

Recall: We study the Gambler's ruin problem ①

$$p(n) = P(R \mid X_0 = n)$$

We showed (by conditioning on the first event)
that $p(n) = p(n+1)p + p(n-1)(1-p)$ $0 \leq n \leq N$
with boundary conditions $p(0) = 1$ and $p(N) = 0$

To solve this recurrence equation, the general method is to first solve the associated characteristic equation (see week 2 online page)

$$X = X^2 p + (1-p) \Leftrightarrow pX^2 - X + (1-p) = 0$$

→ the roots allow to find a general solution of $p(n) = p(n+1)p + p(n-1)(1-p)$ that depends on 2 constants.

→ To determine the constant, we use the boundary conditions.

Case $p = \frac{1}{2}$: By solving $pX^2 - X + (1-p) = 0$ with $p = \frac{1}{2}$
we find $\Delta = 0 \Rightarrow$ one root $X = \frac{1}{2p} = 1$

In this case the general solution is

$$p(n) = \alpha X^n + \beta \cdot n X^n, \text{ where } \alpha \text{ and } \beta \text{ are constants}$$

$$\Rightarrow p(n) = \alpha + \beta \cdot n, \text{ and}$$

(2)

$$\begin{cases} p(0) = 1 \\ p(N) = 0 \end{cases} \Leftrightarrow \begin{cases} \alpha \cdot 1^0 + \beta \cdot 0 \cdot 1^0 = 1 \\ \alpha \cdot 1^N + \beta \cdot N \cdot 1^N = 0 \end{cases}$$

$$\Leftrightarrow \begin{cases} \alpha = 1 \\ \beta = -\frac{1}{N} \end{cases}$$

$$\Rightarrow p(n) = 1 - \frac{n}{N}$$

Conclusion: $P(\text{ruin} | X_0 = n) = \begin{cases} 1 - \frac{a^n - 1}{a^N - 1} & \text{if } p \neq \frac{1}{2} \text{ and where } a = \frac{1-p}{p} \\ 1 - \frac{n}{N} & \text{if } p = \frac{1}{2} \end{cases}$

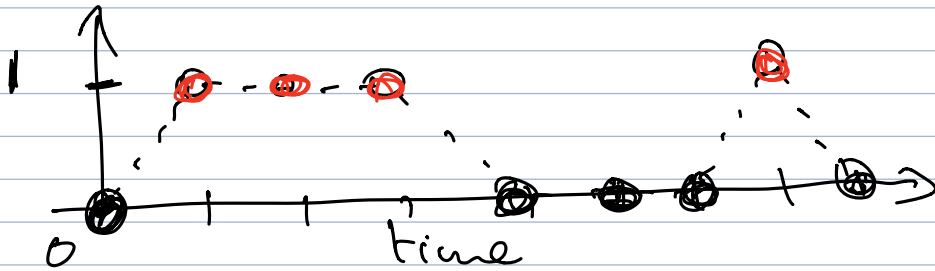
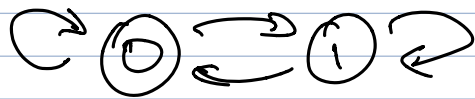
- We saw that the states in $\{1, \dots, N-1\}$ were transient (i.e. the probability of revisiting the state is < 1). Can we find this probability f_i ?

Answer: YES! \rightarrow see Jupyter Notebook and more properties on recurrence and transience.

Properties of transience and recurrence

Let $N_i = \# \{ n \geq 0 \mid X_n = i \}$

ex: If X_n is a M-C with transition diagram: (3)



- N_0 = How many times X_n was eq. to 0
- N_1 = " " " " " " " " " " " "

Remark: • N_i is a random variable on $\mathbb{N} \cup \{+\infty\}$

$$N_i = \sum_{n=0}^{+\infty} \mathbb{1}_{(X_n=i)},$$

where $\mathbb{1}_{(X_n=i)}$ (indicator function) $\approx \begin{cases} 1 & \text{if } X_n=i \\ 0 & \text{else} \end{cases}$

$$E(\mathbb{1}_{(X_n=i)}) = \Pr(X_n=i)$$

$$= 1 \cdot \Pr(\mathbb{1}_{(X_n=i)}=1)$$

$$+ 0 \cdot \Pr(\mathbb{1}_{(X_n=i)}=0)$$

Prop: (i) If i is recurrent, then $P(N_i = \infty | X_0 = i) = 1$ ⁽⁴⁾

(ii) If i is transient and $X_0 = i$,
then $N_i \sim \text{Geom}(1 - f_i)$

$$\text{(i.e. } P(N_i = m) = (1 - f_i) \cdot f_i^{m-1}, m \geq 1)$$

$$\text{and in particular } E(N_i) = \frac{1}{1 - f_i}$$

Interpretation: A recurrent state is re-visited ∞ by many times, while for a transient state, the chain doesn't revisit the state after a certain time

Proof: (i) $f_i = 1 \Rightarrow$ the process returns to i w.p. 1
so $P(N_i \geq 1 | X_0 = i) = 1$, but after returning, the process essentially starts afresh, so it returns again
so $P(N_i \geq 2 | X_0 = i) = 1$ etc.

