

Stochastic Modeling and the Theory of Queues
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Module - 3
Lecture - 24
Strong Law for Renewal Processes - Proof

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Lec 24: Strong Law for RPs

$$\lim_{t \rightarrow \infty} \frac{N(t)}{t} = \frac{1}{\bar{X}} \text{ a.s.}$$

$\frac{N(t)}{S_{N(t)+1}} < \frac{N(t)}{t} \leq \frac{N(t)}{S_{N(t)}} \quad \forall t > 0.$

We will continue discussing the strong law for renewal processes. We will discuss the proof of the strong law. So, remember that the strong law says that limit t tending to infinity $N(t)$ over t is equal to 1 over \bar{X} almost surely. So, the intuitive reason that this works is as follows: You take some t ; at its time, there have been $N(t)$ arrivals. So, the epoch of the arrival that occurred just before t is just $S_{N(t)}$, and the epoch of the arrival that just occurred after the t is $S_{N(t)+1}$.

So, we are looking at $N(t)$ over t which can be sandwiched between $N(t)$ over $S_{N(t)+1}$, where this is less than or equal to, and $N(t)$ over $S_{N(t)}$, because numerators are all the same; denominators, when the denominator gets smaller, the ratio gets bigger, right? So, that is what I have used. So, this is true for all t . Now, if I send t to infinity, please notice that this $N(t)$ over $S_{N(t)+1}$ is like some n over S_{n+1} .

As t becomes large, this $N(t)$, the number of arrivals, you would expect that it goes, it increases 1 at a time, and it seems intuitively reasonable that $N(t)$ should go to infinity as t

goes to infinity. So, this ratio and this ratio, both behave like n over S_n for large n . And S_n over n by the strong law of large numbers goes to \bar{X} almost surely. So, you want to argue that n over S_n goes to $1/\bar{X}$ almost surely.

To formalise this argument, there are 2 ingredients that are required, as 2 lemmas. The first lemma is that N_t goes to infinity almost surely, as t goes to infinity. And the second is that, N over S_n goes to $1/\bar{X}$ almost surely. In fact, the second assertion is true for any continuous function. So, if you take any sequence of random variables X_n going to α , where α is some constant almost surely, then for any continuous function f , f of X_n will converge to f of α almost surely. So, if you have these 2 lemmas, will be able to prove this strong law. So, let us prove these 2 lemmas.

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$S_{N(t)} \leq S_{N(t)+1}$

$$\frac{N(t)}{S_{N(t)+1}} < \frac{N(t)}{t} \leq \frac{N(t)}{S_{N(t)}} \quad \forall t > 0.$$

Lemma For a RP, $\lim_{t \rightarrow \infty} N(t) = \infty$ a.s. (True even when $\bar{X} = \infty$)

Pf $P\{\omega \mid \lim_{t \rightarrow \infty} N(t, \omega) < \infty\} = P\left(\bigcup_{n \geq 1} \{\omega \mid \lim_{t \rightarrow \infty} N(t, \omega) < n\}\right)$

$$\leq \sum_{n \geq 1} P\left(\{\omega \mid \lim_{t \rightarrow \infty} N(t, \omega) < n\}\right)$$

want to show $\underbrace{0}_{!}$ for each $n \geq 1$.

For a renewal process, limit t tending to infinity N of t is equal to infinity almost surely. This is true even when \bar{X} is infinity. So, \bar{X} can be finite or infinite. As t tends to infinity, N of t goes to infinity almost surely. Proof of this is follows: So, you look at those ω for which limit t tending to infinity N of t ω is less than infinity. You want to show that this probability is 0.

See, for those ω for which limit t tending to infinity N of t ω is less than infinity, there must exist an N for which limit t tending to infinity N of t ω is less than n , because the limit is finite, we are saying. So, that limit must be something less than, some n ; so, there must exist such an n . There exists translates to a union, so, that just becomes probability

union n greater than or equal to 1, ω such that limit t tending to infinity $N(t, \omega) < n$.

