

Pattern Matching on Correlated Sources

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Random Sources (of texts)

A random source is a process that construct random words. Each words w is produced with probability p_w such that $\sum_{|w|=k} p_w = 1$

- Uniform sources: each symbol is produced independently with a uniformly probability.
- Memoryless sources: each symbol is produced independently with a fixed probability.
- (Hidden) Markov chains: each symbol is produced with a bounded memory on the past.
- Dynamical Sources, mixing sources, . . . : the memory is not bounded.

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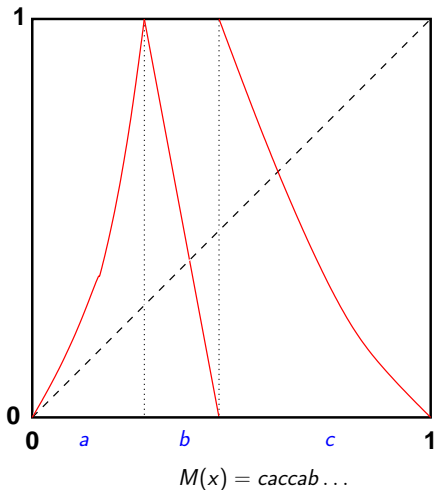
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Deterministic mechanism:

- 1) an alphabet Σ
- 2) an encoding function σ
- 3) A shift function T

Random choice:

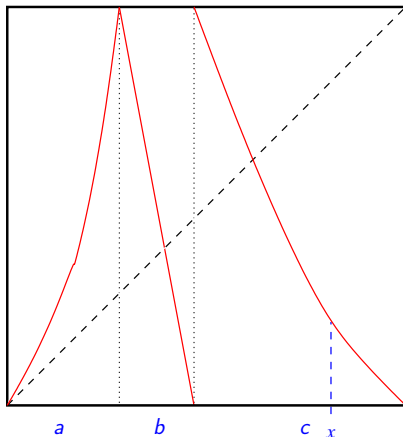
- 4) An initial density f

Word produced:

$$M(x) := (\sigma(x), \sigma(Tx), \sigma(T^2x), \dots)$$

Fundamental intervals:

$$I_w = \{x \mid M(x) \text{ begins with } w\}.$$



$M(x) = caccab\dots$

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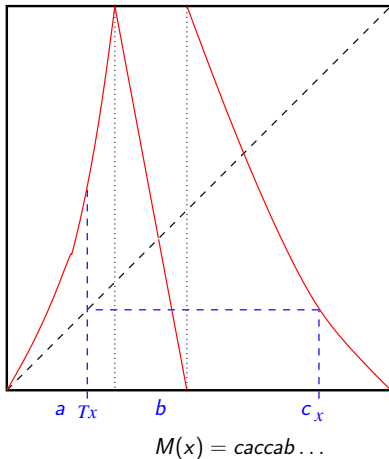
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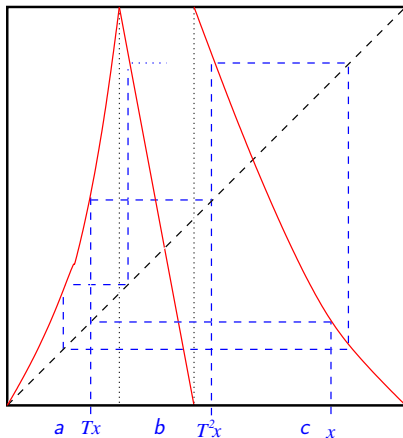
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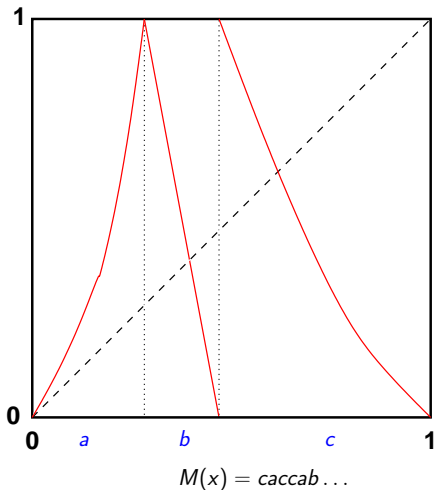
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Density transformer operator

