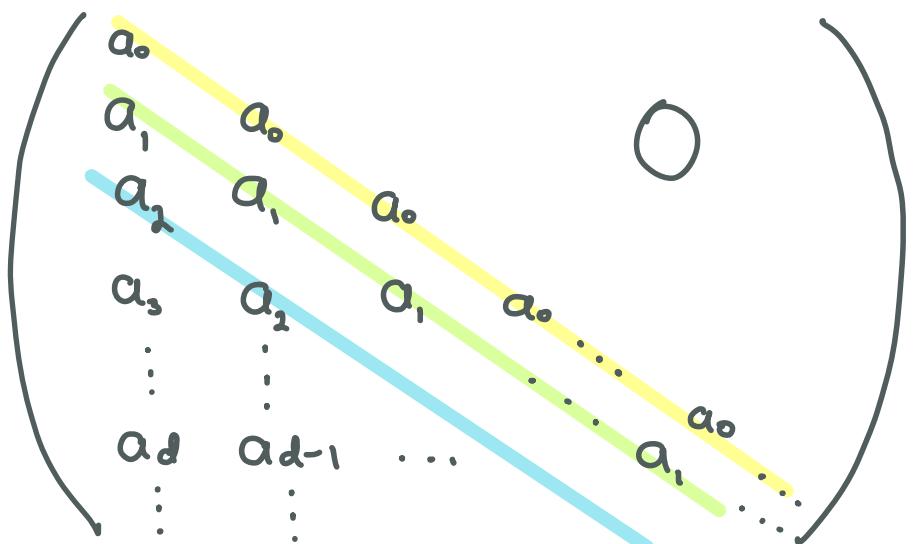
## Toepplitz and Hankel matrices

Dcf To a real seq.  $(a_n)_{n \geq 0}$ , we can associate the infinite

Toepplitz matrix

and

Hankel matrix.



$$T(a) = \begin{pmatrix} 1 & & & & & \\ a_1 & 1 & & & & \\ a_2 & a_1 & 1 & & & \\ a_3 & a_2 & a_1 & 1 & & \\ \vdots & \vdots & & \ddots & & \\ a_d & a_{d-1} & \dots & & 1 & \\ \vdots & \vdots & & & & \ddots \end{pmatrix}$$

Thm ( Schoenberg, Aissen-Schoenberg-Whitney, Edrei '48-'53  
 See also Thoma '64)

$T(a)$  is totally positive iff in some nbhd of  $z=0$

$$1 + a_1 z + a_2 z^2 + \dots = e^{z\alpha} \prod_{i \geq 0} \frac{(1 + \beta_i z)}{(1 - \gamma_i z)}$$

for  $\alpha \geq 0$ ,  $\beta_1 \geq \beta_2 \geq \dots \geq 0$ ,  $\gamma_1 \geq \gamma_2 \geq \dots \geq 0$

with  $\sum_i \beta_i + \sum_i \gamma_i < \infty$ .

Rietsch '01: parametrization of  $n \times n$  totally positive Toeplitz

$$T(a) = \left( \begin{array}{cccccc} a_0 & & & & & \\ a_1 & a_0 & & & & \\ a_2 & a_1 & a_0 & & & \\ a_3 & a_2 & a_1 & a_0 & & \\ \vdots & \vdots & \vdots & \ddots & \ddots & \ddots \\ a_d & a_{d-1} & \dots & & & \end{array} \right)$$

Observe:

$T(a)$  totally positive  
 $\Rightarrow a_0, a_1, \dots$  is log-concave

E.g.  $\binom{n}{0}, \dots, \binom{n}{k}, \dots, \binom{n}{n}$

$\langle \binom{n}{0} \rangle, \dots, \langle \binom{n}{k} \rangle, \dots, \langle \binom{n}{n} \rangle$  Eulerian #'s, e.g.  $\sigma \in S_n$  w/ k ascents

$[n]_0, \dots, [n]_k, \dots, [n]_n$  stirling #1, e.g.  $\sigma \in S_n$  w/ k cycles

$\{n\}_0, \dots, \{n\}_k, \dots, \{n\}_n$  stirling #2, e.g. partitions of  $[n]$  into k parts

What about log convexity?

