

Lesson 3

Dynamical Systems

3-2: Measurable Dynamics

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Stochastic Processes

A finite alphabet, $(X_n)_{n \in \mathbb{N}}$ A -valued process.

- Recall: joint distribution (P_n) , where

$$P_n(X_0 = a_0, \dots, X_n = a_n) \quad (\text{short: } P_n(a_0, \dots, a_n))$$

describes the distribution of the process up to time n , and is subject to the consistency requirement

$$P_n(a_0, \dots, a_n) = \sum_{a \in A} P_{n+1}(a_0, \dots, a_n, a).$$

- Kolmogorov Extension Theorem: Constructs an underlying measure μ on $A^{\mathbb{N}}$ such that

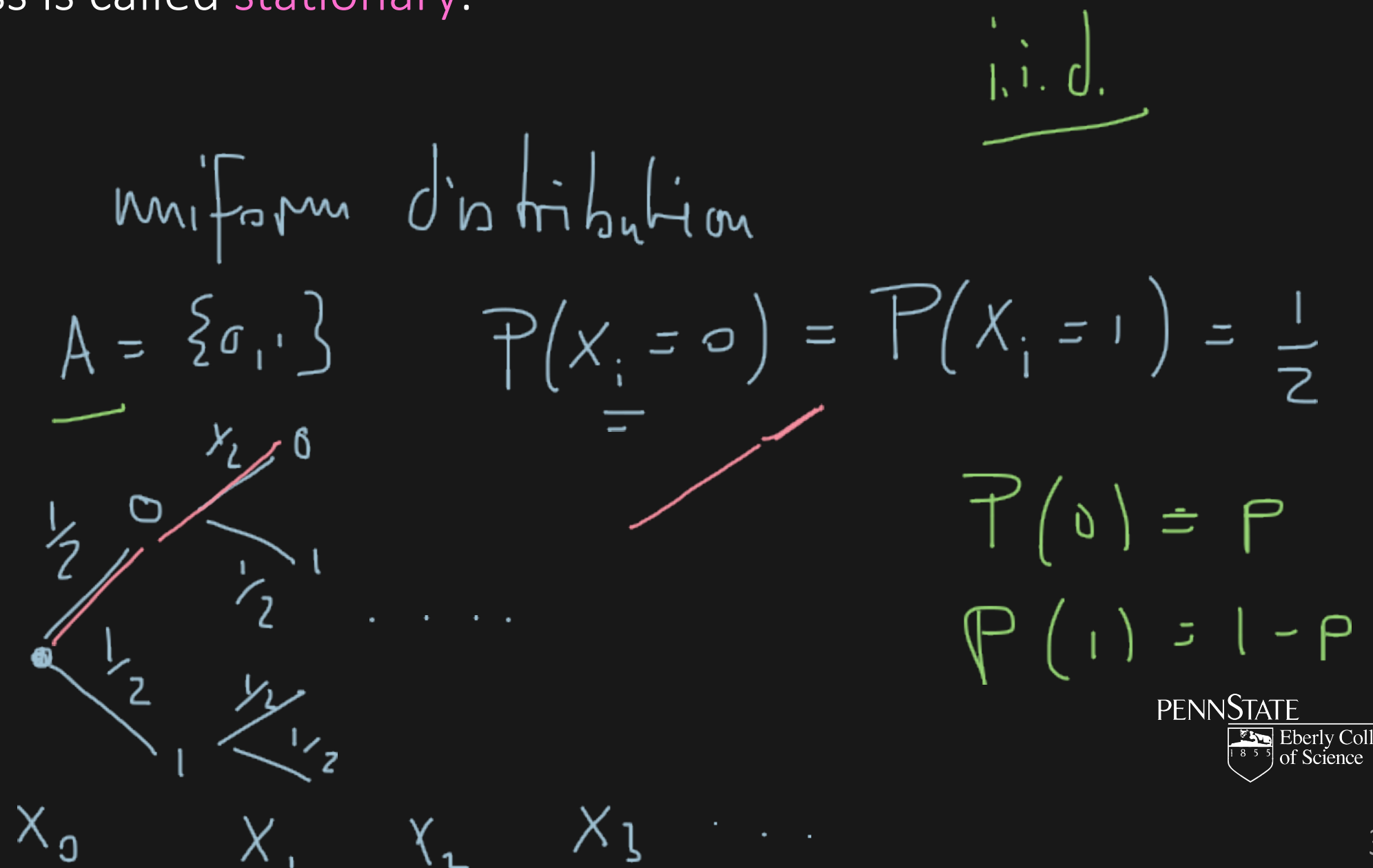
$$\mu[\sigma] = P_{|\sigma|}(\sigma).$$

Stationary Processes

In many cases, the joint distributions of the process will **not depend on the choice of time origin**, i.e. for all $m \leq n$, $a_i \in A$,

$$\text{Prob}(X_i = a_i, m \leq i \leq n) = \text{Prob}(X_{i+1} = a_i, m \leq i \leq n)$$

Such a process is called **stationary**.



