

Stochastic Modeling 9, 05.10.09

Note Title

5.10.2009

Proposition $(f_i)_{i=0}^{\infty}$ homogeneous
Markov chain

$$X = X_T \cup X_P, \quad X_T \cap X_P = \emptyset$$

$$X_P = \bigcup_m C_m$$

↑ disjoint, closed,
irreducible.

Proof

$$k \in X_P$$

$$C(k) = \{ e \in X, k \rightarrow e \}$$

k is persistent: so $k \rightarrow e \Rightarrow \exists i \geq 1$

$$\mathbb{P}(f_i = k \mid f_0 = e) > 0$$

$$k \rightarrow k \Rightarrow \boxed{k \in C(k)}$$

$C(k)$ is closed $m \in C(k), e \in X \setminus C(k)$

assume $\mathbb{P}(f_n = e \mid f_0 = m) > 0$
for some $n \geq 1$.

so $m \rightarrow e$, but because
of $k \rightarrow m$ we have $k \rightarrow e$.

This implies that $e \in C(k)$ by df
of $C(k)$.

$C(k)$ is irreducible $e, m \in C(k)$

$k \rightarrow e$ by df of $C(k)$
 $e \rightarrow k$ by persistence of k
and $k \rightarrow e$ } $k \leftrightarrow e$

In the same way $k \leftrightarrow m$

\leftrightarrow equivalence relation, especially

$$m \leftrightarrow e$$

Either $C(k) \cap C(e) = \emptyset$ or $C(k) = C(e)$

$m \in C(k) \cap C(e)$. Then $k \rightarrow m$
 $m \leftrightarrow \bar{e} \quad \forall \bar{e} \in C(e)$

$$\Rightarrow k \rightarrow \bar{e} \quad \forall \bar{e} \in C(e)$$

By df of $C(k)$ $\bar{e} \in C(k) \quad \forall \bar{e} \in C(e)$

$$\text{or } C(e) \subseteq C(k)$$

By symmetry

$$\text{so } C(k) \subseteq C(e)$$

$$C(k) = C(e)$$



Summary of the classification

arithmetic properties of T^n

$$\text{or } p^{(n)}(k, e) =: p_{k, e}^{(n)}$$

$$\boxed{k \text{ is absorbing:}} \quad p_{kk} = 1$$

$$\boxed{k \rightarrow e} \quad \exists n \geq 1, \quad p_{ke}^{(n)} > 0$$

$$\boxed{k \leftrightarrow e} \quad k \rightarrow e \text{ and } e \rightarrow k$$

k has the period $d(k)$

$d(k)$ is the largest m such that

$$p_{kk}^{(n)} = 0 \quad \text{for all } n \text{ which are not multiples of } m.$$

decomposition of X

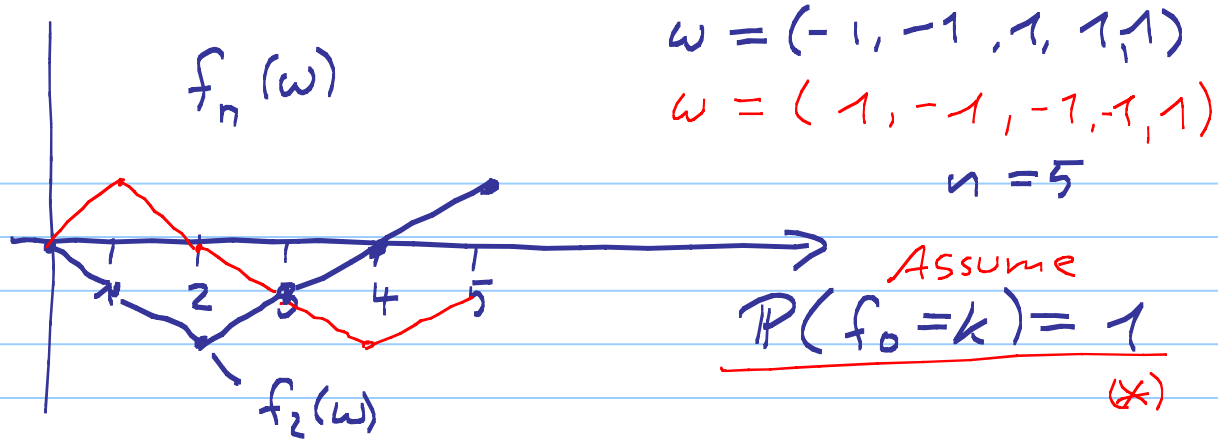
$$X = \bigcup_m X_m$$

↑
disjoint

$$k \rightarrow l \iff k, l \in X_m \text{ for some } m$$

Classification according to
asymptotic properties of $P_{ke}^{(n)}$

Recall $\mathcal{D}_n = \{ w = (\varepsilon_1, \dots, \varepsilon_n) \mid \varepsilon_i = \{1, -1\} \}$



$$T_k(w) = \inf \{ n \geq 1; f_n(w) = k \}$$

$$T_0(w) = 2 \quad T_0(w) = 4,$$

T_k is random (depends on w)
 is the recurrence time

k is persistent (recurrent) $\Leftrightarrow f_{kk} = 1$.