And this of course, by the union bound is less than or equal to sum over n greater than or equal to 1, probability of ω such that limit t tending to infinity $N(t, \omega) < n$. So, now, I want to show that each of these things is 0. This is equal to 0 for each n greater than or equal to 1. So, if each of these terms is 0, then the sum is 0. Then I will be done. So, you fix an n and consider this probability.

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For $n \geq 1$, and consider $P(\{\omega \mid \lim_{t \rightarrow \infty} N(t, \omega) < n\})$

Consider a seq $\{t_k\}$ such that $t_k \uparrow \infty$. $t_1 \quad t_2 \quad t_k \quad t_{k+1}$

$\Rightarrow P(\bigcap_{k \geq 1} \{\omega \mid N(t_k, \omega) < n\})$

Note that $\{\omega \mid N(t_k, \omega) < n\} \supseteq \{\omega \mid N(t_{k+1}, \omega) < n\}$

Thus

$$P(\{\omega \mid \lim_{t \rightarrow \infty} N(t, \omega) < n\}) = \lim_{k \rightarrow \infty} P(\{\omega \mid N(t_k, \omega) < n\})$$

(Continuity of P)

So, you fix n and consider this probability out here; ω such that limit t tending to infinity $N(t, \omega) < n$. Now, this is a limit that is inside this probability. We would like to bring this out. And bringing the limit out, you carry out using a continuity of probability argument. So, the way you do this is, you consider a sequence t_k such that t_k is increasing and it increases to infinity.

You consider any sequence t_k such that $0 < t_1 < t_2 < \dots$, and this t_k , the sequence t_k goes off to infinity, increases to infinity. So, I can write this probability as; so, this guy, let me just write it like this. So, this will just become intersection over k greater than or equal to 1 ω ; so, this will be equal to this; $N(t_k, \omega) < n$. So, let me tell you what I have done.

So, I considered this sequence t_1, t_2, \dots , some t_k , which is going off to infinity. Now, I am telling; so, this is the event I want to consider, is the, those ω s for which limit $N(t, \omega)$

is less than n , limit t tending to infinity $N(t, \omega)$ is less than n , which means that for every t , $N(t, \omega)$ must be less than n . See, if in the limit, $N(t, \omega)$ is less than n , then $N(t, \omega)$ should be less than n ; for if not, your limit will be bigger than n ; that is, it cannot hold.

Conversely, if for every t , $N(t, \omega)$ is less than n , for every t ; and remember, this t is going to infinity; then this limit will be less than n . So, you can easily show that this event is same as this intersection. You can show the equality between this event and what is inside here. It is this countable intersection. Now, it is clear that these kind of sets are nested decreasing sets.

So, what I mean is that; let me write it this way; note that ω for which $N(t, \omega) < n$ is, it contains the ω for which $N(t, \omega) < n$. Because, if your ω is such that $N(t, \omega) < n$, at t , you have less than n arrivals; then the number of arrivals at t must be less than n . So, this event implies this event, which means that these sets are nested decreasing.

And you have a countable intersection over nested decreasing sets, and by continuity of probability, this is equal to the limit of the k th set, probability of the k th set. So, you can say, thus probability of; you go back to what you want; limit t tending to infinity $N(t, \omega) < n$. So, you are fixing an n , like here; is equal to this guy, which is equal to limit k tending to infinity probability of those ω for which $N(t, \omega) < n$; this is by continuity of probability. So, maybe I should recall that for you.

