# On Convergence of Random Series 

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## Outline

1. Review: Numerical series
2. Random Series
3. Two Theorems By Kolmogorov
4. Examples
5. Strong Law of Large Numbers

## Numerical Series

## Series

$\rightarrow$ Sequence $a_{1}, a_{2}, \ldots$ real numbers.

- Series, the "infinite sum" denoted by " $\sum_{m=1}^{\infty} a_{m}$ ", with terms $a_{1}, a_{2}, \ldots$.


## Convergence

Infinite sum $=$ Limits of finite sums.

- Sequence of partial sums $S_{n}=\sum_{m=1}^{n} a_{m}$.
- Series converges $\Leftrightarrow \lim _{n \rightarrow \infty} S_{n}$ exists as a real number, the sum of the sequence.
- Diverges otherwise.


## Examples

- Geometric, $a_{n}=q^{n-1}$.
- Harmonic, $a_{n}=\frac{1}{n}$.
- Alternating harmonic, $a_{n}=\frac{(-1)^{n+1}}{n}$.
- Other? Taylor series, and more fancy stuff.

No closed form in general. Sometimes can get estimates, or, more generally can determine if converges or not.

## Convergence tests

Tools to determine convergence. Many... Here are a few.
Theorem 1 (Comparison)
Suppose $\sum_{n=1}^{\infty} b_{n}$ is converges and $\left|a_{n}\right|<b_{n}$. Then $\sum_{n=1}^{\infty} a_{n}$ converges.
Theorem 2 (Condensation - substitution)
Suppose $a_{n} \geq a_{n+1} \geq \ldots 0$. Then $\sum_{n=1}^{\infty} a_{n}$ converges $\Leftrightarrow \sum_{n=1}^{\infty} 2^{n} a_{2^{n}}$ convegres.

## Examples

- $a_{n}=q^{n-1}$. Then $S_{n}=\frac{1-q^{n}}{1-q}$ and therefore $\sum_{n=1}^{\infty} a_{n}$ converges $\Leftrightarrow|q|<1$.
- $a_{n}=n^{-p}$. Then $2^{n} a_{2^{n}}=2^{n} 2^{-p n}=\left(2^{1-p}\right)^{n}=q^{n}$. Therefore converges iff $p>1$.
- Try: $a_{n}=\frac{1}{n \ln (1+n)}$.

Theorem 3 (Dirichlet's test - summation by parts)
Suppose $b_{n} \searrow 0$ and the sequence $\left(s_{n}: n=1,2, \ldots\right)$ is bounded. Then $\sum_{n=1}^{\infty} b_{n}\left(s_{n+1}-s_{n}\right)$ converges.

## Examples

- Let $a_{n}=\frac{(-1)^{n+1}}{n}$. Apply with $b_{n}=\frac{1}{n}$ and $\left(s_{n}: n=1,2, \ldots\right)=(0,1,0,1, \ldots)$.
- Try: $a_{n}=\frac{\sin n}{\ln (1+n)}$.


## Random Series

## What's the deal?

- We sample $a_{n}$ randomly.
- Each realization of the sampling yields a (possibly) different series.


## Example

- Toss a fair coin repeatedly.
- Set

$$
H_{n}= \begin{cases}1 & \text { n'th toss is } H \\ 0 & \text { n'th toss is } T\end{cases}
$$

- Set $a_{n}=2^{-n} H_{n}$
- The series is

$$
\sum_{n=1}^{\infty} a_{n}=\frac{H_{1}}{2^{1}}+\frac{H_{2}}{2^{2}}+\frac{H_{3}}{2^{3}}+\ldots
$$

essentially randomly picking some of the terms of the geometric sequence $\left(q=\frac{1}{2}\right)$.

- $\Rightarrow$ Converges, due to Theorem 1.
- But the sum can be anywhere between $0\left(0=H_{1}=H_{2}=\ldots\right)$ and 1 $\left(1=H_{1}=H_{2}=\ldots\right)$.


## The series as a Random Variable

Discussion

- Our random series $\sum_{n=1}^{\infty} \frac{H_{n}}{2^{n}}$ always converges.
- Estimating its sum? Nothing beyond the trivial bounds 0 and 1.
- Enter probability.


