

Lesson 3

Dynamical Systems

3-6: The Ergodic Theorem

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From Recurrence to Averages

(X, \mathcal{B}, μ) probability space, $T : X \rightarrow X$ μ -preserving.

Poincaré recurrence: For every measurable $E \subseteq X$ with $\mu(E) > 0$, almost every point in E returns to E infinitely often.

Question: Can we say something about how often a point returns on average?

- ▶ Assume $X = A^{\mathbb{N}}$, A finite, T shift mapping, $E = [\sigma]$.
- ▶ For a sequence x to return to E then means that σ occurs as a substring in x .
- ▶ The average number of returns to $[\sigma]$ by time n is then given as

$$\frac{\#\{i < n : x_i \dots x_{i+|\sigma|-1} = \sigma\}}{n}$$

$[\sigma]$ $\mu[\sigma]$

- ▶ **Q:** Does this average converge? To $\mu[\sigma]$?

→ **Ergodic Theorem**

The Ergodic Theorem

T measure-preserving transformation on probability space (X, \mathcal{B}, μ) .

- If f is a μ -integrable function on X , then the average

$$\frac{1}{n} \sum_{i=0}^{n-1} f(T^i(x))$$

converges for μ -almost every $x \in X$.

- If we denote the limit by $f^*(x)$, then f^* is integrable ($L^1(\mu)$) and $\int f d\mu = \int f^* d\mu$. Furthermore, f^* is T -invariant, i.e. $f^* \circ T = f^*$.
- Finally, if T is ergodic, then f^* is constant μ -a.e. and hence $f^*(x) = \int f d\mu$, which means

$$\frac{1}{n} \sum_{i=0}^{n-1} f(\underbrace{T^i(x)}_{\text{ergodic}}) \xrightarrow{n \rightarrow \infty} \int \underline{f d\mu}.$$

Example - Law of Large Numbers

- ▶ Let (X_n) be a binary Bernoulli process with $P(1) = p$, $P(0) = 1 - p$.
- ▶ μ_p Kolmogorov measure on $2^{\mathbb{N}}$, i.e. $\mu_p[1] = p$, $\mu_p[0] = 1 - p$. μ_p is invariant under shift map T .

- ▶ Let $f = \chi_{[1]}$. We have $f(T^i x) = 1$ iff $T^i x \in [1]$ iff $x_i = 1$.

- ▶ By the ergodic theorem, for μ_p -almost every $x \in 2^{\mathbb{N}}$,

$$\begin{aligned} \lim_n \frac{1}{n} \sum_{i=0}^{n-1} \chi_{[1]}(T^i x) &= \lim_n \frac{|\{i < n : x_i = 1\}|}{n} \\ &= \lim_n \frac{x_0 + x_1 + \cdots + x_{n-1}}{n} = \mu_p[1] = p. \end{aligned}$$

$\mu_p[\cdot]$
 $\rightarrow \int \chi_{[1]} d\mu_p$

Therefore, the ergodic theorem can be seen as an **extension of the strong law of large numbers** to arbitrary stationary processes.

Proving the Ergodic Theorem (I)

It suffices to consider real-valued $f \in L^1(\mu)$. Put

$$f^*(x) = \limsup_n \frac{1}{n} \sum_{i=0}^{n-1} f(T^i x) \quad \text{and} \quad f_*(x) = \liminf_n \frac{1}{n} \sum_{i=0}^{n-1} f(T^i x).$$

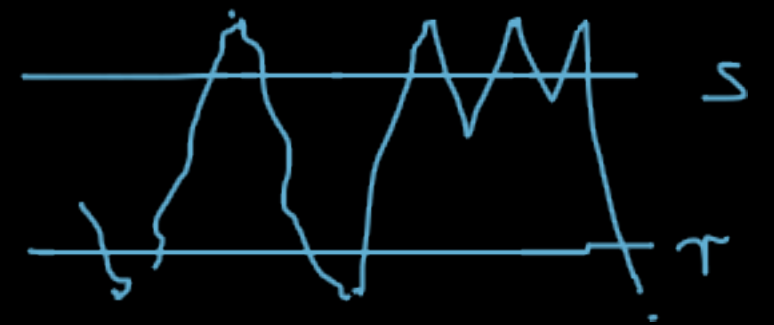
We want to show: $f^*(x) = f_*(x)$ μ -almost everywhere.

For real $r < s$, let

$$E_{r,s} = \{ \underline{x} \in X : f_*(x) < r \text{ and } f^*(x) > s \}.$$

We have

$$\{x : f_*(x) < f^*(x)\} = \bigcup_{r < s \in \mathbb{Q}} E_{r,s}.$$



Hence it suffices to show $\mu(E_{r,s}) = 0$.

