# A q-ANALOGUE OF DE FINETTI'S THEOREM

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ABSTRACT. A q-analogue of de Finetti's theorem is obtained in terms of a boundary problem for the q-Pascal graph. For q a power of prime this leads to a characterisation of random spaces over the Galois field  $\mathbb{F}_q$  that are invariant under the natural action of the infinite group of invertible matrices with coefficients from  $\mathbb{F}_q$ .

### 1. INTRODUCTION

The infinite symmetric group  $\mathfrak{S}_{\infty}$  consists of bijections  $\{1, 2, \ldots\} \to \{1, 2, \ldots\}$  which move only finitely many integers. The group  $\mathfrak{S}_{\infty}$  acts on the product space  $\{0, 1\}^{\infty}$  by permutations of the coordinates. A random element of this space, that is a random infinite binary sequence, is called *exchangeable* if its probability law is invariant under the action of  $\mathfrak{S}_{\infty}$ . De Finetti's theorem asserts that every exchangeable sequence can be generated in a unique way by the following two-step procedure: first choose at random the value of parameter p from some probability distribution on the unit interval [0, 1], then run an infinite Bernoulli process with probability p for 1's.

One approach to this classical result, as presented in Feller [3, Ch. VII, §4], is based on the following exciting connection with the Hausdorff moment problem. By exchangeability, the law of a random infinite binary sequence is determined by the array  $(v_{n,k})$ , where  $v_{n,k}$  equals the probability of every initial sequence of length n with k 1's. The rule of addition of probabilities yields the backward recursion

$$v_{n,k} = v_{n+1,k} + v_{n+1,k+1}, \ 0 \le k \le n, \ n = 0, 1, \dots,$$
(1)

which readily implies that the array can be derived by iterated differencing of the sequence  $(v_{n,0})_{n=0,1,\ldots}$ . Specifically, setting

$$u_l^{(k)} = v_{l+k,k}, \qquad l = 0, 1, \dots, \quad k = 0, 1, \dots,$$
 (2)

and denoting by  $\delta$  the difference operator acting on sequences  $u = (u_l)_{l=0,1,\dots}$  as

$$(\partial u)_l = u_l - u_{l+1},$$

the recursion (1) can be written as

(1)

$$u^{(k)} = \delta u^{(k-1)}, \qquad k = 1, 2, \dots$$
 (3)

G. O. was supported by a grant from the Utrecht University, by the RFBR grant 08-01-00110, and by the project SFB 701 (Bielefeld University).

Since  $v_{n,k} \ge 0$ , the sequence  $u^{(0)}$  must be completely monotone, that is, componentwise

$$\underbrace{\delta \circ \cdots \circ \delta}_{k} u^{(0)} \ge 0, \qquad k = 0, 1, \dots,$$

but then Hausdorff's theorem implies that there exists a representation

$$v_{n,k} = u_{n-k}^{(k)} = \int_{[0,1]} p^k (1-p)^{n-k} \mu(\mathrm{d}p)$$
(4)

with uniquely determined probability measure  $\mu$ . De Finetti's theorem follows since  $v_{n,k} = p^k (1-p)^{n-k}$  for the Bernoulli process with parameter p. See [1] for other proofs and extensive survey of generalisations of this result.

The present note is devoted to variations on the q-analogue of de Finetti's theorem, which was briefly outlined in Kerov [9] within the framework of the boundary problem for generalised Stirling triangles. The boundary problem for other weighted versions of the Pascal triangle was studied in [4], [6], and for more general graded graphs in [5], [9], [10].

**Definition 1.1.** Given q > 0, let us say that a random binary sequence  $\varepsilon = (\varepsilon_1, \varepsilon_2, ...) \in \{0, 1\}^{\infty}$  is *q*-exchangeable if its probability law  $\mathbb{P}$  is  $\mathfrak{S}_{\infty}$ -quasiinvariant with a specific cocycle, which is uniquely determined by the following condition: Denoting by  $\mathbb{P}(\varepsilon_1, ..., \varepsilon_n)$  the probability of an initial sequence  $(\varepsilon_1, ..., \varepsilon_n)$ , we have for any i = 1, ..., n - 1

$$\mathbb{P}(\varepsilon_1,\ldots,\varepsilon_{i-1},\varepsilon_{i+1},\varepsilon_i,\varepsilon_{i+2},\ldots,\varepsilon_n)=q^{\varepsilon_i-\varepsilon_{i+1}}\mathbb{P}(\varepsilon_1,\ldots,\varepsilon_n)$$

In words: under an elementary transposition of the form  $(\ldots, 1, 0, \ldots) \rightarrow (\ldots, 0, 1, \ldots)$ , probability is multiplied by q.

**Theorem 1.2.** Assume 0 < q < 1. There is a bijective correspondence  $\mathbb{P} \leftrightarrow \mu$  between the probability laws  $\mathbb{P}$  of infinite q-exchangeable binary sequences and the probability measures  $\mu$  on the closed countable set

$$\Delta_q := \{1, q, q^2, \ldots\} \cup \{0\} \subset [0, 1].$$

More precisely, a q-exchangeable sequence can be generated in a unique way by first choosing at random a point  $x \in \Delta_q$  distributed according to  $\mu$  and then running a certain q-analogue of the Bernoulli process indexed by x. Each law  $\mathbb{P}$  is uniquely determined by the infinite triangular array

$$v_{n,k} := \mathbb{P}(\underbrace{1,\ldots,1}_{k},\underbrace{0,\ldots,0}_{n-k}), \qquad 0 \le k \le n < \infty, \tag{5}$$

which in turn is given by a q-version of formula (4), with  $\Delta$  being replaced by  $\Delta_q$  (Theorem 3.2). A similar result with switching the roles of 0's and 1's and replacing q by  $q^{-1}$  also holds for q > 1.

The rest of the paper is organized as follows. In Section 2 we introduce the q-Pascal graph and formulate the q-exchangeability in terms of certain Markov chains on this graph. In Section 3 we find a characteristic recursion for the numbers (5), which is a q-deformation of (1), and we prove the main result, equivalent to Theorem 1.2, using the

method of [10]. In Section 4 we discuss three examples: two q-analogues of the Bernoulli process and a q-analogue of Pólya's urn process. Finally, in Section 5, for q a power of a prime number, we provide an interpretation of the theorem in terms of random subspaces in an infinite-dimensional vector space over  $\mathbb{F}_q$ .