$$\text{so } \forall m \quad P(N_i \geq m | X_0 = i) = 1$$

$$\Rightarrow \forall k \in \mathbb{N} \quad P(N_i = k | X_0 = i) = 0$$

$$\text{so } \underline{P(N_i = \infty | X_0 = i) = 1} \quad \checkmark$$

(ii) • let's assume $X_0 = i$, so $P(N_i = 1) = 1 - f_i$ ^(s)
 (by def. of f_i)

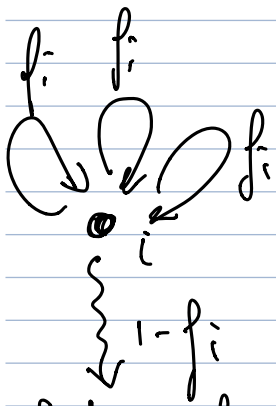
• For $N_i = 2$

\Rightarrow the chain returns once and never after that, so $P(N_i = 2) = f_i \cdot (1 - f_i)$

• For $N_i = 3$

\Rightarrow the chain returns twice and never after, so $P(N_i = 3) = f_i^2 (1 - f_i)$

(...) $P(N_i = n) = f_i^{n-1} (1 - f_i) \checkmark$



(Picture for $N_i = 4$)

Rank : • if i is recurrent $E(N_i | X_0 = i) = +\infty$

• if the state space is finite, then some state must be recurrent

(Proof : if all states are transient, there is no state visited after a certain time)
 Contradiction!

We now characterize recurrence and transience using n -step transition probabilities.

Prop: i is $\begin{cases} \text{recurrent if } \sum_{n=0}^{+\infty} P_{ii}^n = \infty \\ \text{transient if } \sum_{n=0}^{+\infty} P_{ii}^n < +\infty \end{cases}$ (6)

Proof: $E(N_i | X_0 = i) = \sum_{n=0}^{+\infty} E(\mathbb{1}_{(X_n = i)} | X_0 = i)$
 $\quad \quad \quad \uparrow$
 $\sum_{n=0}^{+\infty} \mathbb{1}_{(X_n = i)}$ by def. $= \sum_{n=0}^{+\infty} P(X_n = i | X_0 = i)$
 $= \sum_{n=0}^{+\infty} P_{ii}^n \quad \square$

Corollary: if i is recurrent and $i \leftrightarrow j$, then j is recurrent

Important consequence: Recurrence and transience are **class properties** (all states in a communicating class are either all recurrent or all transient).

Proof: (exercise) \rightarrow (use the Chapman-Kolmogorov equation as in previous proofs)
 \rightarrow if $\sum P_{ii}^n > \infty$ and $i \leftrightarrow j$, show that $\sum P_{jj}^n = \infty$

As recurrence and transience are class properties, we can also directly conclude about these

when we observe a certain type of class. (7)

Def: A communication class is **closed** if $P_{ij} = 0$ whenever $i \in C$ and $j \notin C$ (i.e. there is no escape from C)

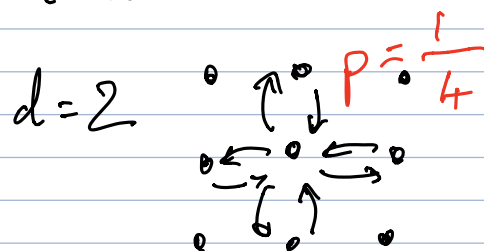
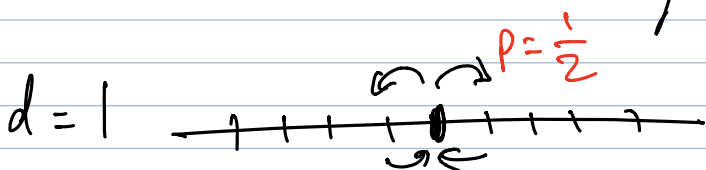
Remark: if $i \rightarrow j$, but $j \not\rightarrow i$, then i is transient.
(exercise, see also practice problems for week 3)

Prop: (i) A communication class which is not closed is always transient
(ii) A **closed finite** communication class is always recurrent

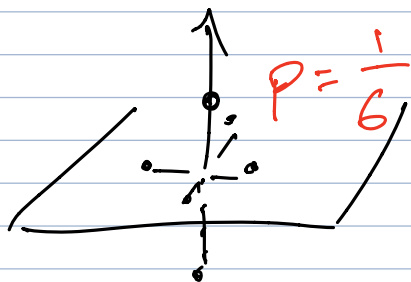
Remark: For an infinite communication class, it is harder to conclude, and we need to rely on the previous proposition, as we will see with the following example.

Example: Random walk in \mathbb{Z}^d

• We consider the uniform R-W in \mathbb{Z}^d



$$d=3$$



In general, $p = \frac{1}{2d}$

(8)

Q: Is the process recurrent or transient?

Result: The walk is $\begin{cases} \text{recurrent if } d = 1, 2 \\ \text{transient if } d = 3, 4, 5, \dots \end{cases}$