f_0 is the initial density on $[0, 1]$

X R.V. of density f_0

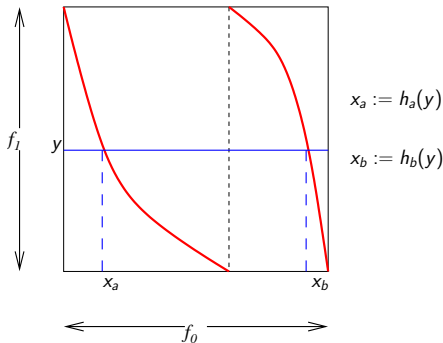
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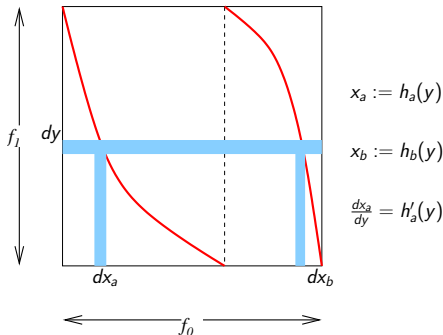
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$$f_1(y) = \sum_{m \in \Sigma} |h'_m(y)| f_0 \circ h_m(y) =: \mathbf{G}[f_0](y)$$



$$p_w := \int_{\mathcal{I}_w} f(t) dt =$$

$$M(z, u) = \sum p_w u^{C(w)} z^{|w|} \quad \leftrightarrow \quad \mathbf{M}(z, u) = \sum \mathbf{G}_w u^{C(w)} z^{|w|}$$

$$p_{w \cdot w'} = p_w p_{w'} \quad \leftrightarrow \quad \mathbf{G}_{w \cdot w'} = \mathbf{G}_{w'} \circ \mathbf{G}_w$$

$$p_w := \int_{\mathcal{I}_w} f(t) dt = \int_0^1 |h'_w| f \circ h_w(t) dt, \quad h_w = (T^{|w|})|_{\mathcal{I}_w}^{-1}$$

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Nice “decomposable” Sources

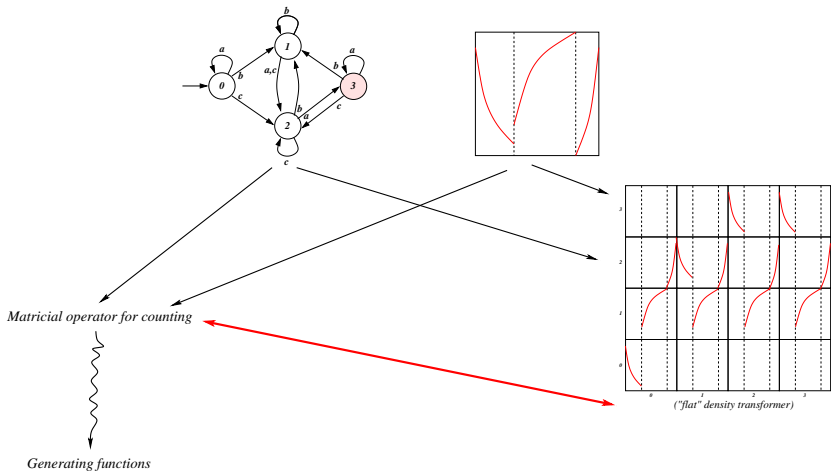
A Dynamical Source is called “decomposable” if its density transformer, when acting on an adapted Banach space, possesses a unique dominant eigenvalue separated from the remainder of the spectrum by a “spectral gap”.

When the alphabet is **finite**, this property is satisfied when branches are **expansives** and when the source is **topologically mixing**.

$$\mathbf{G}^n = \lambda^n \mathbf{P} + \mathbf{N}^n, \quad (\lambda = 1).$$

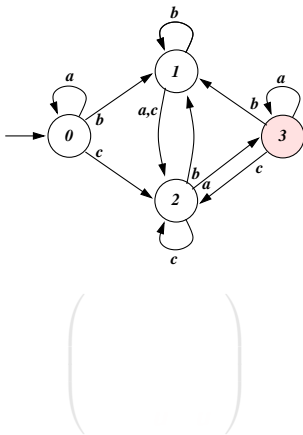
(on the function space $BV(\mathcal{I})$ endowed with the norm $\|f\| = \sup |f| + V(f)$).

Core of the method



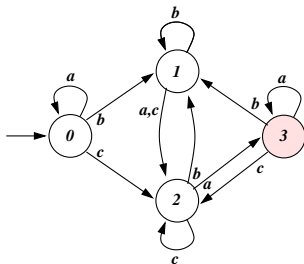
1 - A matrixial operator for counting

We first construct the DFA that recognize $\Sigma^* \mathcal{R}$



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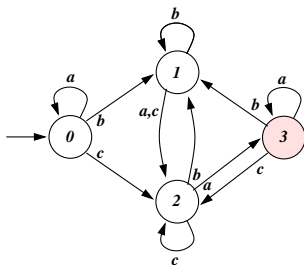
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$$\mathcal{T} = \begin{pmatrix} \{a\} & \emptyset & \emptyset & \emptyset \\ \{b\} & \{b\} & \{b\} & \{b\} \\ \{c\} & \{a, c\} & \{c\} & \{c\} \\ \emptyset & \emptyset & \color{red}u\{a\} & \color{red}u\{a\} \end{pmatrix}$$