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$$\lim_{t \rightarrow \infty} P\left(\bigcap_{k=1}^{\infty} \{\omega \mid N(t_k, \omega) < n\}\right)$$

Note that $\{\omega \mid N(t_k, \omega) < n\} \supseteq \{\omega \mid N(t_{k+1}, \omega) < n\}$

Thus

$$P\left(\{\omega \mid \lim_{t \rightarrow \infty} N(t, \omega) < n\}\right) = \lim_{k \rightarrow \infty} P\left(\{\omega \mid N(t_k, \omega) < n\}\right)$$

(Continuity of P)

Recall: If $A_k \supseteq A_{k+1}$, then

$$P\left(\bigcap_{k=1}^{\infty} A_k\right) = \lim_{k \rightarrow \infty} P(A_k)$$

$$= \lim_{k \rightarrow \infty} P\left(\{\omega \mid S_n(\omega) > \frac{1}{k}\}\right) = 0$$

2/2

Recall: If some events A_k are nested decreasing, then the probability of intersection $\bigcap_k A_k$ is equal to limit; did I write it correctly? This is correct, right? This is what I am using. This you must know; this you would have studied in probability course. This is the fundamental property of probability measures. So, you use that. So, this is equal to this. Now, we are in business.

So, if you look at this; so, this recall, I will put in a box, so that it does not break your proof. So, I will continue here. So, this is equal to limit k tending to infinity probability of ω such that $S_n(\omega) > t_k$. Why is that? $N(t_k) \leq n$ is same as $S_n > t_k$. This equivalence we proved, remember? What we proved was that $N(t) \geq n$ is same as $S_n \leq t$.

We are just taking the complement of it. Now, this is equal to 0. Why? Because t_k is going to infinity. Is the complimentary CDF. So, this t_k is going to infinity and S_n is some random variable. Probability $S_n > t_k$, as this t_k goes to infinity, has to be 0, because this is a property of the complimentary CDF of any random variable. See, this S_n is a legitimate random variable, so, it is finite with probability 1. So, its CCDF has to go to 0.

As k goes to infinity; if you want, you can write one more step; you can write limit t tending to infinity $S_n > t$; and that will be 0. So, what have we shown? We have shown that this guy is 0, for each n . This is true for all n greater than or equal to 1. So, this guy is equal to 0 for all n greater than or equal to 1; therefore, these terms, each of these terms is 0. This is what I wanted to prove. So, I am done with the proof.

So, it requires a continuity of probability argument, which is missing in your book. You can just fill this in. The taking the limit out of the probability requires this nested construction; without that, it is not rigorously correct; I thought I will just point it out. With me? So, that is your first lemma that says that $N(t)$ increases to infinity as t goes to infinity; of course, arrivals come one at a time, so, $N(t)$ increases through all the positive integers, does not make any jumps between 2 integers, because no 2 arrivals can come at the same time.

So, it goes through all the positive integers and it goes off to infinity as t goes to infinity. So, essentially, if you look at this ratio, $N(t)/t$, it is sandwiched between $N(t)/S_{N(t)}$ and $N(t)/S_{N(t)+1}$. And therefore, as t goes to infinity, this ratio goes like $1/S_1$; $2/S_2$

2; 3 over S 3; and so on. And, and this guy goes like 1 over S 2; 2 over S 3; and so on. So, there is just one more ingredient that is missing, which is sort of a continuity property.

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Lemma Let $\{Z_n\}_{n \geq 1}$ be a seq. of RVs such that $Z_n \rightarrow \alpha$ a.s. for some $\alpha \in \mathbb{R}$. Let $f: \mathbb{R} \rightarrow \mathbb{R}$ be continuous at α . Then $f(Z_n) \rightarrow f(\alpha)$ a.s.

Pf Given $P(\{\omega \mid Z_n(\omega) \rightarrow \alpha\}) = 1$

For any $\omega \in \{\omega \mid Z_n(\omega) \rightarrow \alpha\}$ we have $f(Z_n(\omega)) \rightarrow f(\alpha)$ (since f is cont.)

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If Z_n is a sequence of random variables such that Z_n converges to α almost surely for some α in \mathbb{R} . Let me say, let Z_n be a sequence; then say, let f be continuous at α ; then f of Z_n converges to f of α almost surely. So, if you have a sequence of random variables converging to a number almost surely, then any continuous transformation of the sequence of random variables will converge to the corresponding f of α almost surely.

Key issue is here, f should be continuous. So, the key issue here is; so, the proof is fairly easy. So, you are given this, right? Given that the probability of those ω s for which X_n of ω converges to α equal to 1. So, you know that X_n converges to α almost surely. So, the probability of those ω s where X_n of ω converges to α is equal to 1. This is by definition of almost sure convergence.

