

Positivity everywhere

Lecture 1

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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

\rightsquigarrow

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)

$$A = [a_{ij}] = [\overline{a_{jii}}] \quad \text{s.t.} \quad \langle Av, v \rangle \geq 0 \quad \forall v$$

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

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

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

2×2 minors

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

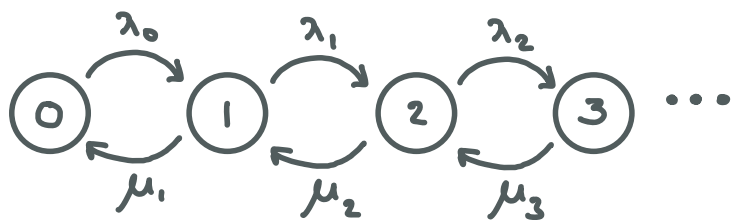
Maximal minors

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



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

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

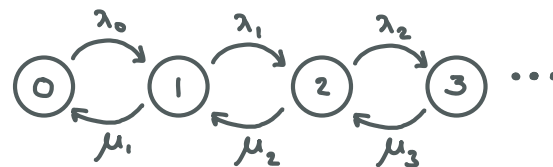
$$p_{k,k-1}(\Delta t) = \mu_k \Delta t + o(\Delta t)$$

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

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):



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

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

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

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

$$P_n(0) = \underline{I}_n$$

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

where $A_n =$

$$\begin{bmatrix} -(\lambda_0 + \mu_0) & \lambda_0 & & & 0 \\ \mu_1 & -(\lambda_1 + \mu_1) & & & \\ & \mu_2 & \dots & & \\ & & & & \lambda_{n-1} \\ 0 & & & \mu_n & -(\lambda_n + \mu_n) \end{bmatrix}$$

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

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

Some further questions you could ask:

- Which solutions P satisfying the backward equation

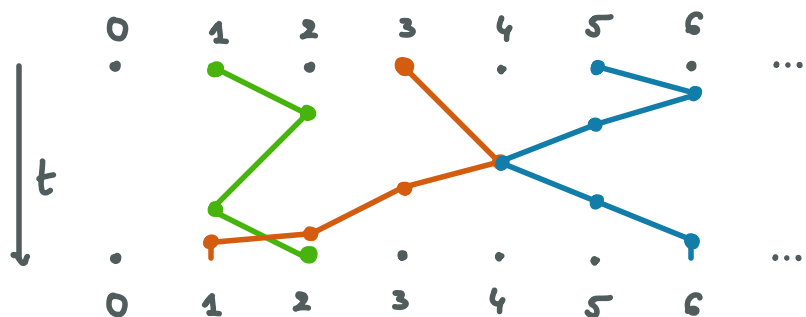
$$\frac{d}{dt} P(t) = A P(t) \text{ and the fwd eg } \frac{d}{dt} P(t) = P(t) A$$

$P(0) = I$ 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



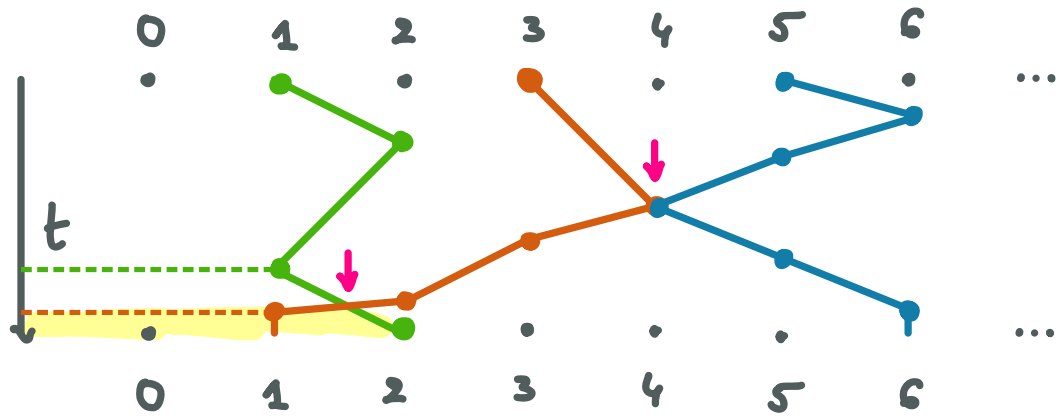
$$\det \left[p_{ij}(t) \begin{matrix} i_1 < i_2 < \dots < i_n \\ j_1 < j_2 < \dots < j_n \end{matrix} \right] = \sum_{\sigma \in S_n} (-1)^{\text{inv}(\sigma)} p_{i_1, j_{\sigma(1)}}(t) p_{i_2, j_{\sigma(2)}}(t) \dots p_{i_n, j_{\sigma(n)}}(t)$$

Claim:

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

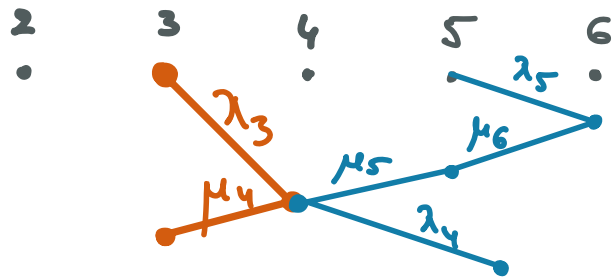
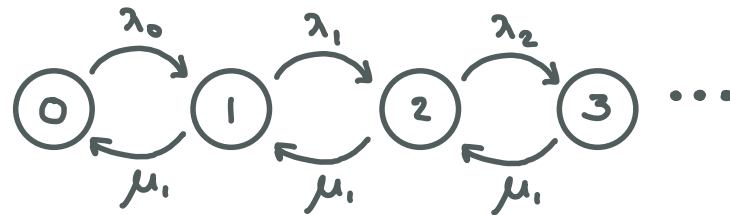
$$\det \begin{bmatrix} p_{ij}(t) \\ i_1 < i_2 < \dots < i_n \\ j_1 < j_2 < \dots < j_n \end{bmatrix} = \sum_{\sigma \in S_n} (-1)^{\text{inv}(\sigma)} p_{i_1, j_{\sigma(1)}}(t) \dots p_{i_n, j_{\sigma(n)}}(t) =$$

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

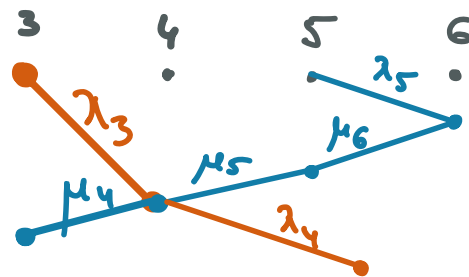


Proof by example:

$p_{ij}(t)$ 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)

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

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. $\forall \mathcal{B}_1, \mathcal{B}_2 \in \mathcal{B}$
and $b_1 \in \mathcal{B}_1 - \mathcal{B}_2$, $\exists b_2 \in \mathcal{B}_2 - \mathcal{B}_1$ w/ $(\mathcal{B}_1 - \{b_1\}) \cup \{b_2\} \in \mathcal{B}$.
("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. 1536427

\longleftrightarrow Grassmann necklaces

\longleftrightarrow J-diagrams

\longleftrightarrow equiv. class. of plabic graphs

(Postnikov '06
Ok '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 $a_k^2 \geq a_{k-1} a_{k+1} \quad \forall k$

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

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: $[n]_q := \frac{1-q^{n+1}}{1-q} = 1+q+\dots+q^n$ with $[0]_q := 0$

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

Exercise: Show that

$$\binom{n}{0}_q, \binom{n}{1}_q, \dots, \binom{n}{k}_q := \frac{[n]_q!}{[n-k]_q! [k]_q!}, \dots, \binom{n}{n}_q$$

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

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

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

The sequence μ_0, \dots, μ_{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.

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

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

Stirling II

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

Take your favorite **combinatorial** sequence (a_n) .

(From now on, $a_0 = 1$)

When does there exist a probability measure

μ on \mathbb{R} s.t.

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

Thm (Hamburger 1920-21)

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

$$\sum_{j, \ell=0}^k a_{j+\ell} z_j \bar{z}_\ell \geq 0,$$

i.e. the Hankel matrices $[a_{i+j}]_{i,j \leq n}$ are positive semidefinite $\forall n$.

<proof>

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

(\Leftarrow) Subsequent lecture

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

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

(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: $a_0 = 1$

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

$H_n(a)$ positive semidefinite
 $\forall n$

$\Leftrightarrow a_n$ 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:

$$n! , C_n , B_n$$

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

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

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

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

⋮

But not to be expected:

- # matroids on $[n]$: 1, 2, 4, 8, 17, 38, 98, ... A055545
- # binary matroids on $[n]$: 1, 2, 4, 8, 18, 32, 68, 148, ... A076766
- # ternary matroids on $[n]$: 2, 4, 8, 17, 36, 85, ... A076892
- # simple matroids on $[n]$: 1, 2, 4, 9, 26, ... A002773
-

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

Positivity shows up in unexpected places:

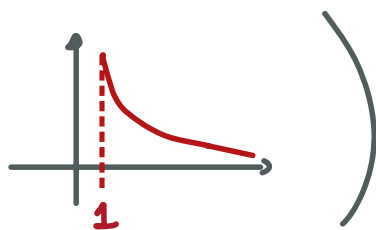
Thm

positroids on $[n]$

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

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

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

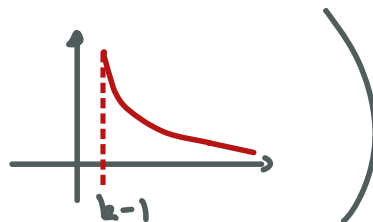


Def (B. - Steingrimssson '21) A k -arrangement on $[n]$ is a permutation $\sigma \in S_n$, together with a k -coloring of its fixed points.

Remark: $k=2 \equiv$ decorated permutations

Thm (B. - Steingrimssson '21)

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

$$\left(= \int_0^{\infty} x^n e^{-(x-k+1)} dx \right)$$


<proof> Generating functions.

→ Various other combinatorial properties

→ Further unexpected occurrence in a different probabilistic setting

} Lec
4

Recall:

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

See Karlin & 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. & Steingrímsson, 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

