

CYS 2020/2021

- Definition of a stochastic process
- Definition of a Markov Chain
 - state space S
 - one-step transition probability
 - initial distribution
 - finite-dimensional distributions
 - some simple examples
- Transition probability matrices
- First Step Analysis
- Absorbing Markov Chains

Stochastic process

$X : \Omega \rightarrow \mathbb{R}$ random variable

$n=2$ $(X, Y) : \Omega \rightarrow \mathbb{R}^2$

$(X_1, X_2) : \Omega \rightarrow \mathbb{R}^2$

$\omega \mapsto (X_1(\omega), X_2(\omega))$

marks
feenle

grade of the first exam

$n \in \mathbb{N}$ (X_1, \dots, X_n)

$\bar{T} = \{0, 1, 2, \dots, n, \dots\}$, $T = [0, +\infty)$

$n \in \bar{T}$ X_n

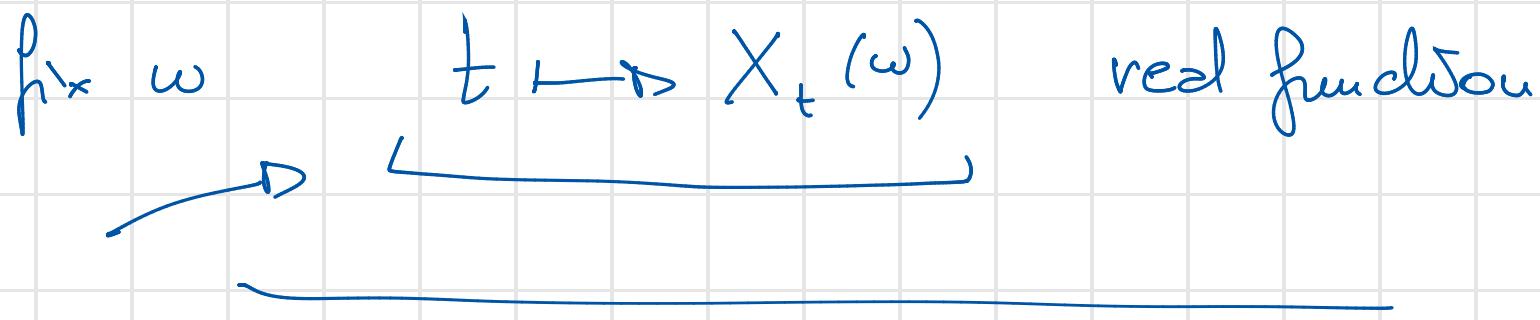
$\rightarrow X_0, X_1, X_2, \dots$ infinite sequence
fix ω $\boxed{\text{path}}$ stochastic process

$X = (X_0, X_1, \dots) : \Omega \times \bar{T} \rightarrow \mathbb{R}$

$\rightarrow \underline{X_n(\omega)} \in \mathbb{R}$

$$X: \Omega \times \bar{T} \rightarrow \mathbb{R} \quad \bar{T} = [0, +\infty)$$

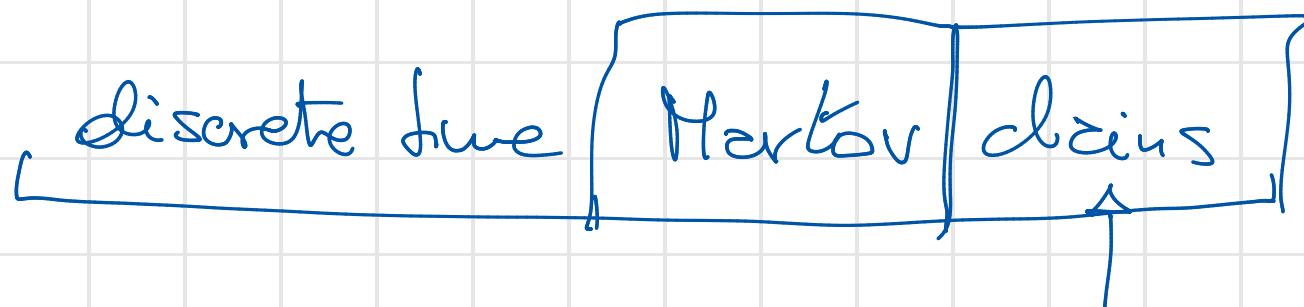
$X_t(\omega)$ random variable (fixed $t \in \bar{T}$)