Now, for all ω s in this set, $f(X_n$ of $\omega)$ will converge to f of α . Why? Continuity. So, for any ω in this; **"Professor - student conversation starts"** Yes. This should be a Z_n , right? That is all; I just put Z_n because I did not want to, X_n 's are our renewal interrenewal times, right? There is no big deal otherwise. **"Professor - student conversation ends"** For any ω prime in this set, we have f of Z_n of ω prime converges to f of α , f is continuous at α . So, what does this mean?

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Pf Given $\mathbb{P}(\{\omega \mid Z_n(\omega) \rightarrow \alpha\}) = 1$

For any $\omega' \in \{\omega \mid Z_n(\omega) \rightarrow \alpha\}$ we have $f(Z_n(\omega')) \rightarrow f(\alpha)$
(since f is cont.)

$$\{\omega \mid f(Z_n(\omega)) \rightarrow f(\alpha)\} \supseteq \{\omega \mid Z_n(\omega) \rightarrow \alpha\}.$$

$$\mathbb{P}(\downarrow) \geq \mathbb{P}(\downarrow) = 1 \quad \square$$

For strong law, we take $f(x) = \frac{1}{x}$ which is continuous for any $x > 0$.



So, if you look at those omegas for which f of Z_n of ω converge to f of α . If you look at those omegas for which f of Z_n of ω converge to f of α , this will contain those omegas for which Z_n of ω converges to α . I have just proven this, right? For every ω contained in this set, I have $f Z_n$ of ω will converge to f of α . So, every ω here is a member here.

Every ω for which Z_n of ω converges to α , for those omegas, f of Z_n of ω will converge to f of α . So, this is a bigger set than this. Have you written it in the correct direction? And so, we are saying that B is contained in A ; so, probability of B has to be greater than or equal to probability of A . So, what can I say? So, I can write, so, probability of all that is greater than or equal to probability of all that.

But what is the probability of all that? 1, because Z_n converges to α on a set of probability 1. So, probability of f of Z_n of ω converging to f of α is greater than or equal to 1; but probability cannot be greater than 1, so, it has to be equal to 1. So, I will explain the containment once more. You take, see, this is the set of omegas for which Z_n of ω converges to α . This is a set of probability 1; we are given that, right?

This is some subset of the sample space; it may not be all of the sample space, but it has all the probability. You fix any ω' you want in this. Then Z_n of ω' is a sequence of real numbers converging to α . Now, f is continuous at α ; therefore, by definition of continuous functions, you have, for that sequence f of Z_n of ω' , you will have convergence to f of α .

So, for every omega prime for which this convergence holds, this convergence necessary holds. So, if you look at this set, any omega here is a member of this set. Of course, this set could be bigger, it could have a few other omegas; so, we will get probability is greater than or equal to 1. Probability cannot be greater than 1, so, it is equal to 1; simple as that. So, in our case, in our setting, for strong law, we take f of X is equal to 1 by x, which is continuous for any X positive. So, now we can prove the strong law, but we can prove it for X bar being finite.

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NPTEL


Proof of Strong Law (Take $\bar{X} < \infty$)

$$\frac{N(t)}{S_{N(t)+1}} < \frac{N(t)}{t} \leq \frac{N(t)}{S_{N(t)}}$$

$$\lim_{t \rightarrow \infty} \frac{N(t)}{S_{N(t)+1}} \leq \lim_{t \rightarrow \infty} \frac{N(t)}{t} \leq \lim_{t \rightarrow \infty} \frac{N(t)}{S_{N(t)}}$$

For finite \bar{X} we already have this, right? $\frac{N(t)}{t}$ is sandwiched between $\frac{N(t)}{S_{N(t)}}$ and $\frac{N(t)}{S_{N(t)+1}}$. So, I can say that limit t tending to infinity, $\frac{N(t)}{t}$ less than or equal to limit t tending to infinity $\frac{N(t)}{S_{N(t)}}$. And here again write limit t tending to infinity $\frac{N(t)}{S_{N(t)+1}}$. So, this less than will become less than or equal to when I take limit. It is like $\frac{1}{N}$ is always greater than 0, but limit $\frac{1}{N}$ is 0. So, this strict inequality will become a non-strict inequality in the limit. I hope you understand that. So, this much you will agree, right? This is just sandwiching property.