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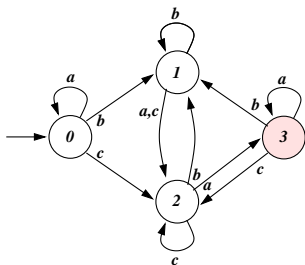
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$$T = \begin{pmatrix} p_a & 0 & 0 & 0 \\ p_b & p_b & p_b & p_b \\ p_c & p_a + p_c & p_c & p_c \\ 0 & 0 & \color{red}p_a & \color{red}p_a \end{pmatrix}$$

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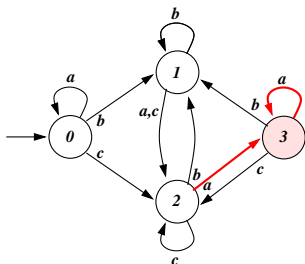
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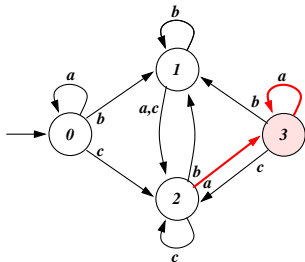
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$$\mathbb{T}(u) = \begin{pmatrix} \mathbf{G}_a & 0 & 0 & 0 \\ \mathbf{G}_b & \mathbf{G}_b & \mathbf{G}_b & \mathbf{G}_b \\ \mathbf{G}_c & \mathbf{G}_a + \mathbf{G}_c & \mathbf{G}_c & \mathbf{G}_c \\ 0 & 0 & u\mathbf{G}_a & u\mathbf{G}_a \end{pmatrix}$$

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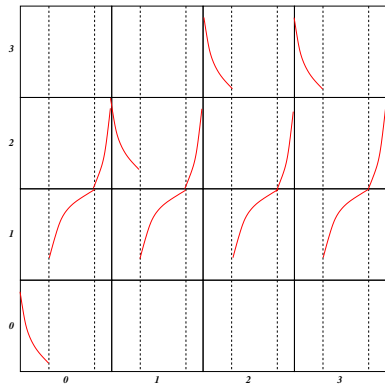


$$\mathbf{M}(z, u) = \sum_n (1, \dots, 1) \begin{pmatrix} \mathbf{G}_a & 0 & 0 & 0 \\ \mathbf{G}_b & \mathbf{G}_b & \mathbf{G}_b & \mathbf{G}_b \\ \mathbf{G}_c & \mathbf{G}_a + \mathbf{G}_c & \mathbf{G}_c & \mathbf{G}_c \\ 0 & 0 & u\mathbf{G}_a & u\mathbf{G}_a \end{pmatrix}^n \begin{pmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{pmatrix} z^n$$

2 - A mixed source

\mathcal{A} automaton with r states, $\delta(m, i)$ its transition function.
The mixed source is defined on interval $]0, r[$, for alphabet $\Sigma \times \{1, \dots, r\}$, by

$$\mathcal{I}_{m,i} = \mathcal{I}_m + i, \quad \mathcal{J}_{m,i} = \mathcal{J}_m + \delta(m, i), \quad h_{m,i}(t) = h_m(t - \delta(m, i)) + i$$



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- The mixed alphabet is finite (of size $|\Sigma| \times r$)
- The mixed branches are expansives
- The mixed source is topologically mixing (if any state of the automaton can be reached for any other state)
- \mathcal{G} acts on $BV(]0, r[)$ and decomposes as

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3 - Matricial operator vs “flat” operator

\mathfrak{G} and \mathbb{T} are conjugated by Ψ ,

$$\mathfrak{G} = \Psi \circ \mathbb{T} \circ \Psi^{-1},$$

où $\Psi : (BV(\mathcal{I}))^r \rightarrow BV(]0, r[)$,

$$\Psi({}^t(g_1, \dots, g_r))(x) = \sum_{i=1}^r \mathbb{1}_{[i-1, i]}(x) \cdot g_i(x - i + 1)$$

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Analytical perturbation

$$\mathbb{E}[C_n] = \lambda'(1)c_1 n + o(n),$$

$$\mathbb{V}[C_n] = (\lambda''(1) + \lambda'(1) - (\lambda'(1))^2)c_2 n + o(n)$$

C_n follow asymptotically a gaussian law.