

6.262: Discrete Stochastic Processes 4/6/11

Lecture 16: Renewals and Countable-state Markov

Outline:

- Review major renewal theorems
- Age and duration at given t
- Countable-state Markov chains

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Sample-path time average (strong law for renewals)

$$\Pr\left\{\lim_{t \rightarrow \infty} \frac{N(t)}{t} = \frac{1}{\bar{X}}\right\} = 1.$$

Ensemble & time average (elementary renewal thm)

$$\lim_{t \rightarrow \infty} E\left[\frac{N(t)}{t}\right] = \frac{1}{\bar{X}}$$

Ensemble average (Blackwell's thm); $m(t) = E[N(t)]$

$$\lim_{t \rightarrow \infty} [m(t+\lambda) - m(t)] = \frac{\lambda}{\bar{X}} \quad \text{Arith. } X, \text{ span } \lambda$$

$$\lim_{t \rightarrow \infty} [m(t+\delta) - m(t)] = \frac{\delta}{\bar{X}} \quad \text{Non-Arith. } X, \text{ any } \delta > 0$$

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$$\lim_{t \rightarrow \infty} [m(t+\lambda) - m(t)] = \frac{\lambda}{\bar{X}} \quad \text{Arith. } X, \text{ span } \lambda$$

can be rewritten as

$$\lim_{n \rightarrow \infty} \Pr\{\text{renewal at } n\lambda\} = \frac{\lambda}{\bar{X}} \quad \text{Arith. } X, \text{ span } \lambda$$

If we model an arithmetic renewal process as a Markov chain starting in the renewal state 0, this essentially says $P_{00}^n \rightarrow \pi_0$.

$$\lim_{t \rightarrow \infty} [m(t+\delta) - m(t)] = \frac{\delta}{\bar{X}} \quad \text{Non-Arith. } X, \text{ any } \delta > 0$$

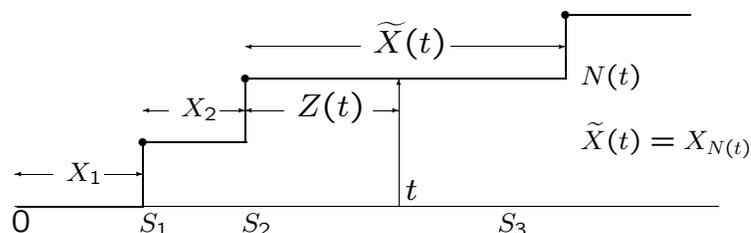
This is the best one could hope for. Note that

$$\lim_{\delta \rightarrow 0} \lim_{t \rightarrow \infty} \frac{m(t+\delta) - m(t)}{\delta} = \frac{1}{\bar{X}}$$

but the order of the limits can't be interchanged.

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Age and duration at given t



Assume an arithmetic renewal process of span 1.

For integer t , $Z(t) = i \geq 0$ and $\tilde{X}(t) = k > i$ iff there are successive arrivals at $t - i$ and $t - i + k$.

Let $q_j = \Pr\{\text{arrival at time } j\} = \sum_{n \geq 1} p_{S_n}(j)$ and let $q_0 = 1$ (nominal arrival at time 0). Then

$$P_{Z(t), \tilde{X}(t)}(i, k) = q_{t-i} p_X(k) \quad \text{for } 0 \leq i \leq t; k > i$$

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$$p_{Z(t), \tilde{X}(t)}(i, k) = q_{t-i} p_X(k) \quad \text{for } 0 \leq i \leq t; k > i$$

Note that

$$q_i = \Pr\{\text{arrival at } j\} = E[\text{arrival at } j] = m(i) - m(i-1),$$

so by Blackwell, $\lim_{j \rightarrow \infty} 1/\bar{X}$.

$$\lim_{t \rightarrow \infty} p_{Z(t), \tilde{X}(t)}(i, k) = \frac{p_X(k)}{\bar{X}} \quad \text{for } k > i \geq 0.$$

$$\lim_{t \rightarrow \infty} p_{Z(t)}(i) = \frac{\sum_{k=i+1}^{\infty} p_X(k)}{\bar{X}} = \frac{F_X^c(i)}{\bar{X}} \quad \text{for } i \geq 0.$$

$$\lim_{t \rightarrow \infty} p_{\tilde{X}(t)}(k) = \frac{\sum_{i=0}^{k-1} p_X(k)}{\bar{X}} = \frac{k p_X(k)}{\bar{X}} \quad \text{for } k \geq 1.$$

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Now look at asymptotic expected duration:

$$\lim_{t \rightarrow \infty} E[\tilde{X}(t)] = \sum_{k=1}^{\infty} k \cdot k p_X(k) / \bar{X} = E[X^2] / \bar{X}$$

This is the same as the sample-path average, but now we can look at the finite t case. More important, we get a different interpretation.

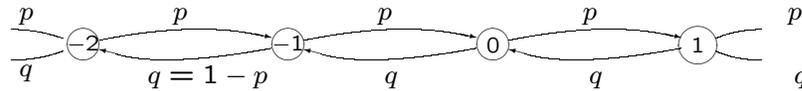
For a given $\tilde{X} = k$, there are k equiprobable choices for age; for each choice, the joint Z, \tilde{X} PMF is $p_X(k)/\bar{X}$. Thus large durations are enhanced relative to inter-renewals.

The expected age (after some work) is $E[X^2]/2\bar{X} - \frac{1}{2}$. This is at integer values of large t . The age increases linearly with slope 1 to the next integer value and then drops by 1.