$$f_{kk} = \sum_{n=1}^{\infty} f_{kk}^{(n)}$$

$$f_{kk}^{(n)} = \mathbb{P}(f_n = k, f_{n-1} \neq k, \dots, f_1 \neq k | f_0 = k)$$

$$= \mathbb{P}(f_n = k, f_{n-1} \neq k, \dots, f_1 \neq k)$$

(*)

$$= \mathbb{P}(T_k = n)$$

So if k is persistent

$$\mathbb{P}_{T_k} = \sum_{n=1}^{\infty} f_{kk}^{(n)} \delta_n$$

discrete measure

$$\delta_n(B) = \begin{cases} 1 & n \in B \\ 0 & n \notin B \end{cases}$$

$$\mathbb{P}_{T_k}(\{e\}) = \mathbb{P}(T_k = e) = f_{kk}^{(e)}$$

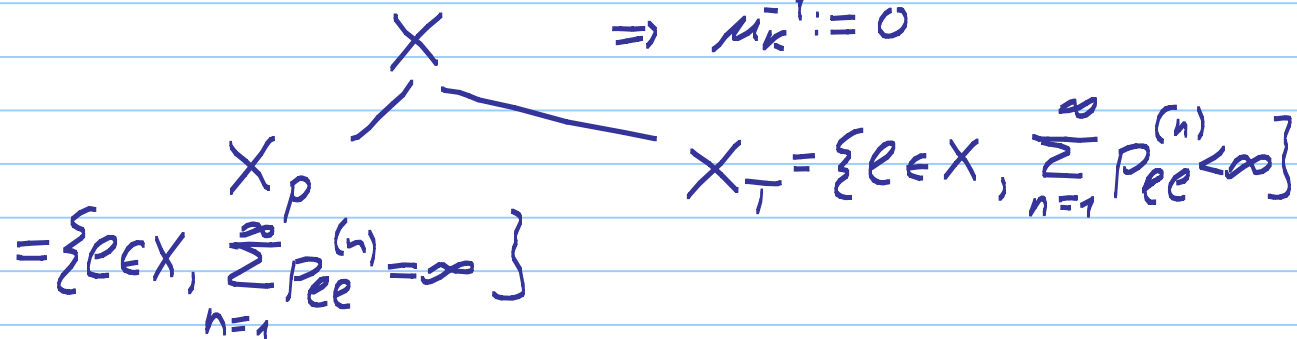
$$\sum_{n=1}^{\infty} f_{kk}^{(n)} \delta_n(\{e\}) = 0 + 0 + f_{kk}^{(e)} + 0 + 0 \dots$$

$$\mu_k = \mathbb{E} T_k = \sum_{n=1}^{\infty} n \underbrace{f_{kk}^{(n)}}_{0 \leq \cdot \leq 1} \geq 0$$

mean recurrence time.

$\mu_k = \infty$ is possible

$$\Rightarrow \mu_k^{-1} := 0$$



positive persistent
 $\mu_k^{-1} > 0$

null persistent
 $\mu_k^{-1} = 0$

Ergodic Theorem (first version)

#X ∞

$$T = \begin{pmatrix} 0.602 & 0.398 \\ 0.166 & 0.695 \end{pmatrix}$$

$$T^5 = \begin{pmatrix} 0.305 & 0.695 \\ 0.296 & 0.710 \end{pmatrix} \quad \square \approx \square$$

Definition

Let $[(f_i)_{i=0}^{\infty}, p, T]$
be a Markov chain with $X = \{0, 1, \dots, K\}$

$P(k) > 0, k \in X,$

$$T = \begin{pmatrix} P_{00} & \dots & P_{0K} \\ \vdots & & \vdots \\ P_{K0} & \dots & P_{KK} \end{pmatrix}$$

$$T^n = \underbrace{T \circ \dots \circ T}_n = \left(P_{ke}^{(n)} \right)_{k,e=0}^K$$

1. Markov chain is ergodic $\stackrel{\text{df}}{\iff} \exists s_e > 0, e \in X$

(a) $s_0 + s_1 + \dots + s_K = 1$

(b) $\lim_{n \rightarrow \infty} P_{ke}^{(n)} = s_e \quad \forall k, e \in X$

2. A distribution $s = (s_0, \dots, s_K)$ is called stationary $\stackrel{\text{df}}{\iff}$

(a) $s_0 + s_1 + \dots + s_K = 1, s_e \in [0, 1],$

$$(b) \quad s \circ T = s \quad \text{i.e.} \quad \sum_{m=0}^K s_m p_{me} = s_e.$$

Proposition:

An ergodic Markov chain has a unique stationary distribution.

Proof:

$$T^n \rightarrow \begin{pmatrix} s_0 & s_1 & \dots & s_K \\ s_0 & s_1 & \dots & s_K \\ \vdots & \vdots & \ddots & \vdots \\ s_0 & s_1 & \dots & s_K \end{pmatrix} = A$$

because $p_{ke}^{(n)} \rightarrow s_e$.

$$\begin{aligned} A &= \lim_{n \rightarrow \infty} T^n = \left(\lim_{n \rightarrow \infty} T^n \right) \circ T \\ &= A \circ T \end{aligned}$$

$\Rightarrow (s_0, \dots, s_K) \circ T = (s_0, \dots, s_K)$
↑
is a stationary distribution.