Proving the Ergodic Theorem (II)

Idea: Show that

$$\int_{\underline{E_{r,s}}} f d\mu \geq s \mu(E_{r,s}),$$

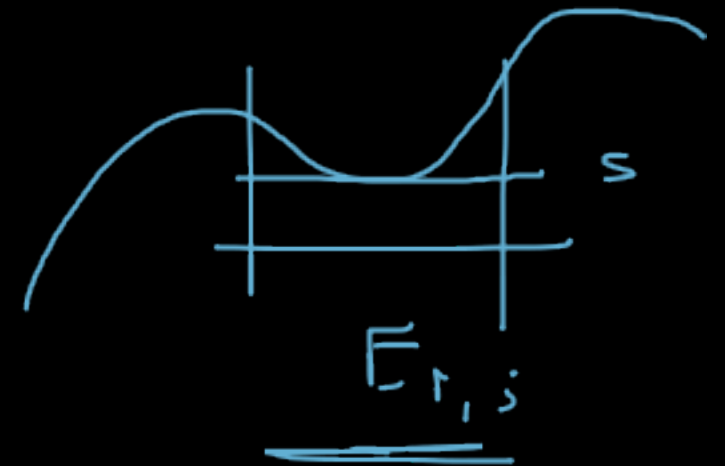
and at the same time

$$\int_{E_{r,s}} f d\mu \leq r \mu(E_{r,s}).$$

If $r < s$, this forces $\mu(E_{r,s}) = 0$.

This follows from the maximal ergodic theorem.

pointwise



Maximal Ergodic Theorem

THM: Let T be a measure-preserving transformation on a probability space (X, \mathcal{B}, μ) , and $f \in L^1(\mu)$. Define

$$f^*(x) = \limsup_n \frac{1}{n} \sum_{i=0}^{n-1} f(T^i x).$$

Then, for any $\lambda \in \mathbb{R}$

$$\int_{\{f^* > \lambda\}} f d\mu \geq \lambda \mu\{f^* > \lambda\}.$$

A Combinatorial Approach

Suppose T is the shift on $2^{\mathbb{N}}$ and μ is shift-invariant and ergodic.

Let's try to give a more "direct" proof that almost surely,

$$\lim_n \frac{1}{n} \sum_{i=0}^{n-1} x_i = \mu[1].$$

Suppose the statement is false, then wlog for some $\varepsilon > 0$, the set

$$E = \{x: \limsup_n \frac{1}{n} \sum_{i=0}^{n-1} x_i \geq \mu[1] + \varepsilon\}$$

has positive measure $\mu(E) > 0$.

E is T -invariant ($TE \subseteq E$), so by ergodicity $\mu(E) = 1$.

So almost every $x \in 2^{\mathbb{N}}$ has a long stretch of "too many" 1's infinitely often.

A Combinatorial Approach

On the other hand, at any finite point in time, the expected value of $\sum_{i < n} x_i / n$ must be $\mu[1]$, since

$$\begin{aligned} \mu[1] &= \int \chi_{[1]}(x) d\mu(x) = \int (\chi_{[1]} \circ T)(x) d\mu(x) \quad (*) \\ &= \int \frac{1}{n} \sum_{i=0}^{n-1} \chi_{[1]}(T^i x) d\mu(x) = \int \frac{\sum_{i < n} x_i}{n} d\mu(x) \end{aligned}$$

Idea: Find a **long interval** that contains many short intervals in which the number of 1's is too large for $(*)$ to hold.



A Combinatorial Approach

Assume $x \in E$. Since $TE \subseteq E$, for every n there exists a minimal $s(n) = s_x(n)$ for which

$$\frac{x_n + x_{n+1} + \cdots + x_{s(n)-1}}{s(n) - n} \geq \mu[1] + \varepsilon.$$

Only a certain fraction of x has a large $s_x(0)$: Given $\delta > 0$, there exists L such that

$$\mu(D) \leq \delta^2, \text{ where } D = \{x: s_x(0) > L\}.$$

Hence if we give up a little bit of measure, we can work with sequences that obtain an ε -excess of 1's by a fixed time L .

A Combinatorial Approach

$$D = \{x : S_x(0) > L\}$$

Furthermore, not many orbits will visit D very often: Let $g_N(x)$ be the average number of visits of $T^i x$ to D by time N ,

$$g_N(x) = \frac{1}{N} \sum_{i=0}^{N-1} \chi_D(T^i x).$$

$\geq \mu(D)$

T measure preserving implies $\int \underline{g_N} d\mu \leq \delta^2$. Then Markov's inequality yields