Some log-convex sequences:  $a_k^2 \leq a_{k-1} a_{k+1}$

$n!$  e.g. permutations

$B_n$  Bell #'s e.g. set partitions

$C_n$  Catalan #'s e.g. non-crossing set partitions

But also:

$$\left( \sum_{k=0}^n \binom{n}{k} x^k \right)_{n \geq 0} \quad \forall x \in \mathbb{R} \quad (\text{trivial})$$

Eulerian

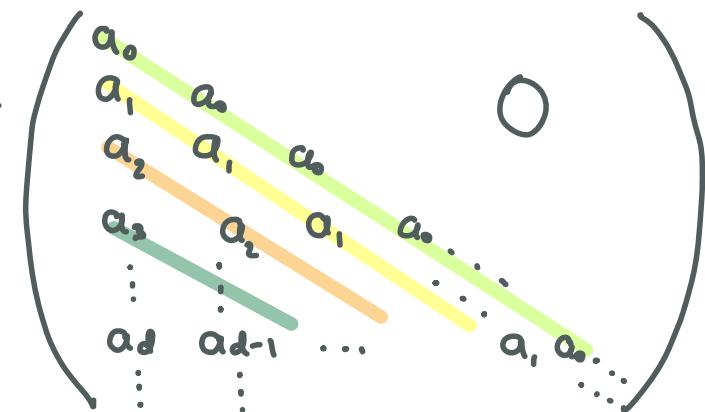
$$\left( \sum_{k=0}^n \langle \binom{n}{k} \rangle x^k \right)_{n \geq 0} \quad (x \geq 0)$$

Stirling II

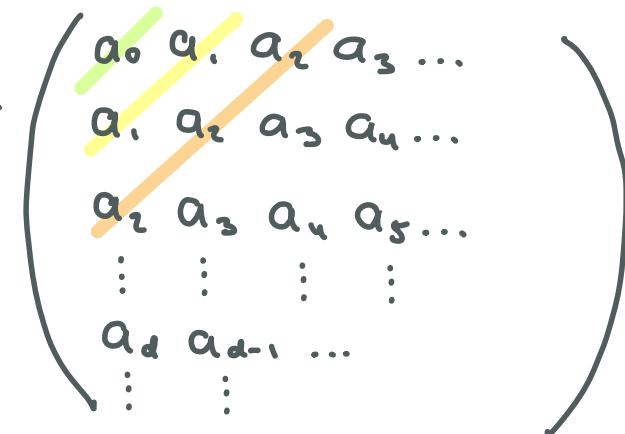
$$\left( \sum_{k=0}^n \{ \binom{n}{k} \} x^k \right)_{n \geq 0} \quad (x \geq 0)$$

Def A real sequence  $(a_n)_{n \geq 0}$  is

- Toeplitz totally positive if  $T(a) =$   
is totally positive.



- Hankel totally positive if  $H(a) =$   
is totally positive.



Exercise:

Toeplitz TP  $\Rightarrow$  log-concave

Hankel TP  $\Rightarrow$  log-convex

Take your favorite combinatorial sequence  $(a_n)$ .

(From now on,  $a_0 = 1$ )

When does there exist a probability measure

$\mu$  on  $\mathbb{R}$  s.t.

$$a_n = \int_{\mathbb{R}} x^n d\mu(x) \quad \forall n \in \mathbb{N} ?$$

Thm (Hamburger 1920-21)

$(a_n)_{n \geq 0}$  is a moment sequence of a positive Borel measure  $\mu$  on  $\mathbb{R}$  iff  $\forall k \in \mathbb{N}, \forall z_0, z_1, \dots, z_k \in \mathbb{C}$ ,

$$\sum_{j, e=0}^k a_{j+e} z_j \bar{z}_e \geq 0, \quad \text{i.e. the Hankel matrices } [a_{i+j}]_{i,j \leq n} \text{ are positive semidefinite } \forall n.$$

<proof>

$$(\Rightarrow) \quad \sum_{j, e=0}^k a_{j+e} z_j \bar{z}_e = \int_{\mathbb{R}} \underbrace{\left| \sum_{j=0}^k z_j x^j \right|^2}_{\geq 0} d\mu(x) \geq 0$$

(<) Subsequent lecture

Thm (Stieltjes 1894-95, Gantmakher-Krein 1937) TFAE:

(1)  $\exists \mu \geq 0$  on  $[0, \infty)$  s.t.  $a_n = \int\limits_{[0, \infty)} x^n d\mu(x)$

(2) The infinite Hankel matrix  $\begin{bmatrix} a_0 & a_1 & a_2 & \dots \\ a_1 & a_2 & a_3 & \dots \\ a_2 & a_3 & \dots \\ \vdots & & & \end{bmatrix}$  is **totally positive**.