discrete time stochastic process

$$X: \Omega \times \mathbb{N} \rightarrow \mathbb{R}$$

X_0, X_1, X_2, \dots



$$X_n: \Omega \rightarrow S$$

discrete space
finite or countable set
infinite

$$S = \mathbb{N}$$

$$\textcircled{e} \quad P[X_{n+1} = j \mid X_0 = i_0, X_1 = i_1, \dots, X_{n-1} = i_{n-1}, X_n = i]$$

$$X_n: \Omega \rightarrow \mathbb{N}$$

$$= P[X_{n+1} = j \mid X_n = i]$$

$\forall n \in \mathbb{N}, i, j, i_0, \dots, i_{n-1} \in \mathbb{N}$

One-step transition probability $P_{ij}^{n, n+1}$

$$P_{ij}^{n, n+1} := P[X_{n+1} = j \mid X_n = i]$$

$$X_n = i \xrightarrow{\quad} X_{n+1} = j$$

Stationary transition probabilities

$$P_{ij}^{n, n+1} := \boxed{P_{ij}} \quad \forall n \in \mathbb{N}$$

(homogeneous in time)

$$P_{ij} := P[X_1 = j \mid X_0 = i] \quad i, j \in \mathbb{N}$$

- $P_{ij} \geq 0$

- P_{ijj}

$$\sum_{j \in \mathbb{N}} P_{ijj} = \sum_{j \in \mathbb{N}} P[X_1 = j \mid X_0 = i]$$

$\{X_1 = j\}$ are disjoint events

$$\forall j \in \mathbb{N}$$

$$= P[\bigcup_j \{X_1 = j \mid X_0 = i\}] = 1$$

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- $P_{ij} \geq 0$

$$\forall i, j \in \mathbb{N}$$

- $\sum_{j \in \mathbb{N}} P_{ij} = 1$

$$\forall i \in \mathbb{N}$$

fixed i

$$P[A|B] = \frac{P[A \cap B]}{P[B]}$$

$$A \mapsto P[A|B]$$

↑
is \approx probability

$$S = \{0, 1, 2\}$$

$$P_{ij}, i, j \in S$$

$$P = \begin{bmatrix} P_{00} & P_{01} & P_{02} \\ P_{10} + P_{11} + P_{12} & & \\ P_{20} & P_{21} & P_{22} \end{bmatrix} = 1 \text{ transition prob.}$$