## 2. The q-Pascal graph

For q > 0, the q-Pascal graph is a *weighted* directed graph  $\Gamma(q)$  on the infinite vertex set

$$\Gamma = \{ (l,k) : l, k = 0, 1, \ldots \}.$$

Each vertex (l, k) has two weighted outgoing edges  $(l, k) \to (l+1, k)$  and  $(l, k) \to (l, k+1)$ with weights 1 and  $q^l$ , respectively. The vertex set is divided into levels  $\Gamma_n = \{(l, k) : l+k=n\}$ , so  $\Gamma = \bigcup_{n\geq 0} \Gamma_n$  with  $\Gamma_0$  consisting of the sole root vertex (0, 0). For a path in  $\Gamma$  connecting two vertices  $(l, k) \in \Gamma_{l+k}$  and  $(\lambda, \varkappa) \in \Gamma_{\lambda+\varkappa}$  we define the weight to be the product of weights of edges along the path. For instance, the weight of  $(2, 3) \to (2, 4) \to (3, 4) \to (3, 5)$  is  $q^5 = q^2 \cdot 1 \cdot q^3$ . Clearly, such a path exists if and only if  $\lambda \geq l, \varkappa \geq k$ .

We shall consider certain transient Markov chains  $S = (S_n)$ , with state-space  $\Gamma$ , which start at the root (0,0) and move along the directed edges, so that  $S_n \in \Gamma_n$  for every  $n = 0, 1, \ldots$  Thus, a trajectory of S is an infinite directed path in  $\Gamma$  started at the root.

**Definition 2.1.** Adopting the terminology introduced by Vershik and Kerov (see [9]), we say that a Markov chain S on  $\Gamma(q)$  is *central* if the following condition is satisfied for each vertex  $(n - k, k) \in \Gamma_n$  visited by S with positive probability: given  $S_n = (n - k, k)$ , the conditional probability that S follows each particular path connecting (0, 0) and (n - k, k) is proportional to the weight of the path.

**Remark 2.2.** If we only require the centrality condition to hold for all  $(l, k) \in \Gamma_{\nu}$  for fixed  $\nu$ , then we have it satisfied also for all (l, k) with  $l + k \leq \nu$ . From this it is easy to see that the centrality condition *implies* the Markov property of S in reversed time  $n = \ldots, 1, 0$ , hence also implies the Markov property in forward time  $n = 0, 1, \ldots$ 

In the special case q = 1 Definition 2.1 means that in the Pascal graph  $\Gamma(1)$  all paths with common endpoints are equally likely.

Recall a bijection between the infinite binary sequences  $(\varepsilon_1, \varepsilon_2, ...)$  and infinite directed paths in  $\Gamma$  started at the root (0,0). Specifically, given a path, the *n*th digit  $\varepsilon_n$  is given the value 0 or 1 depending on whether *l* or *k* coordinate is increased by 1. Indentifying a path with a sequence  $(n - K_n, K_n)$  (where  $0 \le K_n \le n$ ), the correspondence can be written as

$$K_n = \sum_{j=1}^n \varepsilon_j , \ \varepsilon_n = K_n - K_{n-1}, \ n = 1, 2, \dots$$

**Proposition 2.3.** By virtue of the bijection between  $\{0,1\}^{\infty}$  and the paths in  $\Gamma$ , each q-exchangeable sequence corresponds to a central Markov chain on  $\Gamma(q)$ , and vice versa.

*Proof.* This follows readily from Remark 2.2, Definitions 1.1 and 2.1 and the structure of  $\Gamma(q)$ .

We shall use the standard notation

$$[n] := 1 + q + \ldots + q^{n-1}, \ [n]! := [1] \cdot [2] \cdots [n], \ \begin{bmatrix} n \\ k \end{bmatrix} := \frac{[n]!}{[k]![n-k]!}$$

for q-integers, q-factorials and q-binomial coefficients, respectively, with the usual convention that  $\begin{bmatrix} n \\ k \end{bmatrix} = 0$  for n < 0 or k < 0. Furthermore, we set

$$(x,q)_k := \prod_{i=0}^{k-1} (1 - xq^i), \ 1 \le k \le \infty,$$

with the infinite product  $(k = \infty)$  considered for 0 < q < 1.

The following lemma justifies the name of the graph by relating it to the q-Pascal triangle of q-binomial coefficients.

**Lemma 2.4.** The sum of weights of all directed paths from the root (0,0) to a vertex (n-k,k), denoted  $d_{n,k}$ , is given by

$$d_{n,k} = \begin{bmatrix} n\\k \end{bmatrix}. \tag{6}$$

More generally,  $d_{n,k}^{\nu,\varkappa}$ , the sum of weights of all paths connecting two vertices (n-k,k) and  $(\nu-\varkappa,\varkappa)$  in  $\Gamma$  is given by

$$d_{n,k}^{\nu,\varkappa} = q^{(\varkappa-k)(n-k)} \begin{bmatrix} \nu - n \\ \varkappa - k \end{bmatrix}.$$

*Proof.* Note that any path from (0,0) to (n-k,k) has the second component incrementing by 1 on some k edges  $(l_i, i-1) \rightarrow (l_i, i)$ , where i = 1, 2, ..., k and  $0 \le l_1 \le \cdots \le l_k \le n-k$ , thus the sum of weights is equal to

$$d_{n,k} = \sum_{0 \le l_1 \le \dots \le l_k \le n-k} q^{l_1 + \dots + l_k}.$$
(7)

This array satisfies the recursion

$$d_{n,k} = q^{n-k} d_{n-1,k-1} + d_{n-1,k}, \ 0 < k < n$$
(8)

with the boundary conditions  $d_{n,0} = d_{n,n} = 1$ . On the other hand, it is well known that the array of q-binomial coefficients also satisfies this recursion [8], hence by the uniqueness  $d_{n,k}$  is the q-binomial coefficient. In the like way the sum of weights of paths from (n-k, k)to  $(\nu - \varkappa, \varkappa)$  is

$$d_{n,k}^{\nu,\varkappa} = \sum_{n-k \le l_1 \le \dots \le l_{k'} \le \nu - \varkappa} q^{l_1 + \dots + l_{k'}}, \quad k' := \varkappa - k.$$

Comparing with (7) we see that this is equal to  $q^{(n-k)k'} \begin{bmatrix} \nu - \varkappa \\ k' \end{bmatrix}$ .