## Probabilistic viewpoint

- Switch to a "statistical" perspective.
- Though we don't know what the outcome of the first $n$ tosses will be, we do know all $2^{n}$ outcomes have the same probability of appearing.
- So, at least theoretically, we can find the probability that the sum lies some interval.
- What's the probability that the sum will be in the interval $[0,1 / 2)$ ? In the interval $[0,1 / 4]$ ? Equal to $\frac{3}{4}$ ? Between two dyadic numbers?


## Bottom line

1. Consider the sum as a function of the "random" realization - an object known as a random variable, and
2. Look at the probability this random variable lies an any interval - the distribution of the RV.

## Simulations

Simulation $\sum_{n-1}^{\infty} \frac{H_{n}}{2^{n}}$, sampled $10^{6}$ times


Empirical distribution function


## Discussion

- The histogram is a bit noisy, so I added a graph of the corresponding distribution function.
- What is your conclusion?
- Indeed, the distribution of the series $\sum_{n=1}^{\infty} \frac{H_{n}}{2^{n}}$ is uniform on $[0,1]$.
- This gives a bridge between discrete RVs and continuous RVs. Every RV can be generated from an infinite sequence of fair coin tosses.


## More simulations

An interesting example

- Let's change from $\sum_{n=1}^{\infty} \frac{H_{n}}{2^{n}}$ to $\sum_{n=1}^{\infty} \frac{2 H_{n}}{3^{n}}$.
- The 2 in numerator to make sure we cover the same range of $[0,1]$ ( $\sum_{n=1}^{\infty} \frac{1}{3}=\frac{1}{2}$ ).
- What do you think?


## More simulations

An interesting example

- Let's change from $\sum_{n=1}^{\infty} \frac{H_{n}}{2^{n}}$ to $\sum_{n=1}^{\infty} \frac{2 H_{n}}{3^{n}}$.

Simulation $\sum_{n=1}^{\infty} \frac{2 \mathrm{H}_{n}}{3^{n}}$, sampled $10^{6}$ times


Empirical distribution function


## Discussion

- Here the histogram is far from smooth, and again, the picture is much clearer if we look at the empirical CDF.
- The CDF of this RV is the Cantor function.
- This is an example of a RV which is continuous, but has no density.


## Other random series?

## Recall

- The series $\sum_{n=1}^{\infty} \frac{1}{n}$ diverges.
- Yet, the alternating signs series $\sum_{n=1}^{\infty} \frac{(-1)^{n+1}}{n}$ converges.

A random "version"

- Same fair coin, same $H_{n}$
- Form the series

$$
\sum_{n=1}^{\infty} \frac{(-1)^{H_{n}}}{n}
$$

- Much tougher than our previous case.
- Diverges/converges for some "freak" realizations, but what happens in the "bulk"?


## Some framework

## Independent RVs

The RVs $X_{1}, X_{2}, \ldots$ are independent if information on any of them does not alter the distribution of the other.

## Examples

- The RVs $H_{1}, H_{2}, \ldots$ from our examples, are independent.
- $f_{1}\left(X_{1}\right), f_{2}\left(X_{2}\right), \ldots$ where $X_{1}, X_{2}, \ldots$ are independent and $f_{1}, f_{2}, \ldots$ are functions.
- Partial sums $X_{1}=H_{1}, X_{2}=H_{1}+H_{2}, \ldots$ are not independent. If $X_{2}=2$, then necessarily $X_{1}=1$, although $P\left(X_{1}=1\right)=\frac{1}{2}$.


## Events

An event is a collection of realizations.

1. All but finitely many tosses are $H$.
2. Any finite pattern appears infinitely many times.
3. The proportion of $H$ in first $n$ tosses converges to the constant $c$.
4. The random series converges.

## Almost sure

- An event holds almost surely if its probability is 1 ("the bulk").
- It does not necessarily mean the event contains all realizations!


## Theorem 4 (Kolmogorov's 0-1)

Let $\mathbf{X}=\left(X_{1}, X_{2}, \ldots\right)$ be independent. Any event stated in terms of the sequence $\left(X_{1}, X_{2}, \ldots\right)$, not affected by the value of any of the $X_{n}$ 's, has probability 0 or 1 .