Stochastic Matrix

Non-negative matrix

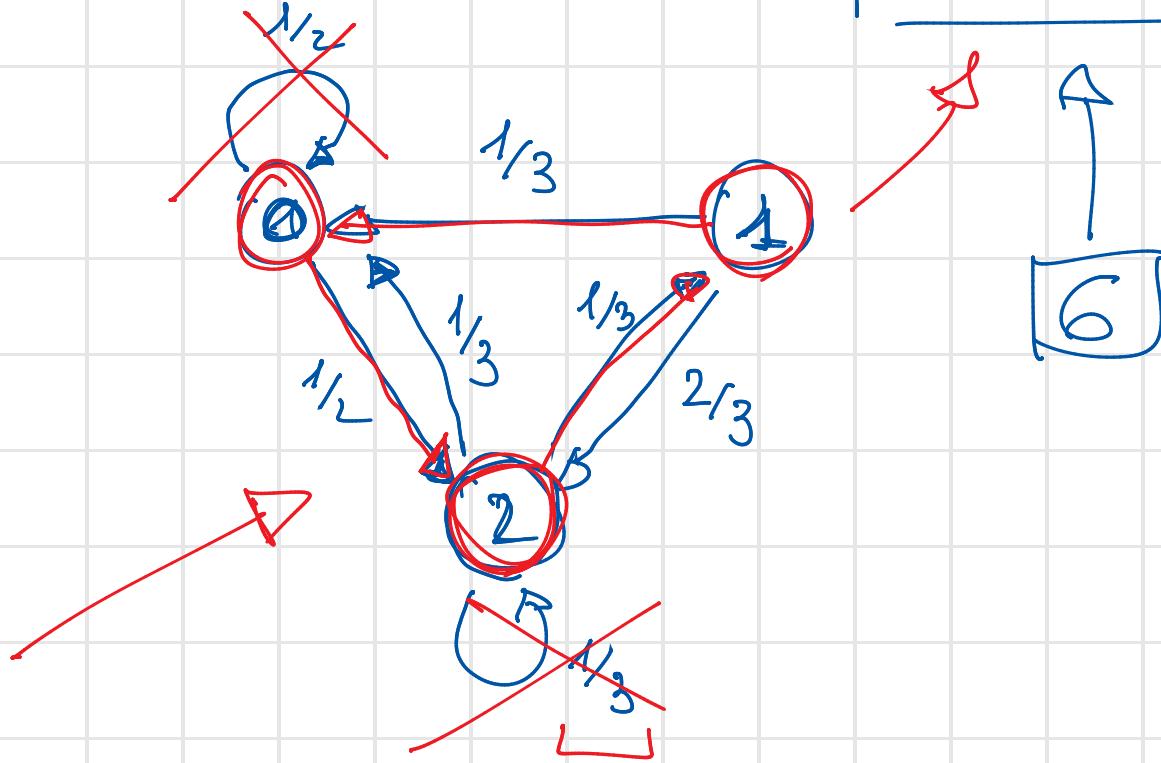
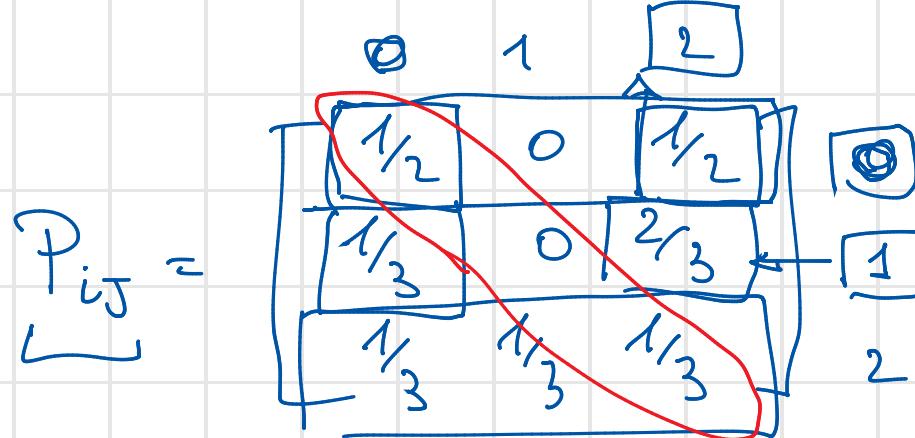
$$\text{Fix } i \quad \sum_j P_{ij} = 1$$

$$S = \mathbb{N}$$

$$\begin{bmatrix} P_{00} & P_{01} & P_{02} & \dots & \dots & \dots \\ P_{10} & P_{11} & P_{12} & \dots & \dots & \dots \\ P_{20} & P_{21} & P_{22} & \dots & \dots & \dots \\ \vdots & \vdots & \vdots & \ddots & \ddots & \ddots \\ \sum_{j=0}^{+\infty} P_{ij} = 1 & & & & & \end{bmatrix}$$

$$\sum_{n=1}^{\infty} \frac{1}{n} = +\infty, \quad \sum_{n=1}^{\infty} \frac{1}{n^2} < \infty$$

Representation



$$S = \{0, 1, 2\}$$

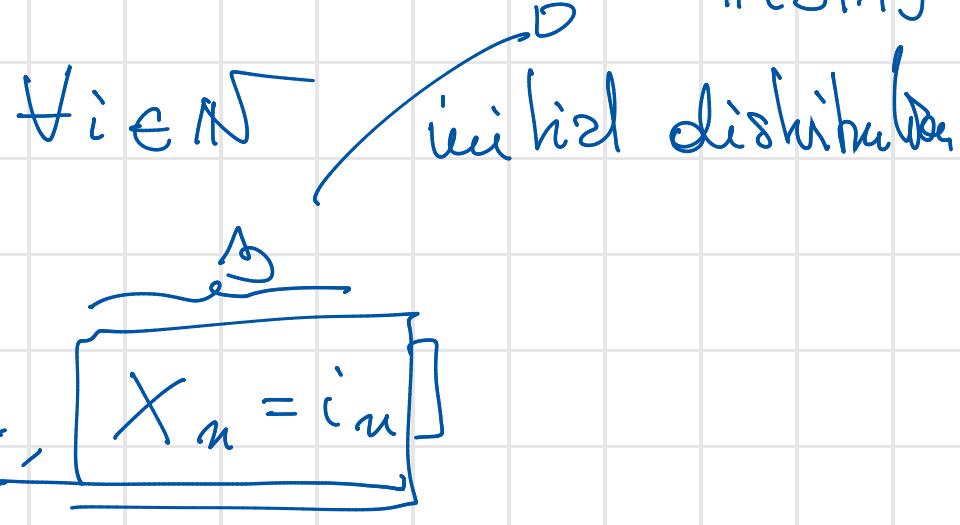
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Finite dimensional probabilities