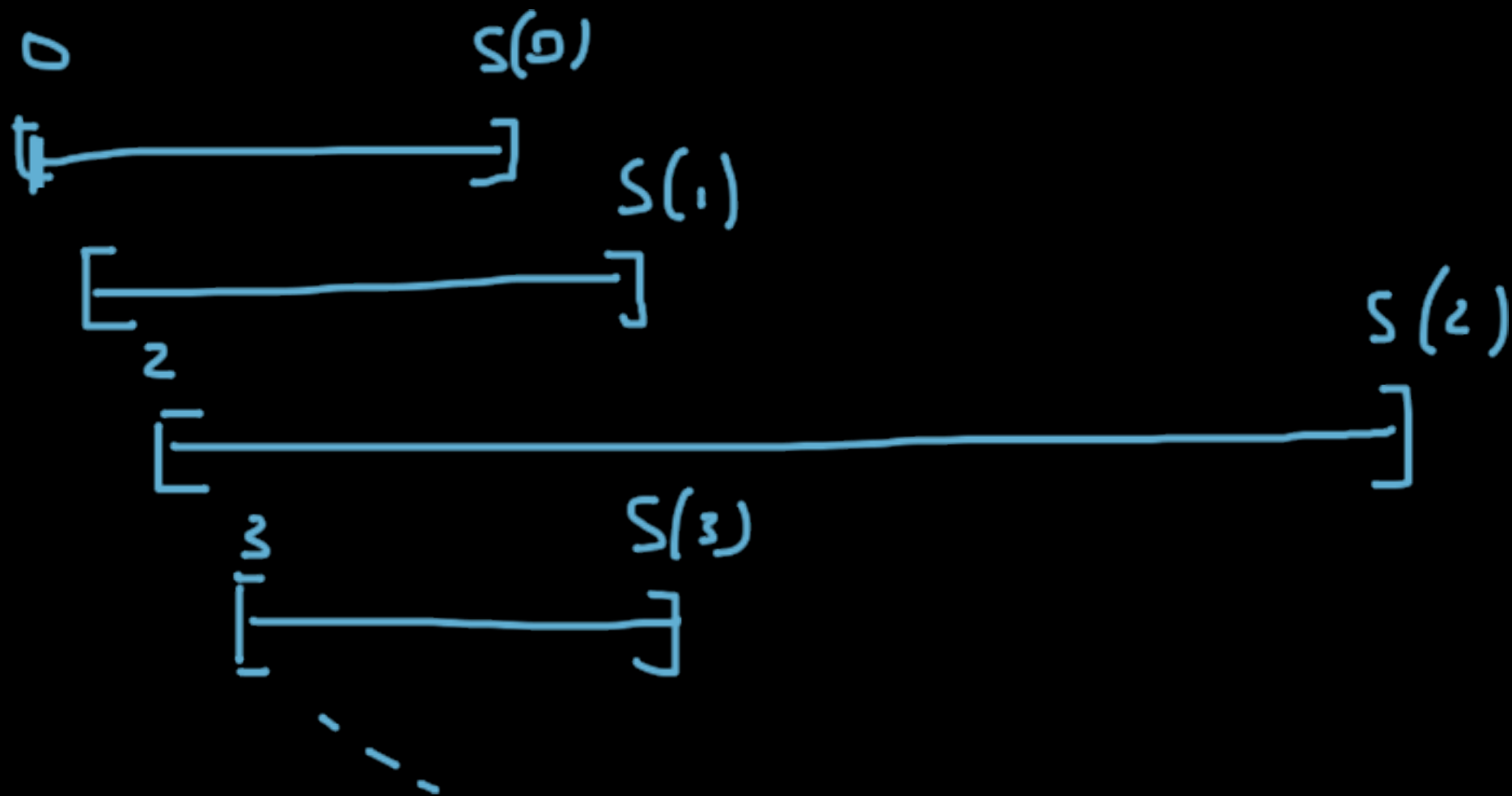
$$\mu(G_N) \geq 1 - \delta, \text{ where } G_N = \{x : g_N(x) \leq \delta\}.$$

$$\mu\{x : g_N \leq \delta\} \geq 1 - \frac{\int g_N d\mu}{\delta} \leq \delta^2$$
$$\geq 1 - \delta$$

A Combinatorial Approach

For $x \in G_N$, the portion of long intervals with starting point in $[0, N - 1]$ is not more than δ :