$S = (s_0, \dots, s_K)$ is unique:

Let $b = (b_0, \dots, b_K)$ be a stationary distribution $\Rightarrow b \circ T = b$

$$b \circ T^2 = (b \circ T) \circ T = b \circ T = b$$

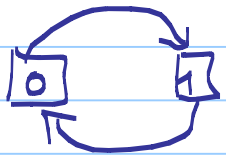
$$\lim_{h \rightarrow \infty} b \circ T^h = b \Rightarrow b \circ A = b$$

$$\begin{aligned} b_e &= b_0 s_e + b_1 s_e + \dots + b_K s_e \\ &= s_e (b_0 + \dots + b_K) \quad \forall e \in X \end{aligned}$$

$\underbrace{\quad}_{=1}$ b is a distribution

$b = s$ or s is unique. \square

Example MC has a unique stationary distribution \Rightarrow MC is ergodic



$$T = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \quad s = (s_0, s_1)$$

$$s_0 + s_1 = 1$$

$$s_i \in [0, 1]$$

$$s \circ T = s$$

$$\Leftrightarrow \begin{cases} s_1 = s_0 \\ s_0 = s_1 \end{cases}$$

so we have only one

$$\text{choice: } s = \left(\frac{1}{2}, \frac{1}{2} \right)$$

$$P_{ek}^{(n)} \xrightarrow[n \rightarrow \infty]{?} S_k :$$

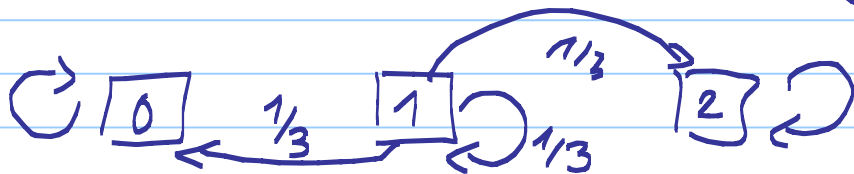
$$T^2 = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$

$$\left. \begin{array}{l} T^3 = T^2 \circ T = T \\ T^4 = T^2 \end{array} \right\} \Rightarrow \begin{array}{l} T^{2n} = T^2 \\ T^{2n+1} = T \end{array}$$

$$P_{ek}^{(n)} = 0, 1, 0, 1 \quad n=1, 2, \dots \quad \text{do not converge}$$

\Rightarrow MC is not ergodic!

Example: \exists more than one stationary distribution



$$T = \begin{pmatrix} 1 & 0 & 0 \\ \frac{1}{3} & \frac{2}{3} & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

$$s_0 T = s \Rightarrow \left. \begin{array}{l} s_0 + \frac{1}{3}s_1 = s_0 \\ \frac{1}{3}s_1 = s_1 \\ \frac{1}{3}s_1 + s_2 = s_2 \end{array} \right\} s_1 = 0$$

$$s_0 + \underbrace{s_1 + s_2}_{=0} = 1 \Rightarrow s_2 = 1 - s_0$$

$$s = (s_0, 0, 1 - s_0)$$

$$s_0 \in [0, 1]$$

\exists ∞ many stationary distributions

Theorem Ergodic Theorem (first version)

Let $[(f_i)_{i=0}^{\infty}, P, T]$ be a Markov chain with $X = \{0, \dots, K\}$

$$T^n = T \circ \dots \circ T = \left(P_{ke}^{(n)} \right)_{k,e=0}^K$$

Let there exist $n_0 \geq 1$ such that

$$\inf_{k,e} P_{ke}^{(n_0)} > 0. \text{ Then}$$

1. Markov chain is ergodic

2. The stationary distribution

$$s = (s_0, \dots, s_K) \text{ satisfies } s_e > 0 \quad \forall e \in X.$$

Proof Assume $n_0 = 1$

$$T^n = \begin{pmatrix} P_{00}^{(n)} & \dots & P_{0K}^{(n)} \\ \vdots & \vdots & \vdots \\ P_{K0}^{(n)} & \dots & P_{KK}^{(n)} \end{pmatrix}$$

e

$$m_e^{(n)} := \min_k P_{ke}^{(n)} \leq \max_k P_{ke}^{(n)} =: M_e^{(n)}$$

$$\boxed{m_e^{(n+1)} \geq m_e^{(n)}} :$$

$$\begin{aligned} m_e^{(n+1)} &= \min_k P_{ke}^{(n+1)} \\ &= \min_k \sum_{s=0}^K P_{ks} P_{se}^{(n)} \end{aligned}$$

Chapman -
Kolm.

$$\geq \min_s P_{se}^{(n)}$$

$$\geq \min_k \underbrace{\min_s P_{se}^{(n)}}_{= m_e^{(n)}} \underbrace{\sum_{s=0}^k P_{ks}}_{= 1}$$

$$= m_e^{(n)}$$

□