$$P[X_0 = i] = p_i$$

$$P[X_0 = i_0, X_1 = i_1, \dots, X_n = i_n]$$

= $P[X_0 = i_0, \dots, X_{n-1} = i_{n-1}]$.

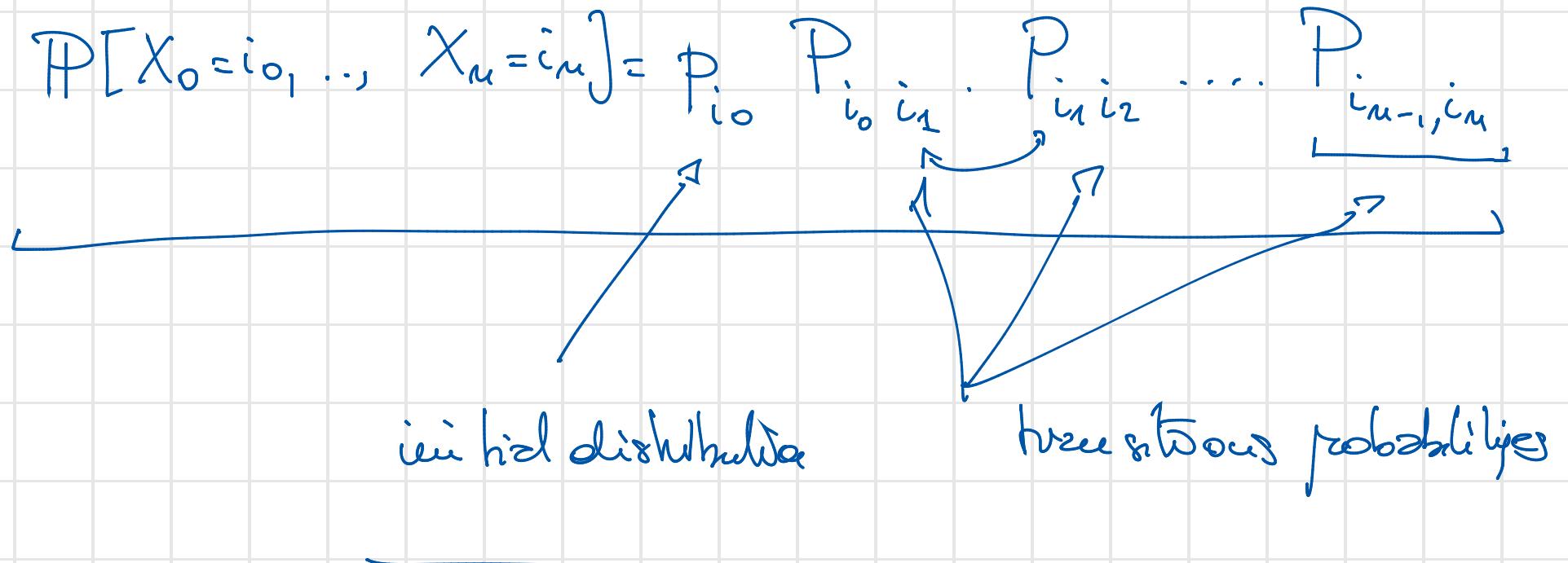


Markov property

$$P[X_n = i_n | X_0 = i_0, \dots, X_{n-1} = i_{n-1}]$$

= $P[X_0 = i_0, \dots, X_{n-1} = i_{n-1}] \cdot P[X_n = i_n | X_{n-1} = i_{n-1}]$

$$P_{i_{n-1}, i_n} = P[X_n = i_n | X_{n-1} = i_{n-1}] \dots$$



Matrix description of Transition probabilities

$$\rightarrow P = (P_{ij})_{i,j \in \mathbb{N}} \quad \text{one-step transition}$$

$$P^{(n)} = (P_{ij}^{(n)})_{i,j \in \mathbb{N}}$$

$$\rightarrow P_{ij}^{(n)} = P[X_n = j | X_0 = i] \quad \begin{matrix} n \in \mathbb{N} \\ i, j \in \mathbb{N} \end{matrix}$$

n -step transition probabilities

Theorem 2.1 (pg 101)