**Remark 2.5.** Changing (l, k) to (k, l) yields the *dual* q-Pascal graph  $\Gamma^*(q)$ , which has the same set of vertices and edges as  $\Gamma(q)$ , but different weights: the edge  $(l, k) \rightarrow (l, k+1)$  has now weight 1, and the edge  $(l, k) \rightarrow (l + 1, k)$  has weight  $q^k$ . The sum of weights of paths in  $\Gamma^*$  from (0,0) to (l, k) is again (6), which is related to another recursion for q-binomial coefficients,  $d_{n,k} = d_{n-1,k-1} + q^k d_{n-1,k}$ .

Consider the recursion

$$v_{n,k} = v_{n+1,k} + q^{n-k} v_{n+1,k+1}$$
, with  $v_{0,0} = 1$ , (9)

which is dual to (8), and denote by  $\mathcal{V}$  the set of nonnegative solutions to (9).

## Proposition 2.6. Formula

$$\mathbb{P}\{S_n = (n-k,k)\} = d_{n,k}v_{n,k}, \qquad (n-k,k) \in \mathbb{I}$$

establishes a bijective correspondence  $\mathbb{P} \leftrightarrow v$  between the probability laws of central Markov chains  $S = (S_n)$  on  $\Gamma(q)$  and solutions  $v \in \mathcal{V}$  to recursion (9).

*Proof.* Let S be a central Markov chain on  $\Gamma$  with probability law  $\mathbb{P}$ . Observe that the property in Definition 2.1 means precisely that the one-step *backward* transition probabilities (that is, transition probabilities in the inverse time) are of the standard form

$$\mathbb{P}\{S_{n-1} = (n-1,k) \mid S_n = (n,k)\} = \frac{d_{n-1,k}}{d_{n,k}} = \frac{\lfloor n-k \rfloor}{\lfloor n \rfloor}$$
(10)

$$\mathbb{P}\{S_{n-1} = (n-1, k-1) \mid S_n = (n, k)\} = \frac{d_{n-1, k-1}q^{n-k}}{d_{n, k}} = q^{n-k} \frac{[k]}{[n]}$$
(11)

for every such S.

Introduce the notation

$$\tilde{v}_{n,k} := \mathbb{P}\{S_n = (n-k,k)\}, \qquad (n-k,k) \in \Gamma.$$
(12)

Consistency of the distributions of  $S_n$ 's amounts to the rule of total probability

$$\tilde{v}_{n,k} = \mathbb{P}\{S_n = (n,k) \mid S_{n+1} = (n+1,k)\}\tilde{v}_{n+1,k} + \mathbb{P}\{S_n = (n,k) \mid S_{n+1} = (n+1,k+1)\}\tilde{v}_{n+1,k+1}.$$
 (13)

Rewriting (13), using (10) and (11), and setting

$$v_{n,k} = d_{n,k}^{-1} \tilde{v}_{n,k} \tag{14}$$

we get (9), which means that  $v \in \mathcal{V}$ . Thus, we have constructed the correspondence  $\mathbb{P} \mapsto v$ .

Conversely, start with a solution  $v \in \mathcal{V}$  and pass to  $\tilde{v} = (\tilde{v}_{n,k})$  according to (14). For each *n* consider the measure on  $\Gamma_n$  with weights  $\tilde{v}_{n,0}, \ldots, \tilde{v}_{n,n}$ . Since the weight of the root is 1, it follows from (9) by induction in *n* that these are probability measures. Again

 $\square$ 

by (9), the marginal measures are consistent with the backward transition probabilities, hence determine the probability law of a central Markov chain on  $\Gamma(q)$ . Thus, we get the inverse correspondence  $v \mapsto \mathbb{P}$ .

By virtue of Propositions 2.3 and 2.6, the law of q-exchangeable infinite binary sequence is determined by some  $v \in \mathcal{V}$ , with the entries  $v_{n,k}$  having the same meaning as in (5). In the sequel this law will be sometimes denoted  $\mathbb{P}_{v}$ .

### 3. The boundary problem

The set  $\mathcal{V}$  is a Choquet simplex, meaning a convex set which is compact in the product topology of the space of functions on  $\Gamma$  and has the property of uniqueness of the barycentric decomposition of each  $v \in \mathcal{V}$  over the set of extreme elements of  $\mathcal{V}$  (see, e. g., [7, Proposition 10.21]).

The boundary problem for the q-Pascal graph amounts to describing extreme nonnegative solutions to the recursion (9). Each extreme solution  $v \in \mathcal{V}$  corresponds to ergodic process  $(S_n)$  for which the tail sigma-algebra is trivial. In this context, the set of extremes is also known as the minimal boundary.

With each array  $v \in \mathcal{V}$ ,  $v = (v_{n,k})$ , it is convenient to associate another array  $\tilde{v} = (\tilde{v}_{n,k})$ related to v via (14). Clearly, the mapping  $v \leftrightarrow \tilde{v}$  is an isomorphism of two Choquet simplexes  $\mathcal{V}$  and  $\tilde{\mathcal{V}} = \{\tilde{v}\}$ . Recall that the meaning of the quantities  $\tilde{v}_{n,k}$  is explained in (12).

A common approach to the boundary problem calls for identifying a possibly larger *Martin boundary* (see [10], [6], [4] for applications of the method). To this end, we need to consider multistep backward transition probabilities, which by Lemma 6 are given by a q-analogue of the hypergeometric distribution

$$\tilde{v}_{n,k}(\nu,\varkappa) := \mathbb{P}\{S_n = (n-k,k) \mid S_\nu = (\nu-\varkappa,\varkappa)\}$$
$$= q^{(\varkappa-k)(n-k)} \begin{bmatrix} \nu - n \\ \varkappa - k \end{bmatrix} \begin{bmatrix} n \\ k \end{bmatrix} / \begin{bmatrix} \nu \\ \varkappa \end{bmatrix}, \ k = 0, \dots, n, \quad (15)$$

and to examine the limiting regimes for  $\varkappa = \varkappa(\nu)$  as  $\nu \to \infty$ , under which the probabilities (15) converge for all fixed  $(n - k, k) \in \Gamma$ . If the limits exist, the limiting array

$$\tilde{v}_{n,k} := \lim_{(\nu,\varkappa)} \tilde{v}_{n,k}(\nu,\varkappa)$$

belongs necessarily to  $\widetilde{\mathcal{V}}$ .

