

# Bernoulli Convolutions and Measures on the Spectra of Algebraic Integers

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# The Spectrum of $\beta$

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- ▶ **Today's Questions:** What can we say about the measures

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where each  $a_i, b_i$  picked independently from  $\{0, 1\}$  with probability  $(\frac{1}{2}, \frac{1}{2})$ .

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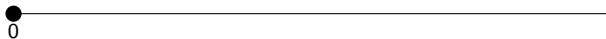
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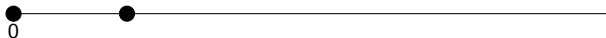
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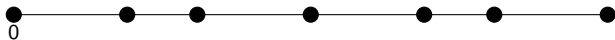
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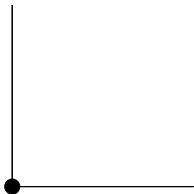
- ▶ Gaps of size  $1$  and  $\phi - 1$ .
- ▶ Gap sequence  $ABAABAABA \dots$  generated by Fibonacci substitution

$$A \rightarrow AB, B \rightarrow A$$



## Where does the structure come from?

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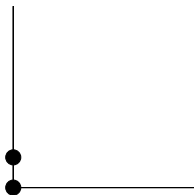


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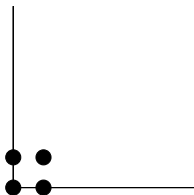
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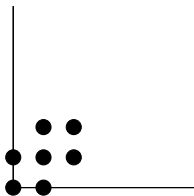
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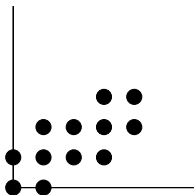
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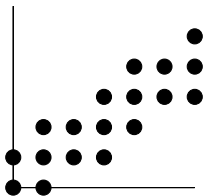
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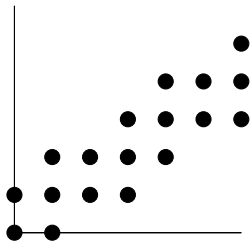
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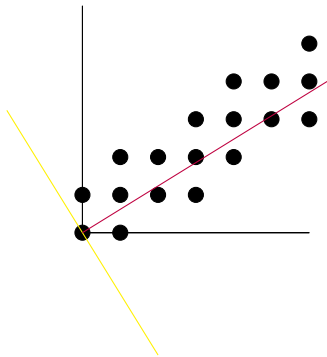
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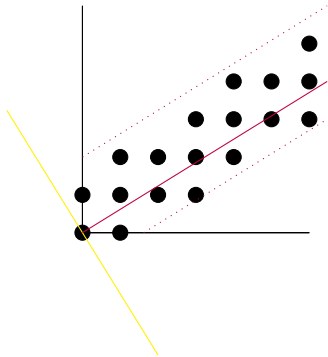
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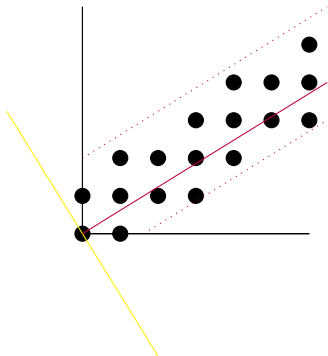




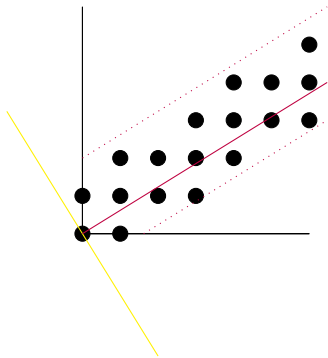








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- ▶ Not so hard to see that we can never escape the strip between the two dotted purple lines. Slightly harder - every lattice point  $(z_1, z_0)$  between the two dotted lines can be reached:

$$\begin{pmatrix} z_1 \\ z_0 \end{pmatrix} = \tilde{T}_{a_1} \circ \dots \circ \tilde{T}_{a_n} \begin{pmatrix} 0 \\ 0 \end{pmatrix}.$$

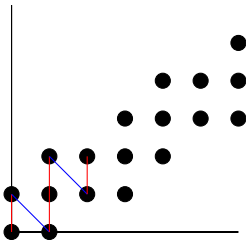




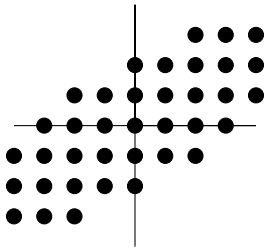
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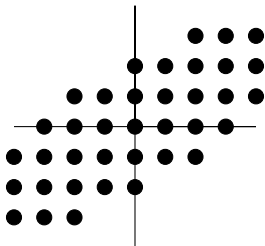
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We can move through the spectrum

### Lemma

Let the map  $G$  on  $X_{\{-1,0,1\}}(\phi) \times (-\phi^2, \phi^2)$  be given by

$$G(x, y) = \begin{cases} (x + 2\phi - 3, y - \frac{2}{\phi} - 3) & y \in [\phi, \phi^2) \\ (x + \phi - 1, y - \frac{1}{\phi} - 1) & y \in (0, \phi) \\ (x + 2 - \phi, y + 2 + \frac{1}{\phi}) & y \in (-\phi^2, 0] \end{cases}$$

Then if  $x$  is the  $n$ th element to the right of 0 in  $X_{\{-1,0,1\}}(\phi)$  and  $x_c$  is the corresponding point in the contracting direction we have that



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- ▶  $\mu_n(x)$  supported on  $X_{\{-1,0,1\}}(\beta)$ . Maps  $T_i(x) = \beta x + i$  expand, so we have a finite recurrent bit near 0 and a dissipative bit.