The n -step transition probabilities of a MC

satisfy

$$P_{ij}^{(n)} = \sum_{k=0}^{\infty} P_{ik}^{(n)} P_{kj}^{(n-1)}$$

where we define $P_{ij}^{(0)} = \text{Id}$

$$\begin{aligned} P_{ij}^{(n)} &= P \times -P^{(n-1)} & \sum_k a_{ik} b_{kj} \\ &= P \times (P \times P^{(n-2)}) & \\ &= P \times P \times (P \times P^{(n-3)}) & \\ \dots &= P^n & n\text{-th power of} \\ && \text{the matrix } P \end{aligned}$$

Proof: first-step analysis

$$\underline{P[A \cap B] = P[A] \cdot P[B|A]}$$

$$\begin{aligned}
 P_{ij}^{(n)} &= P[X_n=j | X_0=i] \\
 &= \sum_{k=0}^{\infty} P[X_n=j, X_1=k | X_0=i] \\
 &= \sum_{k=0}^{\infty} P[X_1=k | X_0=i] \cdot P[X_n=j | X_1=k, X_0=i] \\
 &= \sum_{k=0}^{\infty} P_{ik} P_{kj}^{(n-1)}
 \end{aligned}$$

(Conclusions)

$$P[X_n=k] = \sum_{j=0}^{\infty} P_j P_{jk}^{(n)}$$

$$P_j = P[X_0=j]$$

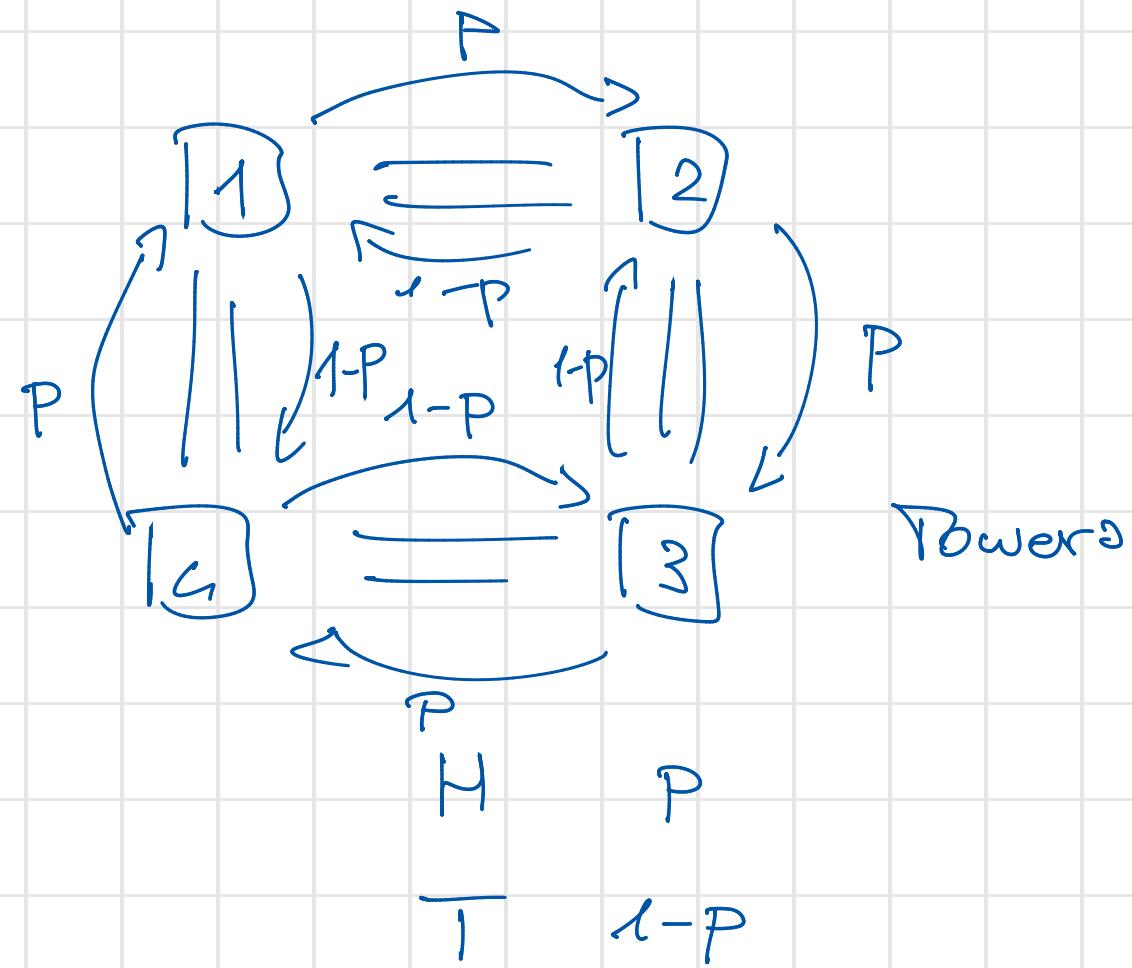
$$\underline{\pi = (P_0, P_1, \dots, P_n, \dots)}$$