Suppose 0 < q < 1 and introduce polynomials

$$\Phi_{n,k}(x) := q^{-k(n-k)} x^{n-k} (x, q^{-1})_k, \qquad \widetilde{\Phi}_{n,k} = d_{n,k} \Phi_{n,k}, \qquad 0 \le k \le n.$$
(16)

Obviously, the degree of  $\Phi_{n,k}$  is n; we will consider the polynomial as a function on  $\Delta_q$ . Observe also that  $\Phi_{n,k}(x)$  vanishes at points  $x = q^{\varkappa}$  with  $\varkappa < k$ , because of vanishing of  $(x, q^{-1})_k$ . **Lemma 3.1.** Suppose 0 < q < 1, and let in (15) the indices n and k remain fixed, while  $\nu \to \infty$  and  $\varkappa = \varkappa(\nu)$  varies in some way with  $\nu$ . Then the limit of (15) is  $\tilde{\Phi}_{n,k}(q^{\varkappa})$  if  $\varkappa$  is constant for large enough  $\nu$ . If  $\varkappa \to \infty$  then the limit is  $\tilde{\Phi}_{n,k}(0) = \delta_{n,k}$ .

*Proof.* Assume first  $\varkappa \to \infty$  and show that the limit of (15) is  $\delta_{nk}$ . Since the quantities  $\tilde{v}_{n,k}(\nu,\varkappa)$ , where  $k = 0, \ldots, n$ , form a probability distribution, it suffices to check that the limit exists and is equal to 1 for k = n. In this case the right-hand side of (15) becomes

$$\prod_{i=1}^{n} \frac{[\varkappa - n + i]}{[\nu - n + i]}$$

Because  $\lim_{m\to\infty} [m] = 1/(1-q)$  for q < 1, this indeed converges to 1 provided that  $\varkappa \to \infty$ .

Now suppose  $\varkappa$  is fixed for all large enough  $\nu$ . The right-hand side of (15) is 0 for  $k > \varkappa$ . For  $k \leq \varkappa$  using  $\lim_{m\to\infty} [m-j]!/[m]! = (1-q)^j$  we obtain

$$\begin{bmatrix} \nu - n \\ \varkappa - k \end{bmatrix} / \begin{bmatrix} \nu \\ \varkappa \end{bmatrix} = \frac{[\nu - n]!}{[\nu]!} \frac{[\nu - \varkappa]!}{[\nu - \varkappa - (n - k)]!} \frac{[\varkappa]!}{[\varkappa - k]!} \rightarrow \frac{(1 - q)^k [\varkappa]!}{[\varkappa - k]!} = \widetilde{\Phi}_{n,n}(q^{\varkappa}).$$
(17)

Part (i) of the next theorem appeared in [9, Chapter 1, Section 4, Corollary 6]. Kerov pointed out that the proof could be concluded from the Kerov-Vershik 'ring theorem' (see [5, Section 8.7]), but did not give details.

For  $\mu$  a measure, we shall write  $\mu(x)$  instead of  $\mu(\{x\})$ , meaning atomic mass at x.

**Theorem 3.2.** Assume 0 < q < 1.

(i) The formulas

$$\tilde{v}_{n,k} = \sum_{x \in \Delta_q} \tilde{\Phi}_{n,k}(x)\mu(x), \qquad v_{n,k} = \sum_{x \in \Delta_q} \Phi_{n,k}(x)\mu(x)$$

establish a linear homeomorphism between the set  $\widetilde{\mathcal{V}}$  (respectively,  $\mathcal{V}$ ) and the set of all probability measures  $\mu$  on  $\Delta_q$ .

(ii) Given  $\tilde{v} \in \tilde{\mathcal{V}}$ , the corresponding measure  $\mu$  is determined by

$$\mu(q^{\varkappa}) = \lim_{\nu \to \infty} \tilde{v}_{\nu,\varkappa}, \qquad \varkappa = 0, 1, \dots; \qquad \mu(0) = 1 - \sum_{\varkappa \in \{0,1,\dots\}} \mu(q^{\varkappa}).$$

*Proof.* As in [10], the assertions (i) and (ii) are consequences of the following claims (a), (b) and (c).

(a) For each  $\nu = 0, 1, 2, ...$ , the vertex set  $\Gamma_{\nu}$  is embedded into  $\Delta_q$  via the map  $(\nu, \varkappa) \mapsto q^{\varkappa}$ . Observe that, as  $\nu \to \infty$ , the image of  $\Gamma_{\nu}$  in  $\Delta_q$  expands and in the limit exhausts the

whole set  $\Delta_q$ , except point 0, which is a limit point. In this sense,  $\Delta_q$  is approximated by the sets  $\Gamma_{\nu}$  as  $\nu \to \infty$ .

(b) The multistep backward transition probabilities (15) converge to  $\Phi_{n,k}(q^{\varkappa})$ , for  $0 \leq \varkappa \leq \infty$ , in the regimes described by Lemma 3.1.

(c) The linear span of the functions  $\widetilde{\Phi}_{n,k}(x)$ ,  $(n-k,k) \in \Gamma$ , is the space of all polynomials, so that it is dense in the Banach space  $C(\Delta_q)$ .

Note that part (ii) of the theorem can be rephrased as follows: given  $\tilde{v} \in \tilde{\mathcal{V}}$ , consider the probability distribution on  $\Gamma_n$  determined by  $\tilde{v}_{n,\bullet}$  and take its pushforward under the embedding  $\Gamma_{\nu} \hookrightarrow \Delta_q$ . The resulting probability measure on  $\Delta_q$  weakly converges to  $\mu$  as  $n \to \infty$ .