$$\frac{|\{n \in [0, N - 1] : s(n) - n > L\}|}{N} \leq \delta.$$



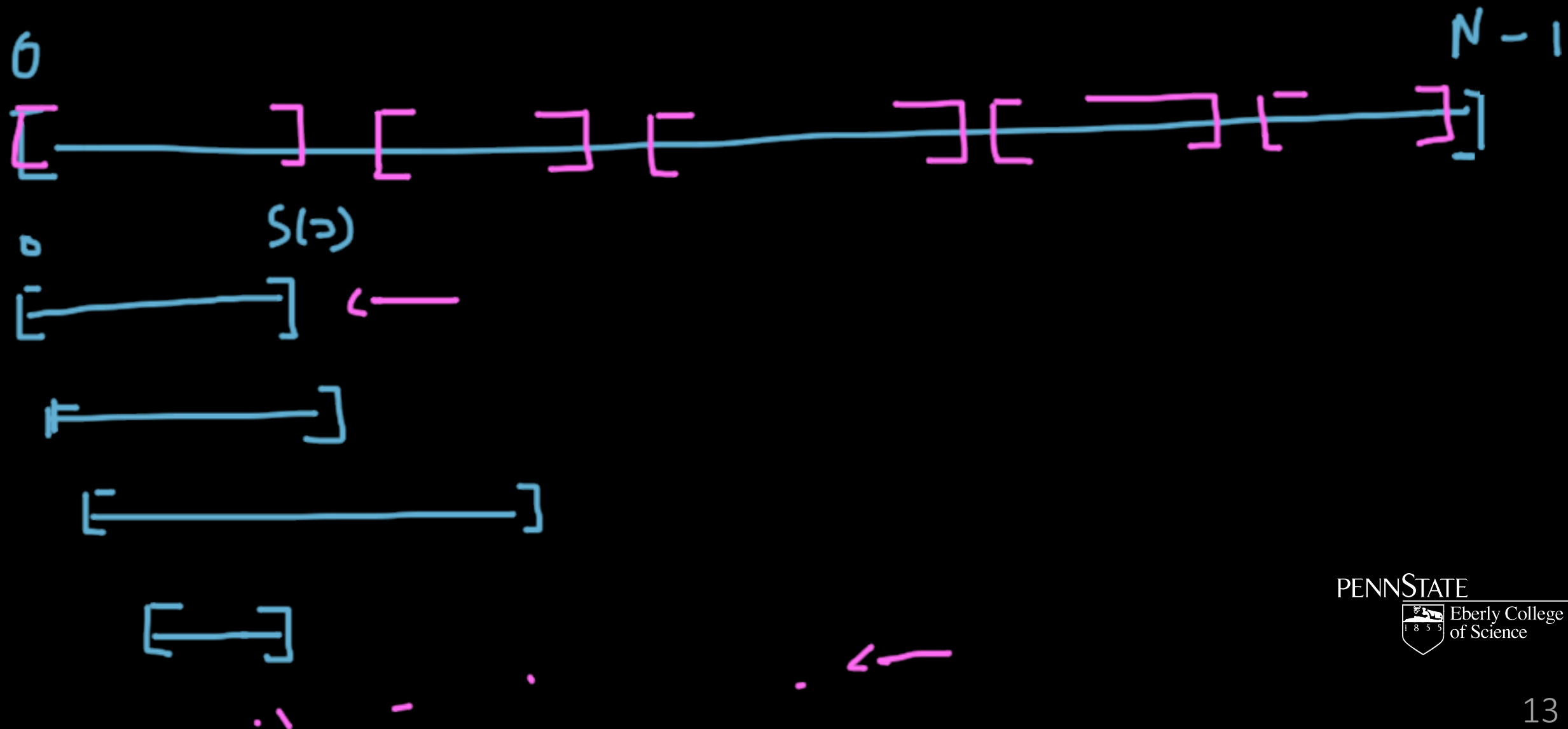
The Packing Lemma

LEMMA: Suppose $x \in G_N$. There exists a sequence $n_1 < n_2 < n_3 < \dots < n_k$ of natural numbers such that

(1) the intervals $[n_i, s_x(n_i)]$ are disjoint,

(2) $[n_i, s_x(n_i)] \subseteq [0, N-1]$ for all $i \leq k$,

(3) $|\bigcup_i [n_i, s_x(n_i)]| \geq \underline{(1-2\delta)N}$.



A Combinatorial Approach

The Lemma allows us to bound the number of 1's in $[0, N - 1]$ from below:

$$\sum_{j=0}^{N-1} x_j \geq \sum_{i=1}^k \sum_{j=n_i}^{s(n_i)} x_j \geq (1 - 2\delta)N(\mu[1] + \varepsilon).$$

Note that this bound is independent of x (though the sequence of n_i 's does depend on x), as long as $x \in G_N$. Hence we can bound the integral

$$\int \sum_{j=0}^{N-1} x_j \geq \underbrace{(1 - 2\delta)}_{\geq 1 - \delta} \underbrace{N(\mu[1] + \varepsilon)}_{\geq 1 - \delta} \mu(G_N).$$

Thus,

$$\underline{\mu[1]} = \int \frac{\sum_{i < N} x_i}{N} d\mu(x) \geq \underline{(1 - 2\delta)} \underline{(1 - \delta)} \underline{(\mu[1] + \varepsilon)},$$

which is impossible, since δ can be chosen arbitrarily small.