$$\underline{\pi \cdot P^n = P[X_n=k]}$$

Remark in general $A \cdot B \neq B \cdot A$.
 A, B matrices

Ex:

Castle



Seehund

$$n \in \mathbb{N}$$

$$X_n \in \{1, 2, 3, 4\}$$

	1	2	3	4
1	0	P	0	$1-P$
2	$1-P$	0	P	0
3	0	$1-P$	0	P
4	P	0	$1-P$	0

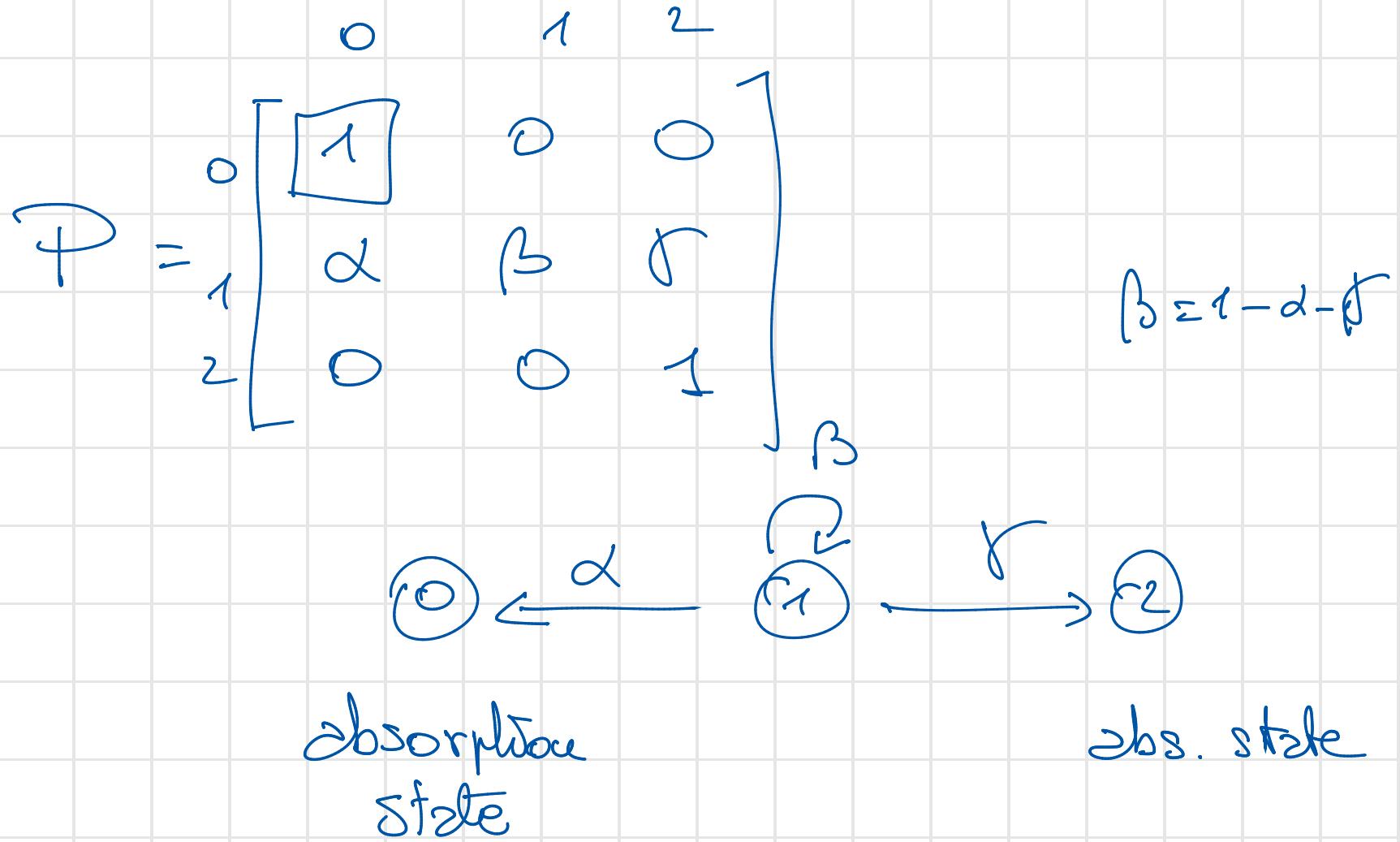
Exercise : $P = \frac{1}{2}$ and compute P^n