Corollary 3.3. For 0 < q < 1 we have:

(i) The extreme elements of  $\mathcal{V}$  are parameterised by the points  $x \in \Delta_q$  and have the form

$$v_{n,k} = \Phi_{n,k}(x), \qquad 0 \le k \le n. \tag{18}$$

(ii) The Martin boundary of the graph  $\Gamma(q)$  coincides with its minimal boundary and can be identified with  $\Delta_q \subset [0, 1]$  via the function  $v \mapsto v_{1,0}$ .

*Proof.* All the claims are immediate. We only comment on the fact the parameter  $x \in \Delta_q$  is recovered as the value of  $v_{1,0}$ : this holds because  $\Phi_{1,0}(x) = x$ .

Letting  $q \to 1$  we have a phase transition: the discrete boundary  $\Delta_q$  becomes more and more dense and eventually fills the whole of [0, 1] at q = 1.

As is seen from (16), the polynomial  $\Phi_{n,k}(x)$  can be viewed as a *q*-analogue of the polynomial  $x^{n-k}(1-x)^k$ , so that (18) is a *q*-analogue of (4). Keep in mind that  $x = q^{\varkappa}$  is a counterpart of 1-p, the probability of  $\varepsilon_1 = 0$ . The following *q*-analogue of the Hausdorff problem of moments emerges. Introduce a modified difference operator acting on sequences  $u = (u_l)_{l=0,1,\dots}$  as

$$(\delta_q u)_l = q^{-l}(u_l - u_{l+1}), \qquad l = 0, 1, \dots$$

**Corollary 3.4.** Assume 0 < q < 1. A real sequence  $u = (u_l)_{l=0,1,\ldots}$  with  $u_0 = 1$  is a moment sequence of a probability measure  $\mu$  supported by  $\Delta_q \subset [0,1]$  if and only if u is 'q-completely monotone' in the sense that for every  $k = 0, 1, \ldots$  we have componentwise

$$\underbrace{\delta_q \circ \cdots \circ \delta_q}_k u \ge 0, \qquad k = 0, 1, \dots$$

*Proof.* Using the notation  $v_{l+k,k} = u_l^{(k)}$  as in (2), we see that the recursion (9) is equivalent to  $u^{(k)} = \delta_q u^{(k-1)}$ , cf. (3). Then we use the fact that  $\Phi_{n,0}(x) = x^n$  and repeat in the reverse order the argument of Section 1.

The case q > 1. This case can be readily reduced to the case with parameter  $0 < \bar{q} < 1$ , where  $\bar{q} := q^{-1}$ . It is convenient to adopt a more detailed notation  $[n]_q$  for the q-integers.

**Lemma 3.5.** For every q > 0,  $\bar{q} = q^{-1}$ , the backward transition probabilities (10), (11) for the graph  $\Gamma(q)$  and the dual graph  $\Gamma^*(\bar{q})$  are the same.

*Proof.* Indeed, by virtue of (10), (11), this is reduced to the equality

$$\frac{[n-k]_q}{[n]_q} = \bar{q}^k \frac{[n-k]_{\bar{q}}}{[n]_{\bar{q}}} \,.$$

The lemma implies that the boundary problem for q > 1 can be treated by passing to  $q^{-1} < 1$  and changing (l, k) to (k, l). In terms of the binary encoding of the path, this means switching 0's with 1's.

Kerov [9, Chapter 1, Section 2.2] gives more examples of 'similar' graphs, which have different edge weights but the same backward transition probabilities.

### 4. Examples

A q-analogue of the Bernoulli process. Our first example is a description of the extreme q-exchangeable infinite binary sequences.

With each infinite binary sequence we associate some T-sequence  $(T_0, T_1, T_2, ...)$  of nonnegative integers, where  $T_j$  is the length of *j*th run of 0's. That is to say,  $T_0$  is the number of 0's before the first 1,  $T_1$  is the number of 0's between the first and second 1's,  $T_2$  is the number of 0's between the second and third 1's, and so on. Clearly, this is a bijection, i.e. a binary sequence can be recovered from its *T*-sequence as

$$(\underbrace{0,\ldots,0}_{T_0},1,\underbrace{0,\ldots,0}_{T_1},1,\underbrace{0,\ldots,0}_{T_2},1,\ldots).$$

If q = 1, then the Bernoulli process with parameter p has a simple description in terms of the associated random T-sequence: all  $T_i$  are independent and have the same geometric distribution with parameter 1 - p.

**Proposition 4.1.** Assume 0 < q < 1. For  $x \in \Delta_q$ , let  $v(x) = (v_{n,k}(x))$  be the extreme element of  $\mathcal{V}$  corresponding to x. Consider q-exchangeable infinite binary sequence  $\varepsilon = (\varepsilon_1, \varepsilon_2, \ldots)$  under the probability law  $\mathbb{P}_{v(x)}$  and let  $(T_0, T_1, \ldots)$  be the associated random T-sequence.

(i) If  $x = q^{\varkappa}$  with  $\varkappa = 1, 2, ...$  then  $T_0, ..., T_{\varkappa - 1}$  are independent,  $T_{\varkappa} \equiv \infty$ , and  $T_i$  has geometric distribution with parameter  $q^{\varkappa - i}$  for  $0 \le i \le \varkappa - 1$ .

(ii) If x = 1 then  $T_0 \equiv \infty$ , which means that with probability one  $\varepsilon$  is the sequence  $(0, 0, \ldots)$  of only 0's.

(iii) If x = 0 then  $T_0 \equiv T_1 \equiv \cdots \equiv 0$ , which means that with probability one  $\varepsilon$  is the sequence  $(1, 1, \ldots)$  of only 1's.

*Proof.* Consider the central Markov chain  $S = (S_n)$  corresponding to the extreme element  $v(q^{\varkappa})$ . Computing the forward transition probabilities, from (18) and (10), for  $0 \le k \le \varkappa$  we have

$$\mathbb{P}\{S_{n+1} = (n+1-k,k) \mid S_n = (n-k,k)\} = \frac{(q^{n+1-k}-1)}{(q^n-1)} \frac{d_{n+1,k} \Phi_{n+1,k}(q^{\varkappa})}{d_{n,k} \Phi_{n,k}(q^{\varkappa})} = q^{\varkappa-k}.$$
 (19)

This implies (i) and (ii). In the limit case x = 0 corresponding to  $\varkappa \to +\infty$ , the above probability equals 0, which entails (iii).

