

# Multivariate Krawtchouk polynomials and a spectral theorem for symmetric tensor powers

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Multivariate Krawtchouk polynomials are constructed.  
A spectral theorem for associated quantum observables is presented.

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# 1 Krawtchouk polynomials in one variable and the binomial distribution

Krawtchouk polynomials may be defined via the generating function

$$(1 + \lambda pv)^{N-j} (1 - \lambda qv)^j = \sum_{0 \leq n \leq N} v^n k_n(j, N)$$

The polynomials  $k_n(j, N)$  are orthogonal with respect to the binomial distribution with parameters  $N, p$ .

► They are part of the legacy of **Mikhail Kravchuk**

N. Virchenko, et al., eds.

[Development of the Mathematical Ideas of Mykhailo Kravchuk \(Krawtchouk\)](#),

Shevchenko Scientific Society, Kyiv-New York, 2004.

Krawtchouk polynomials appear in diverse areas of mathematics and science. Applications range from coding theory to image processing.

## Work with René Schott

- N. Botros, J. Yang, P. Feinsilver, and R. Schott, *Hardware Realization of Krawtchouk Transform using VHDL Modeling and FPGAs*, IEEE Transactions on Industrial Electronics, **49** 6 (2002)1306–1312.
- Ph. Feinsilver and R. Schott, *Finite-Dimensional Calculus*, Journal of Physics A: Math.Theor., **42**:375214, 2009.
- Ph. Feinsilver and R. Schott, *On Krawtchouk Transforms*, Intelligent Computer Mathematics, proceedings 10th Intl. Conf. AISC 2010, LNAI 6167, 64–75.
- F&S, *Algebraic Structures and Operator Calculus*, 3 vols., 1993-1995.

## 2 Symmetric tensor powers

Given a  $d \times d$  matrix  $A$ , the action on the symmetric tensor algebra of the underlying vector space defines its second quantization or “symmetric representation”.

Introduce commuting variables  $x_1, \dots, x_d$ . Map

$$y_i = \sum_j A_{ij} x_j$$

We will use multi-indices,  $m = (m_1, \dots, m_d)$ ,  $m_i \geq 0$ , similarly for  $n$ .

The induced map at level  $N$  has matrix elements  $\bar{A}_{nm}$  determined by the expansion

$$y^n = y_1^{n_1} \cdots y_d^{n_d} = \sum_m \bar{A}_{nm} x^m .$$

The matrix  $\bar{A}$  is often called the *induced matrix* at level  $N$ . The induced matrix maps monomials of homogeneous degree  $N$  to polynomials of homogeneous degree  $N$ .

## 2.1 Transpose Lemma

We introduce the special matrix  $B$  which is a diagonal matrix with multinomial coefficients as entries

$$B_{nm} = \delta_{nm} \binom{N}{n} = \frac{N!}{n_1! n_2! \cdots n_d!} .$$

$B$  is the diagonal of the induced matrix at level  $N$  of the matrix consisting of all 1's.

The level  $N$  is implicit according to context.

If  $p$  is a diagonal matrix with entries  $p_i > 0$ ,  $\sum_i p_i = 1$ , then the matrix

$$B\bar{p}$$

yields the probabilities for the corresponding **multinomial distribution**.

The map  $A \rightarrow \bar{A}$  is at each level a **multiplicative homomorphism**,

$$\overline{A_1 A_2} = \bar{A}_1 \bar{A}_2 .$$

The main lemma is the relation between the induced matrix of  $A$  with that of its transpose,  $A^\top$ .

**Transpose Lemma.**

*The induced matrices at each level satisfy*

$$\overline{A^\top} = B^{-1} \bar{A}^\top B .$$

### 3 Construction of Krawtchouk polynomial systems

Start with  $U$ , an orthogonal (unitary) matrix.

Make all entries of first column positive by taking out phases and form the probability matrix thus

$$p = \begin{pmatrix} U_{00}^2 & & \\ & \ddots & \\ & & U_{d0}^2 \end{pmatrix} = \begin{pmatrix} p_0 & & \\ & \ddots & \\ & & p_d \end{pmatrix}$$

row and column indices running from 0 to  $d$ .

Define

$$A = \frac{1}{\sqrt{p}} U \sqrt{D}$$

where  $D$  is diagonal with all positive entries on the diagonal.

