

8.3.2021

2. The Erdős-Kac Theorem

The "simplest" additive function
is $\omega(n) = \text{nb. of } p \mid n, \text{ without multiplicity,}$
($\omega(12) = 2$)

i.e.

$$\omega(p^v) = 1, \quad \forall p, \forall v \geq 1$$

We saw last week that because

$$\sum_p \frac{1}{p} = +\infty \quad (\text{Euler})$$

ω does not satisfy the assumptions
of Erdős-Wintner.

Nevertheless, we try to understand the
values of ω for $n \leq N, N \rightarrow +\infty$.

Experience of E-W Th. suggests
to compare ω on $\Omega_N = \{1, \dots, N\}$
with

$$\sum_p B_p, \quad \mathbb{P}(B_p = 1) = \frac{1}{p}$$

(B_p) independent

[since $w(n) = \sum_p B_p(n)$,
 $B_p(n) = 1 \Leftrightarrow p|n$]

But this series diverges a.s. because
 $\{B_p = 1\}$ occurs a.s. infinitely
 often. almost surely

But if we note that

$$w(n) = \sum_{p \leq N} B_p(n)$$

we can try an approximation of

w on Ω_N by $\sum_{p \leq N} B_p$.

First question: if

$$X_N = \sum_{p \leq N} B_p,$$

what is the asymptotic behavior as
 $N \rightarrow +\infty$?

Theorem [B.7.2] Central Limit Theorem
[Special case of CLT]

If $(Y_n)_{n \geq 1}$ are independent Bernoulli r.v. with $P(Y_n=1) = \gamma_n$,

and if $\sum_{n \geq 1} \gamma_n(1-\gamma_n) = +\infty$, then

$$\frac{\sum_{n \leq N} Y_n - \sum_{n \leq N} \gamma_n}{\sqrt{\sum_{n \leq N} \gamma_n(1-\gamma_n)}} \xrightarrow{\text{law}} \mathcal{N}(0,1)$$

$$P(\mathcal{N}(0,1) \in [a,b]) = \frac{1}{\sqrt{2\pi}} \int_a^b e^{-t^2/2} dt$$

This suggests strongly:

Theorem (Erdős - Kac, ~ 1939)
[Th. 2.3.1]

Let X_N on Ω_N be defined

by $X_N(n) = \omega(n)$. Then

$$\frac{X_N - \log \log N}{\sqrt{\log \log N}} \xrightarrow[N \rightarrow \infty]{\text{law}} \mathcal{N}(0,1).$$

Note: $\log \log N$ arises as the asymptotic behavior of $\sum_{p \leq N} \frac{1}{p}$.

Sketch of the proof of the CLT

Th. (Lévy Criterion; simple form)

$d \geq 1$, (X_n) r.v. with values in \mathbb{R}^d

X r.v. with values in \mathbb{R}^d

Then $X_n \xrightarrow[n \rightarrow \infty]{\text{law}} X$

\Updownarrow

for all $t \in \mathbb{R}^d$, $\varphi_{X_n}(t) \xrightarrow[n \rightarrow \infty]{} \varphi_X(t)$

where the characteristic functions are

$$\varphi_{X_n}(t) = \mathbb{E} \left(e^{it \cdot X_n} \right)$$

$$\left[t \cdot x = t_1 x_1 + \dots + t_d x_d \right]$$

Note: (1) \Downarrow easy because $\varphi_X(t) = \mathbb{E}(f(x))$
for $f(x) = e^{it \cdot x}$, continuous + bounded

(2) The "difficult" version of the Lévy Criterion does not require to know that the law of X is a probability measure.

Application to CLT: exploit independence to get

$$\mathbb{E} \left(\exp \left(it \frac{\sum_{n \leq N} Y_n - \sum_{n \leq N} y_n}{\sqrt{\sum_{n \leq N} \gamma_n}} \right) \right)$$

$$\stackrel{\text{indep.}}{=} \prod_{n \leq N} \mathbb{E} \left[\exp \left(it \frac{Y_n - y_n}{\sqrt{\sum_{n \leq N} \gamma_n}} \right) \right]$$

and then one uses Taylor expansion of exponential to get the CLT naturally (cf. B. 7.1).