The analogy with the Bernoulli process is evident from the above description of the binary sequence  $\varepsilon(q^{\varkappa})$ . Moreover, the Bernoulli process appears as a limit. Indeed, fix  $p \in (0, 1)$  and suppose  $\varkappa$  varies with q, as  $q \uparrow 1$ , in such a way that

$$\varkappa \sim \frac{-\log(1-p)}{1-q}.$$

In this limiting regime,  $q^{\varkappa - k} \to 1 - p$  for every k, hence  $(T_0, T_1, ...)$  weakly converges to an infinite sequence of i.i.d. geometric variables with parameter 1-p, and the random binary sequence  $\varepsilon(q^{\varkappa})$  converges in distribution to the Bernoulli process with the frequency of 0's equal to 1-p.

Another q-analogue of Bernoulli process. Following [9], another q-analogue of Bernoulli process is suggested by the q-binomial formula (see [8])

$$(-\theta,q)_n = \sum_{k=0}^n q^{k(k-1)/2} \begin{bmatrix} n\\ k \end{bmatrix} \theta^k$$

For  $\theta \in [0,\infty]$  we define a probability law  $\mathbb{P}_{w^{\theta}}$  for  $S = (S_n)$  by setting

$$w_{n,k}^{\theta} := \frac{\theta^k q^{k(k-1)/2}}{(-\theta, q)_n}, \quad \mathbb{P}_{w^{\theta}} \{ S_n = (n-k, k) \} := d_{n,k} w_{n,k}^{\theta}, \quad (n,k) \in \Gamma.$$
(20)

Checking (9) is immediate. Computing forward transition probabilities,

$$\mathbb{P}_{w^{\theta}}\{S_{n+1} = (n+1-k,k) \mid S_n = (n-k,k)\} = 1/(1+\theta q^n),$$

shows that under  $\mathbb{P}_{w^{\theta}}$  the process  $S_n = (n - K_n, K_n)$  has independent inhomogeneous increments, with probability  $\theta q^{n-1}/(1 + \theta q^{n-1})$  for increment  $K_n - K_{n-1} = 1$ . For q = 1 we are back to the ergodic Bernoulli process, but for 0 < q < 1 the process is not extreme. To obtain the barycentric decomposition of  $w^{\theta}$  over extremes,

$$w^{\theta} = \sum_{0 \le \varkappa \le \infty} v^{\varkappa} \mu(q^{\varkappa}),$$

we can apply Theorem 3.2(ii) to compute from (20)

$$\mu(q^{\varkappa}) = \lim_{n \to \infty} \mathbb{P}_{w^{\theta}} \{ S_n = (n - \varkappa, \varkappa) \} = \frac{1}{(-\theta, q)_{\infty}} \frac{q^{\varkappa(\varkappa - 1)/2} \theta^{\varkappa}}{(1 - q)^{\varkappa} [\varkappa]!}$$

This measure  $\mu$  may be viewed as a q-analogue of the Poisson distribution.

A q-analogue of Pólya's urn process. The conventional Pólya's urn process is described in [3, Section 7.4]. Here we provide its natural deformation.

Fix a, b > 0 and 0 < q < 1. Consider the Markov chain  $(S_n)$  on  $\Gamma$  with the forward transition probabilities from (n-k,k) to (n+1-k,k) and from (n-k,k) to (n-k,k+1) given by

$$\frac{[b+n-k]}{[a+b+n]} \quad \text{and} \quad \frac{[a+k]}{[a+b+n]} q^{n-k+b},$$

respectively. Then the distribution at time n is

$$\mathbb{P}\{S_n = (n-k,k)\} = \begin{bmatrix} n \\ k \end{bmatrix} q^{bk} \\
\times \frac{[a][a+1]\cdots[a+k-1][b][b+1]\cdots[b+n-k-1]}{[a+b][a+b+1]\cdots[a+b+n-1]}.$$
(21)

Checking consistency (9) is easy. The conventional Pólya's urn process appears in the limit  $q \rightarrow 1$ . The corresponding probability measure  $\mu$  is computable from Theorem 3.2(ii) as

$$\lim_{n \to \infty} \mathbb{P}\{S_n = (n - \varkappa, \varkappa)\}\$$

For a = 1, the limit distribution of the coordinate  $\varkappa$  is geometric with parameter  $1 - q^b$ . For general a, b we obtain a measure on  $\Delta_q$ 

$$\mu(q^{\varkappa}) = \frac{(q^a, q)_{\varkappa}(q^b, q)_{\infty}}{(q, q)_{\varkappa}(q^{a+b}; q)_{\infty}} q^{\varkappa b}, \qquad q^{\varkappa} \in \Delta_q,$$

which may be viewed as a q-analogue of the beta distribution on [0, 1].

## 5. Grassmannians over a finite field

For q a power of a prime number, let  $\mathbb{F}_q$  be the Galois field with q elements. Define  $V_n$  to be the *n*-dimensional space of sequences  $(\xi_1, \xi_2, \ldots)$  with entries from  $\mathbb{F}_q$ , which satisfy  $\xi_i = 0$  for i > n. The spaces  $\{0\} = V_0 \subset V_1 \subset V_2 \subset \ldots$  comprise a complete flag, and the union  $V_{\infty} := \bigcup_{n>0} V_n$  is a countable, infinite-dimensional space over  $\mathbb{F}_q$ .

By the Grassmannian  $\operatorname{Gr}(V_{\infty})$  we mean the set of all vector subspaces  $X \subseteq V_{\infty}$ . Likewise, for  $n \geq 0$  let  $\operatorname{Gr}(V_n)$  be the set of all vector subspaces in  $V_n$ , with  $\operatorname{Gr}(V_0)$  being a singleton. Consider the projection  $\pi_{n+1,n} : \operatorname{Gr}(V_{n+1}) \to \operatorname{Gr}(V_n)$  which sends a subspace of  $V_{n+1}$  to its intersection with  $V_n$ .

**Lemma 5.1.** There is a canonical bijection  $X \leftrightarrow (X_n)$  between the Grassmannian  $\operatorname{Gr}(V_{\infty})$  and the set of sequences  $(X_n \in \operatorname{Gr}(V_n), n \ge 0)$  satisfying the consistency condition  $X_n = \pi_{n+1,n}(X_{n+1})$  for each n.