From Stationarity to Shift-Invariance

Let (X_n) be stationary, and let $(A^{\mathbb{N}}, \mathcal{B}, \mu)$ its Kolmogorov extension (\mathcal{B} Borel sets on $A^{\mathbb{N}}$). Then μ is **shift-invariant**.

- Let T be the shift-map on $A^{\mathbb{N}}$. Then $\{x : x_0 = a_0, \dots, x_{k+n-1} = a_{n-1}\}$
- $$T^{-1}[a_0 \dots a_{n-1}]_k = [a_0 \dots a_{n-1}]_{k+1}.$$

Hence the preimage of a cylinder is a cylinder, and it follows that $T^{-1}(B)$ is Borel for any Borel set $B \subseteq A^{\mathbb{N}}$. $\Rightarrow T$ is **measurable**.

- Since (X_n) is stationary,

$$\mu[a_0 \dots a_{n-1}]_k = \mu[a_0 \dots a_{n-1}]_{k+1} = \mu(T^{-1}[a_0 \dots a_{n-1}]_k).$$

This in turn extends to all Borel sets:

$$\mu(T^{-1}(B)) = \mu(B) \quad \text{for all } B \subseteq A^{\mathbb{N}} \text{ Borel.}$$

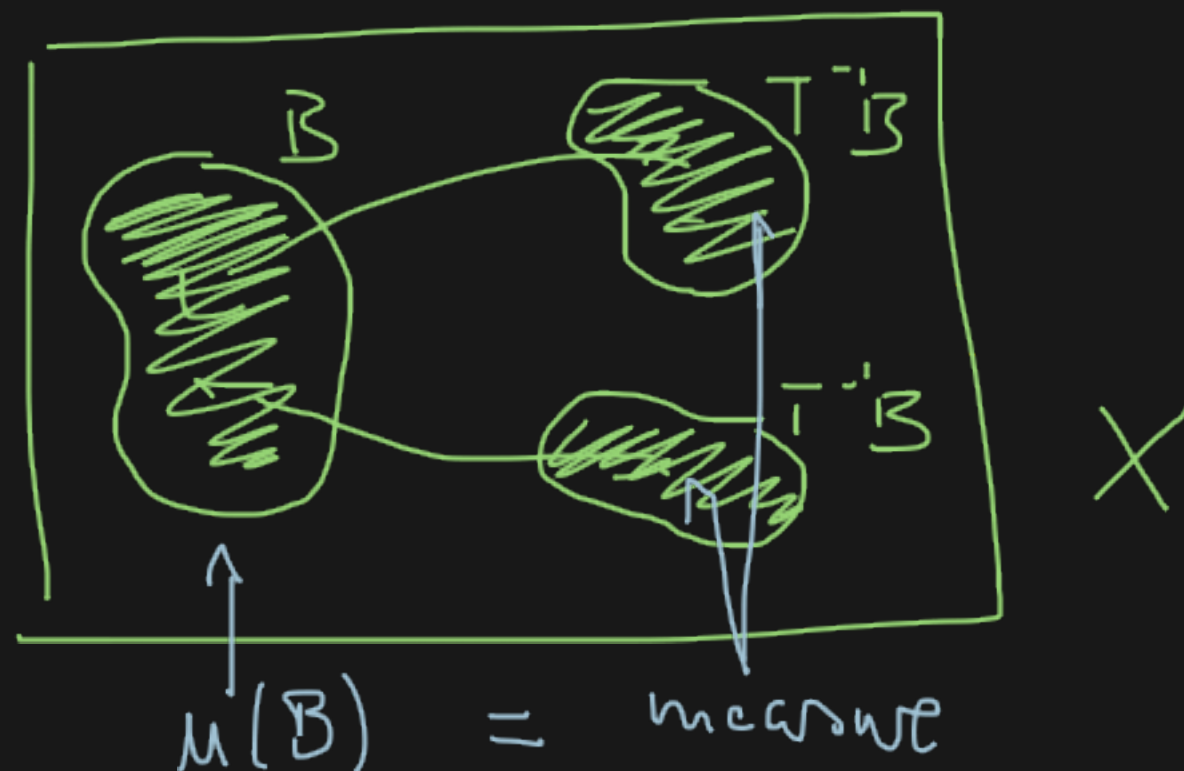
T is a μ -preserving transformation

Measure-theoretic Dynamical Systems

Together with the shift map T , the Kolmogorov measure space $(A^{\mathbb{N}}, \mathcal{B}, \mu)$ corresponding to a stationary process (X_n) forms a **measure-preserving dynamical system**.