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Countable -state Markov chains

The biggest change from finite-state Markov chains to countable-state chains is the concept of a recurrent class. Example:

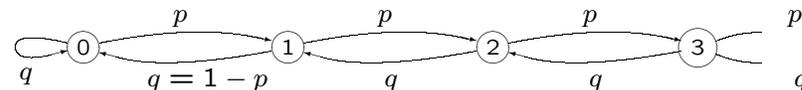


This Markov chain models a Bernoulli ± 1 process. The state at time n is $S_n = X_1 + X_2 + \dots + X_n$. The state S_n at time n is $j = 2k - n$ where k is the number of positive transitions in the n trials.

All states communicate and have a period $d = 2$; $\sigma_{S_n}^2 = n[1 - (p - q)^2]$. $P_{0,j}^n$ approaches 0 at least as $1/\sqrt{n}$ for every j .

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Another example (called a birth-death chain)



In this case, if $p > 1/2$, the state drifts to the right and P_{0j}^n approaches 0 for all j . If $p < 1/2$, it drifts to the left and keeps bumping state 0.

A truncated version of this was analyzed in the homework. With $p > 1/2$, the steady-state increases to the right, with $p < 1/2$, it increases to the left, and at $p = 1/2$ it is uniform.

As the truncation point increases, the 'steady-state' remains positive only for $p < 1/2$.

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We want to define recurrent to mean that, given $X_0 = i$, there is a future return to state i WP1. We will see that the birth-death chain above is recurrent if $p < 1/2$ and not recurrent if $p > 1/2$. The case $p = 1/2$ is strange and will be called null-recurrent.

We can use renewal theory to study recurrent chains, but first must understand first-passage-times.

Def: The first-passage-time probability, $f_{ij}(n)$, is

$$f_{ij}(n) = \Pr\{X_n=j, X_{n-1}\neq j, X_{n-2}\neq j, \dots, X_1\neq j | X_0=i\}.$$

It's the probability, given $X_0 = i$, that n is the first epoch at which $X_n = j$. Then

$$f_{ij}(n) = \sum_{k \neq j} P_{ik} f_{kj}(n-1); \quad n > 1; \quad f_{ij}(1) = P_{ij}.$$

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$$f_{ij}(n) = \sum_{k \neq j} P_{ik} f_{kj}(n-1); \quad n > 1; \quad f_{ij}(1) = P_{ij}.$$

Recall that Chapman-Kolmogorov says

$$P_{ij}^n = \sum_k P_{ik} P_{kj}^{n-1},$$

so the difference between $f_{ij}(n)$ and P_{ij}^n is only in cutting off the outputs from j (as before in finding expected first-passage-times).

Let $F_{ij}(n) = \sum_{m \leq n} f_{ij}(m)$ be the probability of reaching j by time n or before. If $\lim_{n \rightarrow \infty} F_{ij}(n) = 1$, there is a rv T_{ij} with distribution function F_{ij} that is the first-passage-time rv.

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We can also express $F_{ij}(n)$ as

$$F_{ij}(n) = P_{ij} + \sum_{k \neq j} P_{ik} F_{kj}(n-1); \quad n > 1; \quad F_{ij}(1) = P_{ij}$$

Since $F_{ij}(n)$ is nondecreasing in n , the limit $F_{ij}(\infty)$ must exist and satisfy

$$F_{ij}(\infty) = P_{ij} + \sum_{k \neq j} P_{ik} F_{kj}(\infty).$$

Unfortunately, choosing $F_{ij}(\infty) = 1$ for all i, j satisfies these equations. The correct solution turns out to be the smallest set of $F_{ij}(\infty)$ that satisfies these equations.

If $F_{jj}(\infty) = 1$, then an eventual return from state j occurs with probability 1 and the sequence of returns is the sequence of renewal epochs in a renewal process.

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If $F_{jj}(\infty) = 1$, then there is a rv T_{jj} with the distribution function $F_{jj}(n)$ and j is recurrent. The renewal process of returns to j then has inter-renewal intervals with the distribution function $F_{jj}(n)$.

From renewal theory, the following are equivalent:

- 1) state j is recurrent.
- 2) $\lim_{t \rightarrow \infty} N_{jj}(t) = \infty$ with probability 1.
- 3) $\lim_{t \rightarrow \infty} E [N_{jj}(t)] = \infty$.
- 4) $\lim_{t \rightarrow \infty} \sum_{1 \leq n \leq t} P_{jj}^n = \infty$.

None of these imply that $E [T_{jj}] < \infty$.

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Two states are in the same class if they communicate (same as for finite-state chains).

If states i and j are in the same class then either both are recurrent or both transient (not recurrent).

Pf: If j is recurrent, then $\sum_n P_{jj}^n = \infty$. Then

$$\sum_{n=1}^{\infty} P_{ii}^n \geq \sum_{k=1}^{\infty} P_{ij}^m P_{jj}^k P_{jk}^l = \infty$$

All states in a class are recurrent or all are transient.

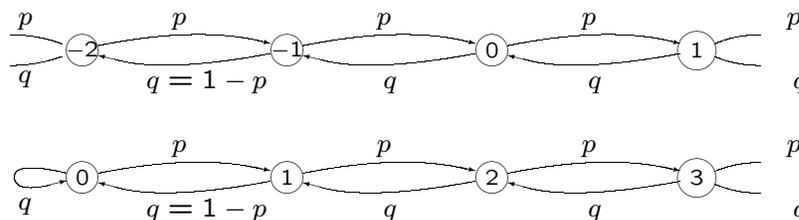
By the same kind of argument, if i, j are recurrent, then $F_{ij}(\infty) = 1$.

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If a state j is recurrent, then T_{jj} might or might not have a finite expectation.

Def: If $E[T_{jj}] < \infty$, j is positive recurrent. If T_{jj} is a rv and $E[T_{jj}] = \infty$, then j is null recurrent. Otherwise j is transient.

For $p = 1/2$, each state in each of the following is null recurrent.



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