## Example

- Sounds weird?
- All examples from the last slide are of this type!
- In particular, if $\mathbf{X}=\left(X_{1}, X_{2}, \ldots\right)$ are independent, then the series

$$
\sum_{n=1}^{\infty} \frac{x_{n}}{b_{n}}
$$

either converges a.s. or diverges a.s. Whichever alternative holds? We have a theorem for that too.

How to prove? Show that any such event is independent of itself.

## 3 Series

Theorem 5 (Kolmogorov's Three Series Theorem)
Let $\mathbf{Y}=\left(Y_{1}, Y_{2}, \ldots\right)$ be independent. Let

$$
Z_{n}= \begin{cases}Y_{n} & \left|Y_{n}\right| \leq 1 \\ 0 & \text { otherwise }\end{cases}
$$

Then the series $\sum_{n=1}^{\infty} Y_{n}$ converges a.s. if and only if all of the following conditions hold:

1. $\sum_{n=1}^{\infty} P\left(\left|Y_{n}\right|>1\right)<\infty$ (large finitely otfen)
2. $\sum_{n=1}^{\infty} E\left[Z_{n}\right]<\infty$ (expectation of partial sums)
3. $\sum_{n=1}^{\infty} E\left[\left(Z_{n}-E\left[Z_{n}\right]\right)^{2}\right]<\infty$ (variance of partial sums)

Application
Consider the series $\sum_{n=1}^{\infty} \underbrace{\frac{(-1)^{H_{n}}}{n}}_{=Y_{n}}$.

- $\left|Y_{n}\right| \leq 1 \Rightarrow Z_{n}=Y_{n}$ and $1 \checkmark$
- $E\left[Z_{n}\right]=0, \Rightarrow 2 \checkmark$
- $E\left[Z_{n}^{2}\right]=\frac{1}{n^{2}} \Rightarrow 3 \checkmark$

Conclusion: converges a.s.
Other proofs? This Math Stack Exchange post.

## Generalization

Random $p$-harmonic

- Reminder: $\sum_{n=1}^{\infty} \frac{(-1)^{n}}{n^{p}}$ converges for all $p>0$ (Theorem 3).
- What about

$$
\begin{equation*}
\sum_{n=1}^{\infty} \frac{(-1)^{H_{n}}}{n^{p}} \tag{*}
\end{equation*}
$$

- Conditions 1,2 in Theorem 5 trivially hold, with $Z_{n}=Y_{n}$.
- Check condition 3: $E\left[Z_{n}^{2}\right]=\frac{1}{n^{2 p}}$.

Corollary 1
$(*)$ converges a.s. if $p>\frac{1}{2},(*)$ diverges a.s. if $p \leq \frac{1}{2}$.

## Simulations

Let's look at simulations for the random harmonic series.


- Much nicer one on poster.
- More on the distribution? Read Byron Schmuland, Random Harmonic Series

What about other values of $p$ ?

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## Random L-functions ${ }^{1}$, $1 / 2$

## Construction

- Let $H_{1}=1$.
- For prime $p$, define $H_{p}$ as before
- Extend to all natural numbers through the formula $H_{n m}=H_{n}+H_{m}$ (you can do this $\bmod 2$ ).
- Example: $H_{p^{n}}=n H_{p}, H_{6}=H_{2}+H_{3}$, etc. Define

$$
\begin{equation*}
L(s)=\sum_{n=1}^{\infty} \frac{(-1)^{H_{n}}}{n^{s}} \tag{**}
\end{equation*}
$$

Almost the same as $(*)$, but here $H_{n}$ are not independent! $H_{2}$ determines $H_{2^{n}}$, etc.

Corollary 2

$$
(* *) \text { converges a.s. if } s>\frac{1}{2} \text {. }
$$

[^0]Random L-functions, $2 / 2$

## Proof of Corollary 2

Key idea: bring this to the form of Theorem 5.