▶  $\phi = \frac{1+\sqrt{5}}{2}$ ,  $a_1, \dots, a_n, b_1, \dots, b_n$  i.i.d. with probability  $(\frac{1}{2}, \frac{1}{2})$ .

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### Theorem (Batsis, K.)

There exist  $\theta > 1$  and a function  $f : X_{\{-1,0,1\}}(\phi) \rightarrow \mathbb{R}^+$  such that

$$\lim_{n \rightarrow \infty} \theta^n \mathbb{P} \left\{ \sum_{i=0}^n a_i \phi^i - \sum_{i=0}^n b_i \phi^i = x \right\} = f(x).$$

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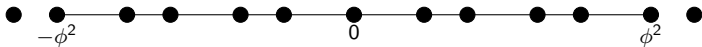
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- ▶  $\phi$  can be replaced by any algebraic integer  $\beta \in (1, 2)$  for which all of the other Galois conjugates  $\beta_i$  have  $|\beta_i| \neq 1$ .



Let  $T_i(x) = \phi x + i$ .

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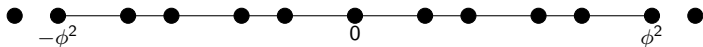
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spends a long time in the recurrent piece before spending a bounded (indep of  $n$ ) amount of time reaching  $x$ .

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decays like  $\theta^n$  where  $\theta$  is the max eigenvalue of the matrix encoding the dynamics on the recurrent piece.

Let  $\mu$  be the infinite, locally finite measure on  $X_{\{-1,0,1\}}(\beta)$  given by

$$\mu = \sum_{x \in X_{\{-1,0,1\}}(\beta)} f(x) \delta_x.$$

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**Goal for next section:** We can move through  $X_{\{-1,0,1\}}(\beta)$  using an odometer map. Can we do something similar for the measure?

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Strictly positive matrices contract projective space. Our matrices aren't strictly positive, but we mess around a bit. For  $x \in X_{\{-1,0,1\}}(\phi)$  let  $x_c \in [0, 1]$  be the corresponding point in the contracting window and  $x'$  be the nearest neighbour of  $x$ .

**Theorem [Batsis, K.]** There exists a function  $\psi : [0, 1] \rightarrow \mathbb{R}$ , continuous except on a set of Hausdorff dimension  $< 1$ , such that  $\frac{\mu(x')}{\mu(x)} = \psi(x_c)$ .

## A Concrete Example

Theorem (Batsis, K.)

Let the map  $F$  on  $X_{\{-1,0,1\}}(\phi) \times (-\phi^2, \phi^2) \times \mathbb{R}$  be given by

$$F(x, y, z) = \begin{cases} (x + 2\phi - 3, y - \frac{2}{\phi} - 3, z + \psi(y)) & y \in [\phi, \phi^2) \\ (x + \phi - 1, y - \frac{1}{\phi} - 1, z + \psi(y)) & y \in (0, \phi) \\ (x + 2 - \phi, y + 2 + \frac{1}{\phi}, z + \psi(y)) & y \in (-\phi^2, 0] \end{cases}$$

Then if  $x$  is the  $n$ th element to the right of 0 in  $X_{\{-1,0,1\}}(\phi)$  we have that

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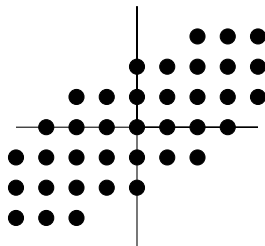
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## Bernoulli Convolutions:

- ▶ Given  $\beta \in (1, 2)$  the Bernoulli convolution  $\nu_\beta$  is the weak\* limit of the measures  $\nu_{\beta, n}$  given by

$$\nu_{\beta, n} = \frac{1}{2^n} \sum_{a_1 \cdots a_n \in \{0, 1\}^n} \delta_{\sum_{i=1}^n a_i \beta^{-i}}$$

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- ▶ Very few specific examples of absolutely continuous Bernoulli convolutions, due to Garsia (1950s) and Varju (2020).

If  $\nu_\beta$  is absolutely continuous then the density  $h_\beta$  also satisfies a self-similarity relation.

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### Theorem (Batsis, K.)

Suppose there exists a constant  $C$  that the total number  $\mathcal{N}_n$  of words  $a_1 \cdots a_n, b_1 \cdots b_n \in \{0, 1\}^n$  satisfying

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**Goal:** Let  $\beta \in (1, 2)$  be an algebraic unit, non-Pisot, no Galois conjugates of absolute value one. Use the cut and project structure of the sets  $X_A(\beta)$  to study pairs of words  $a_1, \cdots, a_n, b_1, \cdots, b_n$  as above. Use the measures  $\mu_n$  to count them.