$\forall n \in \mathbb{N}$

Section 4

First Step Analysis

$$\mathcal{S} = \{0, 1, 2\}$$



If I start from 1, what is the

prob. that I goes (forever) to 0?

to 2?

$$\overline{T} = \min_{t \in \mathbb{N}} \left\{ n \geq 0 : X_n = 0 \text{ or } X_n = 2 \right\}$$

true

$$u = \mathbb{P}[X_T = 0 \mid X_0 = 1]$$

$$\sigma = \mathbb{E}[T \mid X_0 = 1]$$



$$\mathbb{P}[X_T = 0 \mid X_1 = 0] = 1$$

$$\mathbb{P}[X_T = 0 \mid X_1 = 2] = 0$$

$$\mathbb{P}[X_T = 0 \mid X_1 = 1] = u$$

$$u = \mathbb{P}[X_T = 0 \mid X_0 = 1] =$$

$$= \sum_{k=0}^2 \mathbb{P}[X_T = 0 \mid X_1 = k, X_0 = 1] \cdot \mathbb{P}[X_1 = k \mid X_0 = 1]$$

$$= 1 \cdot \alpha + u \cdot \beta + 0 \cdot \gamma$$

$$u = \alpha + \beta \cdot u$$

$$u = \frac{\alpha}{1 - \beta} = \frac{\alpha}{\alpha + \gamma}$$

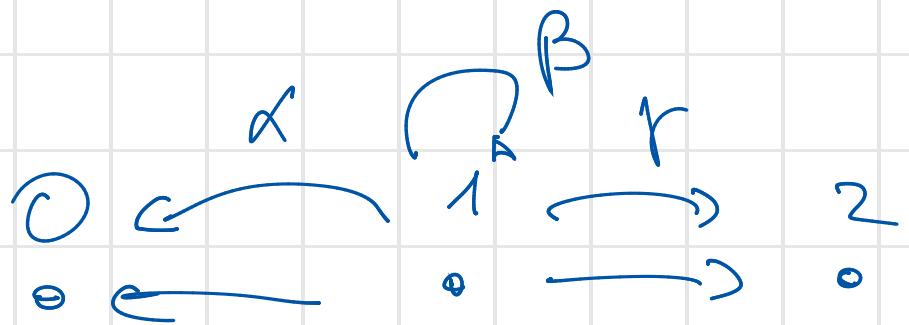
Exercise

$$\omega = \mathbb{P}[X_T = 2 \mid X_0 = 1] = 1 - u = \frac{\gamma}{\alpha + \gamma}$$

$$U = \mathbb{E}[T \mid X_0 = 1]$$

expected
absorption
time

$$X_1 = n$$



$$U = \underbrace{1}_{\text{initial state}} + \alpha \cdot 0 + \gamma \cdot 0 + \beta \cdot U$$

$$U = \mathbb{E}[T \mid X_0 = 1]$$

$$U = \mathbb{E}[T \mid X_1 = 1]$$

$$U = 1 + \beta U$$

$$U = \frac{1}{1-\beta} = \frac{1}{\alpha+\gamma}$$

pag. 118 - 119 - 120

(since $\alpha + \beta + \gamma = 1$
 $\Rightarrow 1 - \beta = \alpha + \gamma$)

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