

6.041SC Probabilistic Systems Analysis and Applied Probability, Fall 2013 Transcript – Recitation: Convergence in Probability Example

In this problem, we're given a random variable X which has a uniform distribution in the interval negative 1 to 1. In other words, if we were to draw out the PDF of X , we see that in the interval negative 1 to 1, it has value $1/2$. Now we're given a sequence random variables X_1, X_2 , and so on, where each X_i has the same distribution as X and different X_i 's are independent.

For part a, we would like to know if the sequence X_i converges to some number-- let's call it c -- in probability as i goes to infinity-- whether this is true. Let's first recall the definition of convergence in probability. If this does happen, then by definition, we'll have that for every ϵ greater than 0, the probability X_i minus c greater equal to ϵ , this quantity will go to 0 in the limit of i going to infinity. In other words, with very high probability, we will find X_i to be very concentrated around the number c if this were to be the PDF of X_i .

Now, can this be true? Well, we know that each X_i is simply a uniform distribution over negative 1 to 1. It doesn't really change as we increase i . So intuitively, the concentration around any number c is not going to happen. So we should not expect a convergence in probability in this sense.

For part b, we would like to know whether the sequence Y_i , defined as X_i divided by i , converges to anything in probability. Well, by just looking at the shape of Y_i , we know that since the absolute value of X_i is less than 1, then we expect the absolute value of Y_i is less than $1/i$. So eventually, Y_i gets very close to 0 as i goes to infinity. So it's safe to bet that maybe Y_i will converge to 0 in probability.

Let's see if this is indeed the case. The probability of Y_i minus 0 greater equal to ϵ is equal to the probability of Y_i absolute value greater equal to ϵ . Now, previously we know that the absolute value of Y_i is at most $1/i$ by the definition of Y_i . And hence the probability right here is upper bounded by the probability of $1/i$ greater equal to ϵ .

Notice in this expression, there is nothing random. i is simply a number. Hence this is either 1 if i is less equal to $1/\epsilon$, or 0 if i is greater than $1/\epsilon$.

Now, this tells us, as long as i is great enough-- it's big enough compared to ϵ -- we know that this quantity here is [INAUDIBLE] 0. And that tells us in the limit of i goes to infinity probability of Y_i deviating from 0 by more than ϵ goes to 0. And that shows that indeed, Y_i converges to 0 in probability because the expression right here, this limit, holds for all ϵ .

Now, in the last part of the problem, we are looking at a sequence Z_i defined by X_i raised to the i -th power. Again, since we know X_i is some number between negative 1 and 1, this number raised to the i -th power is likely to be very small. And likely to be small in the sense that it will have absolute value close to 0. So a safe guess will be the sequence Z_i converges to 0 as well as i goes to infinity.

How do we prove this formally? We'll start again with a probability that Z_i stays away from 0 by more than ϵ and see how that evolves. And this is equal to the probability that X_i raised to the i -th power greater equal to ϵ . Or again, we can write this by taking out the absolute value that X_i is less equal to negative ϵ raised to the 1 over i -th power or X_i greater equal to ϵ 1 over i -th power.

So here, we'll divide into two cases, depending on the value of ϵ . In the first case, ϵ is greater than 1. Well, if that's the case, then we know ϵ raised to some positive power is still greater than 1. But again, X_i cannot have any positive density be on the interval negative 1 or 1. And hence we know the probability above, which is X_i less than some number smaller than negative 1 or greater than some number bigger than 1 is 0. So that case is handled.

Now let's look at a case where ϵ is less than 1, greater than 0. So in this case, ϵ to the $1/i$ will be less than 1. And it's not that difficult to check that since X_i has uniform density between negative 1 and 1 of magnitude $1/2$, then the probability here was simply 2 times $1/2$ times the distance between ϵ to the 1 over i -th power and 1.

So in order to prove this quantity converge to 0, we simply have to justify why does ϵ to the $1/i$ converge to 1 as i goes to infinity. For that, we'll recall the properties of exponential functions. In particular, if a is a positive number and x is its exponent, if we were to take the limit as x goes to 0 and look at the value of a to the power of x , we see that this goes to 1.

So in this case, we'll let a be equal to ϵ and x be equal to $1/i$. As we can see that as i goes to infinity, the value of x , which is $1/i$, does go to 0. And therefore, in the limit i going to infinity, the value of ϵ to the 1 over i -th power goes to 1.

And that shows if we plug this limit into the expression right here that indeed, the term right here goes to 0 as i goes to infinity. And all in all, this implies the probability of Z_i minus 0 absolute value greater equal to ϵ in the limit of i going to infinity converges to 0 for all positive ϵ . And that completes our proof that indeed, Z_i converges to 0 in probability.

MIT OpenCourseWare
<http://ocw.mit.edu>

6.041SC Probabilistic Systems Analysis and Applied Probability
Fall 2013

For information about citing these materials or our Terms of Use, visit: <http://ocw.mit.edu/terms>.