Recap:  $a_0 = 1$

$$H_n(a) = \begin{bmatrix} a_0 & a_1 & \dots & a_n \\ a_1 & a_2 & \dots & a_{n+1} \\ \vdots & & & \\ a_n & a_{n+1} & \dots & a_{2n} \end{bmatrix}$$

$H_n(a)$  positive semidefinite  
then

$\Leftrightarrow a_n$  is a sequence of  
moments of a probability  
measure on the real line  
(Hamburger moment problem)

$H(a)$  totally positive

$\Leftrightarrow a_n$  is a sequence of  
moments of a probability  
measure on  $[0, \infty)$   
(Stieltjes moment problem)

$\Rightarrow a_n$  is log-convex

$T(a)$  totally positive

$\Rightarrow a_n$  is log-concave

Positivity is natural:

$$n!, \quad C_n, \quad B_n$$

$$\binom{n}{0}, \dots, \binom{n}{k}, \dots, \binom{n}{n}$$

$$\langle \binom{n}{0} \rangle, \dots, \langle \binom{n}{k} \rangle, \dots, \langle \binom{n}{n} \rangle$$

$$[\binom{n}{0}], \dots, [\binom{n}{k}], \dots, [\binom{n}{n}]$$

$$\{ \binom{n}{0} \}, \dots, \{ \binom{n}{k} \}, \dots, \{ \binom{n}{n} \}$$

⋮

But not to be expected:

- # matroids on  $[n]$ : 1, 2, 4, 8, 17, 38, 98, ... A05545
- # binary matroids on  $[n]$ : 1, 2, 4, 8, 12, 32, 68, 148, ... A076766
- # ternary matroids on  $[n]$ : 2, 4, 8, 17, 36, 85, ... A076892
- # simple matroids on  $[n]$ : 1, 2, 4, 5, 26, ... A002773
- :

And many more examples (coming soon) that  
are NOT moment sequences.

Positivity shows up in unexpected places:

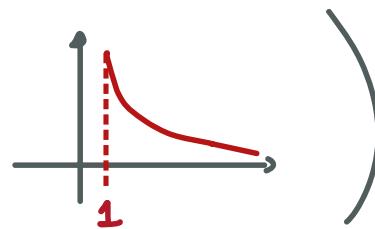
Thm

# positroids on  $[n]$

= # decorated permutations on  $[n]$  (Postnikov)

=  $n^{\text{th}}$  moment of  $1 + \text{Exp}(i)$  (Ardila, Zincoñ, Williams '16)

$$\left( = \int_1^\infty x^n e^{-(x-1)} dx \right)$$



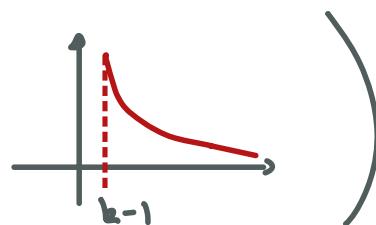
Def (B. - Steingrimsson '21) A  $k$ -arrangement on  $[n]$  is a permutation  $\sigma \in S_n$ , together with a  $k$ -coloring of its fixed points.

Remark:  $k=2 \equiv$  decorated permutations

Thm (B. - Steingrimsson '21)

#  $k$ -arrangements on  $[n]$  =  $n^{\text{th}}$  moment of  $k-1 + \text{Exp}(1)$

$$= \int_1^\infty x^n e^{-(x-k+1)} dx$$



<proof> Generating functions.

→ Various other combinatorial properties

→ Further unexpected occurrence in a different probabilistic setting } lec 4

Recall:

$[P_{d,k}(t)]_{d,k \geq 0}$  is totally positive  $\forall t > 0$

See Kaclin & McGregor '57 for consequences  
'59 for generalizations

Lec 4 will contain recent examples of a  
probabilistic "artifact" carrying additional  
probabilistic structure.

Positivity shows up in unexpected places:

### Observation / Program of work

(B. & Steingrímsson, Elvey Price & Guttman):

Hard combinatorial problems often display some form of positivity. (Focus: moment sequences)

- Deeper structural understanding
- Better asymptotics
- New tools

## Observation / Program of work

(B. & Steingrimsson, Sokal & Zeng, ... ) :

Moment sequences in combinatorics tend to be "related",  
e.g. following from some general combinatorial principle

- Unifying frameworks
- Interesting juxtapositions
- New tools / New definitions