*Proof.* Indeed, the mapping  $X \mapsto (X_n)$  is given by setting  $X_n = X \cap V_n$  for each n, while the mapping  $(X_n) \mapsto X$  is defined by  $X = \bigcup X_n$ .

The lemma shows that  $\operatorname{Gr}(V_{\infty})$  can be identified with a projective limit of the finite sets  $\operatorname{Gr}(V_n)$ , the projections being the maps  $\pi_{n+1,n}$ . Using this identification we endow  $\operatorname{Gr}(V_{\infty})$  with the corresponding topology, in which  $\operatorname{Gr}(V_{\infty})$  becomes a totally disconnected compact space. For  $X \in \operatorname{Gr}(V_{\infty})$ , a fundamental system of its neighborhoods is comprised of the sets of the form  $\{X' \in \operatorname{Gr}(V_{\infty}) : X'_n = X_n\}$ , where  $n = 1, 2, \ldots$ .

Let  $\mathscr{G}_n = GL(n, \mathbb{F}_q)$  be the group of invertible linear transformations of the space  $V_n$ , realised as the group of transformations of  $V_\infty$  which may only change the first n coordinates. We have then  $\{e\} = \mathscr{G}_0 \subset \mathscr{G}_1 \subset \mathscr{G}_2 \subset \ldots$  and we define  $\mathscr{G}_\infty := \cup \mathscr{G}_n$ . The countable group  $\mathscr{G}_\infty$  consists of infinite invertible matrices  $(g_{ij})$ , such that  $g_{ij} = \delta_{ij}$  for large enough i + j. The group  $\mathscr{G}_\infty$  acts on  $V_\infty$  hence also acts on  $Gr(V_\infty)$ .

A probability distribution on  $\operatorname{Gr}(V_{\infty})$  defines a random subspace of  $V_{\infty}$ . We look at random subspaces of  $V_{\infty}$  whose distribution is invariant under the action of  $\mathscr{G}_{\infty}$ . Observe that the action of  $\mathscr{G}_n$  splits  $\operatorname{Gr}(V_n)$  into orbits

$$G(n,k) = \{ X \in \operatorname{Gr}(V_n), \dim X = k \}, \ 0 \le k \le n,$$

where  $\#G(n,k) = d_{n,k}$  is the number of k-dimensional subspaces of  $V_n$ . Therefore, a probability distribution on  $Gr(V_{\infty})$  is  $\mathscr{G}_{\infty}$ -invariant if and only if the conditional distribution on each G(n,k) is uniform.

It must be clear that this setting of 'q-exchangeability' of linear spaces is analogous to the framework of de Finetti's theorem: exchangeability of a random binary sequence means that the conditional measure is uniform on sequences of length n with k 1's. See [1], [2] for more on symmetries and sufficiency.

Lemma 5.2. Formula

$$\tilde{v}_{n,k} = P\{X \in \operatorname{Gr}(V_{\infty}) : X \cap V_n \in G(n,k)\}, \quad (n,k) \in \Gamma$$

establishes a linear homeomorphism between  $\widetilde{\mathcal{V}}$  and  $\mathscr{G}_{\infty}$ -invariant probability measures on the Grassmannian  $\operatorname{Gr}(V_{\infty})$ .

*Proof.* We first spell out more carefully the remark before the lemma. Consider projections

$$\pi_{\infty,n}$$
:  $\operatorname{Gr}(V_{\infty}) \to \operatorname{Gr}(V_n), \qquad X \mapsto X \cap V_n, \quad X \in \operatorname{Gr}(V_{\infty}), \quad n = 1, 2, \dots$ 

If P is a Borel probability measure on the space  $\operatorname{Gr}(V_{\infty})$ , then, for any n, the pushforward  $P_n := \pi_{\infty,n}(P)$  is a probability measure on  $\operatorname{Gr}(V_n)$ , and the measures  $P_n$  are consistent with respect to the projections  $\pi_{n+1,n}$ , that is,

$$P_n = \pi_{n+1,n}(P_{n+1}), \qquad n = 0, 1, 2, \dots$$

Conversely, if a sequence  $(P_n)$  of probability measures is consistent, then it determines a probability measure P on  $\operatorname{Gr}(V_{\infty})$ . Moreover, P is  $\mathscr{G}_{\infty}$ -invariant if and only if each  $P_n$ is  $\mathscr{G}_n$ -invariant. Next, observe that if  $P_n$  is a  $\mathscr{G}_n$ -invariant probability measure, then it assigns the same weight to each k-dimensional space  $X_n \in G(n,k)$ ; let us denote this weight by  $v_{n,k}$ .

Fix  $X_n \in G(n, k)$ . We claim that there are precisely  $q^{n-k}+1$  subspaces  $X_{n+1} \in Gr(V_{n+1})$ such that  $X_{n+1} \cap V_n = X_n$ : one subspace from G(n+1, k) and  $q^{n-k}$  subspaces from G(n+1, k+1). Indeed, dim  $X_{n+1}$  equals either k or k+1. In the former case  $X_{n+1} = X_n$ , while in the latter case  $X_{n+1}$  is spanned by  $X_n$  and a nonzero vector from  $V_{n+1} \setminus V_n$ . Such a vector is defined uniquely up to a scalar multiple and addition of an arbitrary vector from  $X_n$ . Therefore, the number of options is equal to the number of lines in  $V_{n+1}/X_n$ not contained in  $V_n/X_n$ , which equals

$$\frac{q^{n+1-k}-1}{q-1} - \frac{q^{n-k}-1}{q-1} = q^{n-k}.$$

Now, let P be a  $\mathscr{G}_{\infty}$ -invariant probability measure on  $\operatorname{Gr}(V_{\infty})$ , with projections  $(P_n)$  specified by the corresponding array of weights  $v = (v_{n,k})$ . Then the relations  $P_n = \pi_{n+1,n}(P_{n+1})$  together with the dimension computation imply that v satisfies (9).