In general, such a system is a tuple (X, \mathcal{A}, μ, T) , where (X, \mathcal{A}, μ) is any probability space (\mathcal{A} is a σ -algebra), and $T: X \rightarrow X$ is measurable and **measure-preserving**, i.e.

$$\mu(T^{-1}(B)) = \mu(B) \quad \text{for all } B \in \mathcal{A}.$$

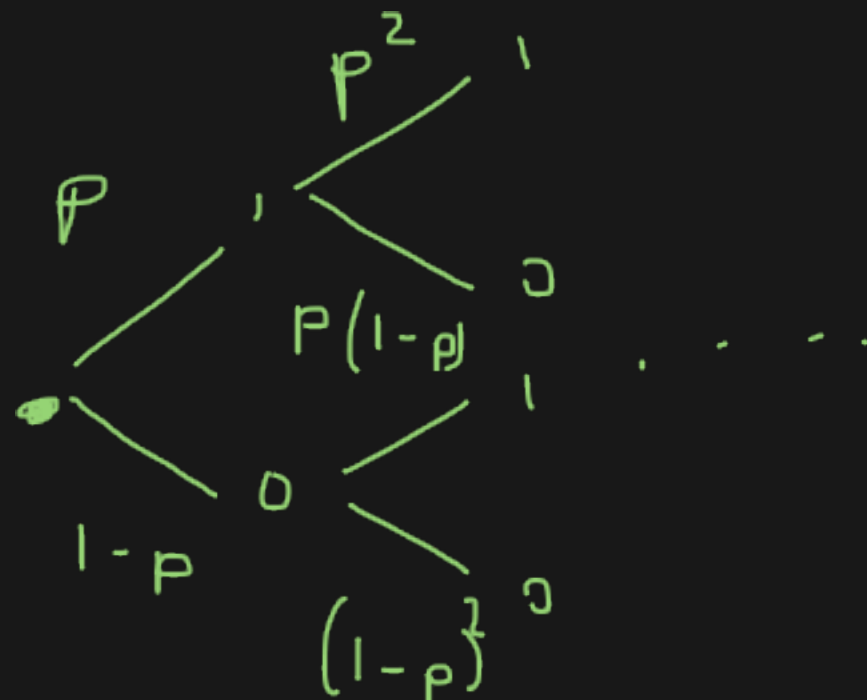


Example: Bernoulli Shifts

Let $A = \{0, 1\}$, and $0 \leq p \leq 1$. Then the measure μ_p given by

$$\mu_p[\sigma] = p^N (1 - p)^{|\sigma| - N},$$

where $N = \#\{i: \sigma(i) = 1\}$, is **shift-invariant**. The system $(A^{\mathbb{N}}, \mathcal{B}, \mu_p, T)$ is the most simple example of a **Bernoulli shift**.



- We can also look at the **two-sided Bernoulli shift** on $A^{\mathbb{Z}}$. This has the advantage that the shift is **invertible**.

Example: Bernoulli Shifts

More generally, if (X, \mathcal{A}, μ) is a probability space, then

$$(Y, \mathcal{F}, \nu) = \prod_{i=-\infty}^{\infty} (X, \mathcal{A}, \mu)$$

is invariant under the shift $T: Y \rightarrow Y$, where

$T(y) = T((x_n)_{n \in \mathbb{Z}}) = (z_n)_{n \in \mathbb{Z}}$ with $z_n = x_{n+1}$. This is called the (two-sided) **Bernoulli shift with state space (X, \mathcal{A}, μ)** .

From Shifts to Processes

If μ is a shift-invariant measure on $A^{\mathbb{N}}$, then we can **derive a stationary process** from it as follows:

- ▶ **Partition** $A^{\mathbb{N}}$ into $\mathcal{P} = \{P_a : a \in A\}$, where $P_a = \{x : x_0 = a\}$.
- ▶ Define the **A-valued random variable** $X_{\mathcal{P}}$ by mapping $x \in A^{\mathbb{N}} \mapsto X_{\mathcal{P}}(x) = a$ where a is such that $x \in P_a$.
- ▶ The random variable X_n is then given by

$$X_n(x) = X_{\mathcal{P}}(T^n(x)) \quad (n \geq 1).$$

$\cong [a]$

Kolmogorov
partition