- By prime factorization,

$$
\sum_{n=1}^{\infty} \frac{1}{n^{s}}=\prod_{p \text { prime }}\left(1+\frac{1}{p^{s}}+\frac{1}{p^{2 s}}+\ldots\right)=\prod_{p \text { prime }}\left(1-\frac{1}{p^{s}}\right)^{-1}
$$

Side note: can you see why $\sum_{p}$ prime $\frac{1}{p^{s}}$ converges if and only if $s>1$ ?

Random L-functions, $2 / 2$

## Proof of Corollary 2

Key idea: bring this to the form of Theorem 5.

- Because $n \rightarrow(-1)^{H_{n}}$ is multiplicative,

$$
\begin{aligned}
\sum_{n=1}^{\infty} \frac{(-1)^{H_{n}}}{n^{s}} & =\prod_{p \text { prime }}\left(1+\frac{(-1)^{H_{p}}}{p^{s}}+\frac{(-1)^{2 H_{p}}}{p^{2 s}}+\ldots\right) \\
& =\prod_{p \text { prime }}\left(1-\frac{(-1)^{H_{p}}}{p^{s}}\right)^{-1} \\
& =\exp \left(-\sum_{p \text { prime }} \ln \left(1-\frac{(-1)^{H_{p}}}{p^{s}}\right)\right)
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Random L-functions, $2 / 2$

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\end{aligned}
$$

- Use Taylor expansion $-\ln (1-x)=x+x^{2} / 2+\ldots$, to recover

$$
\sum_{p \text { prime }} \underbrace{\frac{(-1)^{H_{p}}}{p^{s}}}_{(I)}+\underbrace{\frac{1}{2 p^{2 s}}+\ldots}_{(I I)}
$$

Random L-functions, $2 / 2$

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$$

- So... when $s>\frac{1}{2}$
- (II) converges (we mentioned earlier this slide).
- (I) converges a.s., similarly to Corollary 1.


## Simulations

You're probably curios, so here it is.

Density of Random $L$ function with $s=1$


## Discussion

- Very different from the distribution of the random harmonic series.
- Why positive?


## Strong Law of Large Numbers

With the aid of the all-mighty Kronecker's Lemma one can use Theorem 5 to give an easy proof the SLLN, generalizations, and analogous results.

## Theorem 6 (Strong Law of Large Numbers)

Let $X_{1}, X_{2}, \ldots$ be independent and identically distributed with finite expectation $\mu$. Let $S_{n}=X_{1}+\cdots+X_{n}$. Then

$$
\lim _{n \rightarrow \infty} \frac{S_{n}}{n} \rightarrow \mu \text { a.s. }
$$

## Discussion

In MATH3160 we usually cover the Weak Law of Large Numbers:

- The WLLN claims the the difference between the empirical mean $S_{n} / n$ and $\mu$ is "large" with asymptotically vanishing probability:

$$
\lim _{n \rightarrow \infty} P\left(\left|\frac{S_{n}}{n}-\mu\right|>\epsilon\right)=0
$$

There is no statement on actual convergence of the empirical means.

- The proof you usually see is based on Chebychev's inequality and assumes finite second moment.


## Proof of Theorem 6

Lemma 7 (Kronecker's Lemma: summation by parts)
Suppose that
$-0<a_{1}<a_{2}<\ldots$ with $\lim _{n \rightarrow \infty} a_{n}=\infty$; and

- $\sum_{n=1}^{\infty} \frac{x_{n}}{a_{n}}$ converges.
then

$$
\lim _{N \rightarrow \infty} \frac{1}{a_{N}} \sum_{n=1}^{N} x_{n}=0
$$

Now for the proof.

- WLOG, assume $\mu=0$.
- Apply Theorem 5 to the series $Y_{n}=X_{n} / n$ to conclude that $\sum_{n=1}^{\infty} \frac{X_{n}}{n}$ converges a.s.
- Apply Kronecker's lemma with $x_{n}=X_{n}$ and $a_{n}=n$, to conclude that

$$
\frac{S_{N}}{N}=\frac{\sum_{n=1}^{N} X_{n}}{N} \rightarrow 0 \text { a.s. }
$$

Done. Thank you.


[^0]:    ${ }^{1}$ From Robert Hugh's lecture