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$$\lim_{t \rightarrow \infty} \frac{N(t)}{S_{N(t)+1}} \leq \lim_{t \rightarrow \infty} \frac{N(t)}{t} \leq \lim_{t \rightarrow \infty} \frac{N(t)}{S_{N(t)}}$$

|| a.s.

$$\lim_{n \rightarrow \infty} \frac{n}{S_n} \leq \lim_{t \rightarrow \infty} \frac{N(t)}{t} \leq \lim_{n \rightarrow \infty} \frac{n}{S_n} = \frac{1}{\bar{X}}$$

||

$$\frac{1}{\bar{X}} \stackrel{a.s.}{=} \lim_{n \rightarrow \infty} \frac{n+1}{S_{n+1}} \cdot \frac{n}{n+1} \Rightarrow \lim_{t \rightarrow \infty} \frac{N(t)}{t} = \frac{1}{\bar{X}} \text{ a.s. (for } \bar{X} < \infty)$$

If $\bar{X} = \infty$, proof uses a truncation argument. 4/4



Now, as limit t tends to infinity, so, this is the same as, this guy is the same as limit n tends to infinity n over S_n by the lemma that we proved. $N(t)$ increases 1 at a time and it increases to infinity almost surely. So, this equality is true almost surely. And likewise, this will be limit N tending to infinity n over S_{n+1} . Now, by the continuous mapping theorem, this guy is equal to, what is this equal to? 1 over \bar{X} . Why?

It is a reciprocal mapping is continuous, and \bar{X} I have assumed is, \bar{X} is finite I have assumed. If \bar{X} is not finite, then I cannot invoke continuity. So, if \bar{X} is finite, this limit will be equal to, this guy will be equal to almost surely equal to 1 over \bar{X} . This n over S_{n+1} also, you can write it as $n+1$ over S_{n+1} times n over $n+1$ and manipulate. So, this guy, you can write it as limit n tending to infinity $n+1$ over S_{n+1} times n over $n+1$.

I think I made a mistake; Correct no? So, this will of course go to; same logic; it will go to 1 over \bar{X} ; this will go to 1 . So, this will also be equal to 1 over \bar{X} almost surely. So, what have I shown? So, I have shown this, limit t tending to infinity $N(t)$ over t is equal to 1 over \bar{X} almost surely. I have shown this for \bar{X} finite; but the result I also claimed is true for \bar{X} being infinite. So, that I will not go into in great detail.

I will just mention that, if \bar{X} is infinity, proof uses a truncation argument. Have you seen truncation arguments before? So, you basically truncate the random variable. So, you are saying \bar{X} is infinity, meaning that these X 's are random variables which are finite with

probability 1, but the expected value is infinite. There exist such random variables; I suppose you know.

If you take a one-sided Cauchy distribution for example, it is finite probability 1, but expected value is infinite. In those cases, you define a new renewal process with interarrival times X_i tilde equal to $\max(X_i, B)$, for some B , very big B . So, you chop off the random variable at B , if it becomes too large. And because you are chopping off, this random variable will now have finite X bar; and for that, the strong law will hold, as we have proved it; and then, you have to take B to infinity.

You have to do this in the right order; you have to be a little careful, but it can be done. There are many theorems in probability which uses truncation argument. Even if you want to prove weak law for unbounded variance case; so, for bounded variance case, you can prove weak law using Chebyshev's inequality; for unbounded variance case, you use a truncation argument for example. So, this is a standard trick in proving many results. But this I will not spend class time on. I think there is an exercise in your book to actually guide you through this steps of strong law.