Conversely, given  $v \in \mathcal{V}$ , we can construct a sequence  $(P_n)$  of measures such that  $P_n$ lives on  $\operatorname{Gr}(V_n)$ , is invariant under  $\mathscr{G}_n$  and agrees with  $P_{n+1}$  under  $\pi_{n+1,n}$ . Since  $P_0$ , which lives on a singleton, is obviously a probability measure, we obtain by induction that all  $P_n$ are probability measures. Taking their projective limit we get a  $\mathscr{G}_{\infty}$ -invariant probability measure P on  $\operatorname{Gr}(V_{\infty})$ .

Rephrasing Theorem 3.2 we have from the lemma

**Corollary 5.3.** The ergodic  $\mathscr{G}_{\infty}$ -invariant probability measures on  $\operatorname{Gr}(V_{\infty})$  are parameterised by  $\varkappa \in \{0, 1, \ldots, \infty\}$ . For  $\varkappa = 0$  the measure is the Dirac mass at  $V_{\infty}$ , for  $\varkappa = \infty$  it is the Dirac mass at  $V_0$ , and for  $0 < \varkappa < \infty$  the measure is supported by the set of subspaces of  $V_{\infty}$  of codimension  $\varkappa$ .

The following random algorithm describes explicitly the dynamics of the growing space  $X_n \in \operatorname{Gr}(V_n)$  as n varies, under the ergodic measure with parameter  $\varkappa$ . Recall the notation  $\bar{q} = q^{-1}$ . Start with  $X_0 = V_0$ . With probability  $\bar{q}^{\varkappa}$  choose  $X_1 = V_1$ , and with probability  $1 - \bar{q}^{\varkappa}$  choose  $X_1 = X_0$ . Suppose  $X_n \subseteq V_n$  has been constructed and has dimension n - k with  $k \leq \varkappa$ . Then let  $X_{n+1} = X_n$  with probability  $1 - \bar{q}^{\varkappa - k}$ , and with probability  $\bar{q}^{\varkappa - k}$  choose uniformly at random a nonzero vector  $\xi \in V_{n+1} \setminus V_n$  and let  $X_{n+1}$  be the linear span of  $X_n$  and  $\xi$ .

Duality. We finish with a dual version of our construction. Let  $V^{\infty}$  denote the set of all sequences  $\eta = (\eta_1, \eta_2, ...)$  with entries from  $\mathbb{F}_q$ . This is again a vector space over  $\mathbb{F}_q$ , strictly larger than  $V_{\infty}$  since we do not require  $\eta$  to have finitely many nonzero entries. That is to say,  $V^{\infty}$  is just the infinite product space  $(\mathbb{F}_q)^{\infty}$ , which we endow with the product topology. Let  $\operatorname{Gr}(V^{\infty})$  denote the set of all *closed* subspaces  $Y \subseteq V^{\infty}$ . A dual version of Lemma 5.1 says that such subspaces Y are in a bijective correspondence with the sequences  $(Y_n \in \operatorname{Gr}(V_n), n \ge 0)$  such that  $Y_n = \pi'_{n+1,n}(Y_{n+1})$ , where  $\pi'_{n+1,n}$  is induced by the projection map  $V_{n+1} \to V_n$  which sets the (n+1)th coordinate of a vector  $\xi \in V_{n+1}$ equal to 0. The branching of G(n, k)'s under these projections corresponds to the dual q-Pascal graph.

**Lemma 5.4.** The operation of passing to the orthogonal complement with respect to the bilinear form

$$\langle \xi, \eta \rangle := \sum_{i=1}^{\infty} \xi_i \eta_i, \qquad \xi \in V_{\infty}, \quad \eta \in V^{\infty},$$

is a bijection  $\operatorname{Gr}(V_{\infty}) \leftrightarrow \operatorname{Gr}(V^{\infty})$ .

*Proof.* First of all, note that the bilinear form is well defined, because the coordinates  $\xi_i$  of  $\xi \in V_{\infty}$  vanish for *i* large enough. This form determines a bilinear pairing  $V_{\infty} \times V^{\infty} \to \mathbb{F}_q$ . We claim that it brings the spaces  $V_{\infty}$  and  $V^{\infty}$  into duality, where  $V^{\infty}$  is viewed as a vector space with nontrivial topology, and the topology on  $V_{\infty}$  is discrete.

Indeed, it is evident that the pairing is nondegenerate and that any linear functional on  $V_{\infty}$  is given by a vector of  $V^{\infty}$ . A minor reflection also shows that, conversely, any *continuous* linear functional on  $V^{\infty}$  is given by a vector from  $V_{\infty}$ . Thus, the spaces  $V_{\infty}$  and  $V^{\infty}$  are indeed dual to one another. They are also dual as commutative locally compact topological groups: one is discrete and the other is compact.

Using the duality, it is readily checked that if X is an arbitrary subspace in  $V_{\infty}$ , then its orthogonal complement  $X^{\perp}$  is a closed subspace in  $V^{\infty}$ , whose orthogonal complement  $(X^{\perp})^{\perp}$  coincides with X. Likewise, starting with a closed subspace  $Y \subseteq V^{\infty}$ , we have  $Y^{\perp} \subseteq V_{\infty}$  and  $(Y^{\perp})^{\perp} = Y$ . Thus, the operation of taking the orthogonal complement is a bijection.

The group  $\mathscr{G}_{\infty}$  acts on both  $V_{\infty}$  and  $V^{\infty}$  and preserves the pairing between these vector spaces. Under the identification  $\operatorname{Gr}(V^{\infty}) = \operatorname{Gr}(V_{\infty})$ , the group  $\mathscr{G}_{\infty}$  acts by homeomorphisms on this compact space. In the dual picture, the ergodic measures with  $\varkappa < \infty$  live on the set of  $\varkappa$ -dimensional subspaces of  $V^{\infty}$ . The case  $\varkappa = \infty$  corresponds then to the zero subspace in  $V_{\infty}$  (or the full space  $V^{\infty}$ ). There is a simple explanation why we have to fix codimension in the  $V_{\infty}$ -picture and dimension in the  $V^{\infty}$ -picture, and not vice versa. Namely, the subspaces in  $V_{\infty}$  of fixed nonzero finite dimension form a countable set, which is a single  $\mathscr{G}_{\infty}$ -orbit, and such a  $\mathscr{G}_{\infty}$ -space cannot carry a finite invariant measure.

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