The **essential property** satisfied by  $A$  is

$$A^\top p A = D .$$

### 3.1 Krawtchouk systems

In any degree  $N$ , the induced matrix  $\bar{A}$  satisfies

$$\overline{A^\top} \bar{\rho} \bar{A} = \bar{D} .$$

Using the **Transpose Lemma**

$$B \overline{A^\top} = \bar{A}^\top B$$

with  $B$  the special multinomial diagonal matrix yields

$$\Phi B \bar{\rho} \Phi^\top = B \bar{D}$$

the **Krawtchouk matrix**  $\Phi$  being thus defined as  $\bar{A}^\top$ .

The matrix elements, i. e. entries, of  $\Phi$  are the values of the **multivariate Krawtchouk polynomials** thus determined.

The matrix  $B \bar{D}$  is the diagonal matrix of squared norms according to the orthogonality of the Krawtchouk polynomial system.



## 4 Columns Theorem for symmetric powers

► **MacMahon's Master Theorem** yields the diagonal matrix elements of the symmetric tensor powers. Namely,

Let  $U = \text{diag}(u_1, \dots, u_d)$ . Then, the coefficient of  $u^m = u_1^{m_1} \cdots u_d^{m_d}$  in the expansion of  $\det(I - UA)^{-1}$  is the diagonal matrix element  $\bar{A}_{mm}$ .

► **We present** a variation that reproduces all of the matrix elements.

Given a matrix  $A$ , with each column of  $A$  form a diagonal matrix. Thus,

$$\Lambda_j = \text{diag}((A_{ij}))$$

where

$$(\Lambda_j)_{ii} = A_{ij} .$$

➔ **Columns Theorem.**

For any matrix  $A$ , let  $\Lambda_j$  be the diagonal matrix formed from column  $j$  of  $A$ . Let

$$\Lambda = \sum v_j \Lambda_j .$$

Then the coefficient of  $v^n$  in the level  $N$  induced matrix  $\bar{\Lambda}$  is a diagonal matrix with entries the  $n^{\text{th}}$  column of  $\bar{A}$ .

*Proof:* Setting  $\vec{y} = \Lambda \vec{x}$ , we have

$$y_k = \left( \sum v_j A_{kj} \right) x_k \Rightarrow y^m = \left( \sum \bar{A}_{mn} v^n \right) x^m .$$

A careful reading of the coefficients yields the result. □

We may express this in the following useful way:

the diagonal entries of  $\bar{\Lambda}$  are generating functions for the matrix elements of  $\bar{A}$ .

## 4.1 Quantum observables

Define **observables** by

$$X_j = A^{-1} \Lambda_j A .$$

Let  $X = \sum v_j X_j$ . Then

$$AX = \Lambda A$$

and the symmetric tensor powers satisfy

$$\bar{A}\bar{X} = \bar{\Lambda}\bar{A}$$

the induced spectral formula for  $\bar{X}$ .

## 4.2 Quantum random walks

- **Write**, the superscript denoting the level  $N$  symmetric tensor power,

$$(I + t \sum X_j)^{(N)} = \sum t^m \xi_m(N).$$

So  $\xi_m(N)$  is the sum of all elementary symmetric tensors of order  $N$  having exactly  $m$  factors not equal to the identity.

- **For example**, with a single  $X_j = X$ ,

$$\xi_1(3) = X \otimes I \otimes I + I \otimes X \otimes I + I \otimes I \otimes X$$

the quantum random walk after three steps.

- **Taking**  $\Lambda_0 = I$ ,  $v_0 = 1$ , and  $t$  for the remaining  $v_j$ 's in the discussion above yields the spectral representation for the quantum random walks and their extension to higher levels.

### 4.3 Spectral representation as a recurrence formula

Now take  $A$  corresponding to a **Krawtchouk system**, with  $\Phi = \bar{A}^\top$ . Then

$$\bar{X}^\top \Phi = \Phi \bar{\Lambda}$$

with  $\bar{X}^\top$  combining rows of  $\Phi$  resulting in multiplying the entries of a given row according to the spectrum.

For  $n = 1$ , this is a recurrence formula for the corresponding orthogonal polynomials. Namely, it shows the effect of multiplying  $\phi_m$ , say, by  $\phi_1$ .