Exercise - Check that ("obvious")

adaptation of the idea of proof of

The converse is true: from $X_n \xrightarrow{\text{law}} X$

one can deduce

$$\forall h \geq 0, \quad \mathbb{E}(X_n^h) \longrightarrow \mathbb{E}(X^h).$$

Note: (1) This is easy if all X_n are uniformly bounded ^(say $|X_n| \leq A$), because then

$f(x) = x^h$ can be restricted to a

bounded continuous function, and

moreover they generate a dense

subspace of $C([-A, A])$

[Weierstrass Theorem]

(2) Without condition (3), there

are known counterexamples.

Proof of The Erdős - Kac Theorem

Step 1 - Truncation

Step 2 - Convergence of moments:

$$\text{let } X_{-N} = \sum_{p \leq N} B_p$$

$$X_N = \sum_{p \leq N} B_p$$

$$\sigma_N = \sum_{p \leq N} \frac{1}{p} \quad (\sim \log \log N)$$

$$\mathbb{E} \left(\left(\frac{X_N - \sigma_N}{\sqrt{\sigma_N}} \right)^k \right) = \mathbb{E} \left(\left(\frac{X_N - \sigma_N}{\sqrt{\sigma_N}} \right)^k \right) + (\text{error})$$

Step 3 - Converse of method of moments

applies to $\frac{X_N - \sigma_N}{\sqrt{\sigma_N}} \left[\xrightarrow[\text{CLT}]{\text{law}} \mathcal{N}(0, 1) \right]$

so we get

$$\mathbb{E} \left(\left(\frac{X_N - \sigma_N}{\sqrt{\sigma_N}} \right)^k \right) \xrightarrow[N \rightarrow \infty]{} \mathbb{E} (\mathcal{N}^k)$$

and hence convergence in law by direct method of moments.

Step 1. We note that "very few integers $n \leq N$ have many large prime factors": let

$$Q = N^{1/(\log \log N)^{1/3}}$$

Then

$$\omega(n) - (\log \log N)^{1/3} \leq \sum_{\substack{p|n \\ p \leq Q}} 1 \leq \omega(n)$$

(because the difference is the number

of $p|n$ s.t. $Q < p \leq N$, and if

there are d of them, then

$$N^{d/(\log \log N)^{1/3}} = Q^d \leq n \leq N$$

$$\text{so } d \leq (\log \log N)^{1/3} \text{]}.$$

One can see then that E-K Th.

follows from the variant where

we consider

$$\frac{\sum_{p \leq Q} \frac{B_p}{p} - \sigma_N}{\sqrt{\sigma_N}}.$$

Step 2.

Notation change:

centered
r.v.

$$\frac{X_N}{\sigma_N} = \sum_{p \leq Q} \left(\frac{B_p}{P} - \frac{1}{P} \right)$$

$$X_N = \sum_{p \leq Q} \left(B_p - \frac{1}{P} \right)$$

$$\sigma_N = \sum_{p \leq N} \frac{1}{P} \sim \sum_{p \leq Q} \frac{1}{P}$$

We compute [for fixed $h \geq 0$]

$$\mathbb{E}_N \left(\left(\frac{X_N}{\sigma_N} \right)^h \right)$$

$$= \frac{1}{\sigma_N^{h/2}} \mathbb{E}_N \left[\left(\sum_{p \leq Q} \left(\frac{B_p}{P} - \frac{1}{P} \right) \right)^h \right]$$

$$= \frac{1}{\sigma_N^{h/2}} \sum_{\substack{p_1, \dots, p_h \\ p_i \leq Q}} \mathbb{E}_N \left[\prod_{i=1}^h \left(\frac{B_{p_i}}{p_i} - \frac{1}{p_i} \right) \right]$$

$f(n)$
where f only
depends on modulo
 $q = p_1 \cdots p_h$

So by Th. 1.3.3 + its Corollary

we have

$$\left[\mathbb{E}_N (f(n \bmod q)) = \mathbb{E}_{\mathbb{Z}/q\mathbb{Z}} (f) + O\left(\frac{\|f\|_1}{N}\right) \right.$$

Here

$$f(x) = \prod_{i=1}^k \left(\delta_{p_i}(x) - \frac{1}{p_i} \right)$$

where

$$\delta_{p_i}(x) = \begin{cases} 1 & \text{if } x \bmod p_i = 0 \\ 0 & \text{otherwise} \end{cases}$$

so $\|f\|_1 = \sum_x |f(x)| \leq q \leq Q^k$
 $(= o(N))$

and moreover

$$\left[\mathbb{E}_{\mathbb{Z}/q\mathbb{Z}} (f) = \mathbb{E} \left(\prod_{i=1}^k \left(\beta_{p_i} - \frac{1}{p_i} \right) \right) \right.$$

by a small computation.