The higher powers of  $v$  yield higher-level recursion formulas. They correspond to *linearization formulas* of the type

$$\phi_n \phi_m = \sum_{\ell} c_{mn}^{\ell} \phi_{\ell} .$$

## 5 Contexts

► **Gaussian quadrature** Let  $\{\phi_0, \dots, \phi_n\}$  be an orthogonal polynomial sequence with positive weight function on an interval  $I$  of the real line. For Gaussian quadrature,

$$\int_I f \approx \sum_k w_k f(x_k)$$

with  $x_k$  the zeros of  $\phi_n$  and appropriate weights  $w_k$ . Let

$$A_{ij} = \phi_{i-1}(x_j)$$

Then, with  $\Gamma$  the diagonal matrix of squared norms,

$\Gamma_{ii} = \|\phi_i\|^2$ , we have

$$AWA^\top = \Gamma$$

where  $W$  is the diagonal matrix with  $W_{kk} = w_k$ .

► **Association schemes** Given an association scheme with adjacency matrices  $A_i$ , the  $P$  and  $Q$  matrices correspond to the decomposition of the algebra generated by the  $A_i$  into an orthogonal direct sum, the entries  $P_{ij}$  being the corresponding eigenvalues. A basic result is the relation

$$P^\top D_\mu P = v D_v$$

where  $D_\mu$  is the diagonal matrix of multiplicities and  $D_v$  the diagonal matrix of valencies of the scheme.

Work of Delsarte, Bannai, . . . .



## Example

Start with the orthogonal matrix

$$U = \begin{pmatrix} 1/\sqrt{2} & 1/\sqrt{2} \\ 1/\sqrt{2} & -1/\sqrt{2} \end{pmatrix}.$$

Factoring out the squares from the first column we have

$$p = \begin{pmatrix} 1/2 & 0 \\ 0 & 1/2 \end{pmatrix} \quad \text{and} \quad A = \begin{pmatrix} 1 & 1 \\ 1 & -1 \end{pmatrix}.$$

The binomial coefficient matrix is  $B = \text{diag}(1, 4, 6, 4, 1)$ .

We have the Kravchuk matrix

$$\Phi = (A^{(4)})^\top = \begin{pmatrix} 1 & 1 & 1 & 1 & 1 \\ 4 & 2 & 0 & -2 & -4 \\ 6 & 0 & -2 & 0 & 6 \\ 4 & -2 & 0 & 2 & -4 \\ 1 & -1 & 1 & -1 & 1 \end{pmatrix}.$$

The entries of  $p$  become  $\bar{p} = \frac{1}{16} I_5$ .



Take the second column of  $A$  and form the diagonal matrix

$$\Lambda_1 = \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix}.$$

The corresponding observable is

$$X_1 = A^{-1} \Lambda_1 A = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}.$$

Let  $\Lambda = I + v \Lambda_1$  and  $X = I + v X_1$ .

Then  $\Lambda^{(4)} = \text{diag}(\$

$(1+v)^4, (1+v)^3(1-v), (1+v)^2(1-v)^2, (1+v)(1-v)^3, (1-v)^4)$

And  $X^{(4)} =$

$$\begin{pmatrix} 1 & 4v & 6v^2 & 4v^3 & v^4 \\ v & 1 + 3v^2 & 3v + 3v^3 & 3v^2 + v^4 & v^3 \\ v^2 & 2v + 2v^3 & 1 + 4v^2 + v^4 & 2v + 2v^3 & v^2 \\ v^3 & 3v^2 + v^4 & 3v + 3v^3 & 1 + 3v^2 & v \\ v^4 & 4v^3 & 6v^2 & 4v & 1 \end{pmatrix}$$

Now we have the spectrum via the coefficient of  $v$  in  $\Lambda^{(4)}$

$$\text{Spec} = \text{diag}(4, 2, 0, -2, -4)$$

and the coefficient of  $t$  in the transpose of  $X^{(4)}$  give the recurrence coefficients

$$\text{Rec} = \begin{pmatrix} 0 & 1 & 0 & 0 & 0 \\ 4 & 0 & 2 & 0 & 0 \\ 0 & 3 & 0 & 3 & 0 \\ 0 & 0 & 2 & 0 & 4 \\ 0 & 0 & 0 & 1 & 0 \end{pmatrix}$$

satisfying the relation

$$(\text{Rec}) \Phi = \Phi (\text{Spec})$$

which is essentially the recurrence relation for the corresponding Krawtchouk polynomials.

## 6 Further aspects

► **We acknowledge** the seminal paper of R. C. Griffiths  
Orthogonal polynomials and multinomial distributions,  
Australian J. Stat. **13**(1971) 27–35.

► **As Bernoulli systems** , systems of orthogonal  
polynomials related to representations of the Heisenberg  
algebra,  $\mathfrak{sl}(n)$ , etc., with probabilistic interpretations relating  
to exponential martingales of associated processes.