So we conclude:

$$\begin{aligned} & \mathbb{E}_N \left(\left(\frac{\chi_N}{\sqrt{\sigma_N}} \right)^k \right) \\ &= \frac{1}{\sigma_N^{k/2}} \sum_{\substack{p_i \leq Q \\ p_1, \dots, p_k}} \left(\mathbb{E} \left(\prod_{i=1}^k \left(\beta_{p_i} - \frac{1}{p_i} \right) \right) + O\left(\frac{Q^k}{N}\right) \right) \end{aligned}$$

$$= \mathbb{E} \left[\left(\frac{X_N}{\sqrt{\sigma_N}} \right)^h \right] + O \left(\frac{Q^{2h}}{N} \right)$$

(by reversing the computation).

Our choice of $Q = N^{1/(\log \log N)^{1/3}}$
 ensures that $\frac{Q^{2h}}{N} \xrightarrow{N \rightarrow \infty} 0$.

Step 3. By the C-L-T we know that

$$\frac{X_N}{\sqrt{\sigma_N}} \xrightarrow[N \rightarrow \infty]{\text{law}} \mathcal{N}(0, 1)$$

and so we will get

$$\mathbb{E} \left(\left(\frac{X_N}{\sqrt{\sigma_N}} \right)^h \right) \rightarrow \mathbb{E} (\mathcal{N}^h)$$

if the converse of method of moments applies, and we conclude

$$\frac{X_N}{\sqrt{\sigma_N}} \xrightarrow{\text{law}} \mathcal{N}(0, 1)$$

[because one knows that the

moments are not too big so that

$$\sum_{k \geq 0} \frac{\mathbb{E}(|W|^k)}{k!} z^k$$

has positive radius of convergence
(in fact $+\infty$)]

We need to prove that there exists

$$c_k \geq 0 \text{ s.t. } \mathbb{E} \left(\left(\frac{X_N}{\sqrt{\sigma_N}} \right)^k \right) \leq c_k \quad (*)$$

for all $N \geq 1$.

Claim: $\exists c \geq 0$ s.t. for all N

$$\left[\max \left(\mathbb{E} \left(\exp \left(\frac{X_N}{\sqrt{\sigma_N}} \right) \right), \mathbb{E} \left(\exp \left(- \frac{X_N}{\sqrt{\sigma_N}} \right) \right) \right) \leq c \right]$$

[In fact $c = \exp(1)$ works...]

Observe that

$$\exists c_k, \quad |x|^k \leq c_k \left(e^x + e^{-x} \right)$$

for all $x \in \mathbb{R}$, to conclude proof

of $(*)$

Let $t \in [-1, 1]$. B_p independence

$$\begin{aligned} \mathbb{E} \left(\exp \left(t \frac{X_N}{\sqrt{\sigma_N}} \right) \right) \\ = \prod_{p \leq N} \mathbb{E} \left(\exp \left(\frac{t (B_p - 1/p)}{\sqrt{\sigma_N}} \right) \right) \end{aligned}$$

For $N \geq N_0$ (so that $\sigma_{N_0} \geq 1$) we get \exp (reals in $[-1, 1]$) and use

$$\forall x \in [-1, 1], e^x \leq 1 + x + x^2$$

so

$$\begin{aligned} \mathbb{E} \left(\exp \left(\frac{t (B_p - 1/p)}{\sqrt{\sigma_N}} \right) \right) \\ \leq 1 + \mathbb{E} \left(\frac{t (B_p - 1/p)}{\sqrt{\sigma_N}} \right) \\ = 0 \\ + \frac{t^2}{\sigma_N} \underbrace{\mathbb{E} \left((B_p - 1/p)^2 \right)}_{\mathbb{V}(B_p)} \end{aligned}$$

Using $1 + x \leq e^x$ for $x \in \mathbb{R}$

we get

$$\prod_{p \leq Q} \mathbb{E} \left(\exp \left(\frac{t (B_p - 1/p)}{\sqrt{\sigma_N}} \right) \right)$$

$$\leq \exp \left(\frac{t^2}{\sigma_N} \underbrace{\sum_{p \leq Q} \psi(B_p)} \right)$$

$$= \sum_{p \leq Q} \frac{1}{p} \left(1 - \frac{1}{p} \right)$$

$$\leq \exp(t^2)$$