

THM (ERDŐS-KAC, 1940)

LET  $P(U_n = k) = \frac{1}{n}$ ,  $k \in \{1, 2, \dots, n\}$

LET  $Z_n$  DENOTE THE NUMBER OF PRIME DIVISORS OF  $U_n$ . THEN

$$\frac{Z_n - \ln(\ln(n))}{\sqrt{\ln(\ln(n))}} \Rightarrow \mathcal{N}(0, 1)$$

PROOF: DENOTE BY  $\mathcal{P}$  THE SET OF PRIMES

CLAIM:  $\sum_{\substack{p \in \mathcal{P} \\ p \leq n}} \frac{1}{p} = \ln(\ln(n)) + O(1)$

PROOF: (WE WILL ONLY PROVE THE  $\geq$  PART)

$$\ln(n) = \int_1^n \frac{1}{x} dx \leq \sum_{k=1}^n \frac{1}{k} \leq \prod_{\substack{p \in \mathcal{P} \\ p \leq n}} \left(1 + \frac{1}{p} + \frac{1}{p^2} + \dots\right)$$

$$\prod_{\substack{p \in \mathcal{P} \\ p \leq n}} \frac{1}{1 - \frac{1}{p}}, \text{ THUS } \ln(\ln(n)) \leq \sum_{\substack{p \in \mathcal{P} \\ p \leq n}} -\ln\left(1 - \frac{1}{p}\right)$$

$$= \sum_{\substack{p \in \mathcal{P} \\ p \leq n}} -\left(-\frac{1}{p} + O\left(\frac{1}{p^2}\right)\right) = \sum_{\substack{p \in \mathcal{P} \\ p \leq n}} \frac{1}{p} + O(1) \checkmark$$

BACK TO PROOF OF ERDŐS-KAC:

LET  $\mathbb{Y}_{m|P} := \mathbb{1}[P \text{ DIVIDES } U_m]$

THEN  $Z_m = \sum_{\substack{P \in \mathcal{P} \\ P \leq m}} \mathbb{Y}_{m|P}$

$E(\mathbb{Y}_{m|P}) = P(P \text{ DIVIDES } U_m) = \frac{1}{m} \cdot \lfloor \frac{m}{P} \rfloor \approx \frac{1}{P}$

(NOTE:  $(\mathbb{Y}_{m|P})_{P \in \mathcal{P}}$  ARE NOT INDEPENDENT)

TRICK: TRUNCATION. LET US DEFINE

$T_m = \sum_{\substack{P \in \mathcal{P} \\ P \leq d_m}} \mathbb{Y}_{m|P}$

WHERE

$d_m := \frac{m}{\ln(\ln(m))}$

ENOUGH TO SHOW:



$\frac{T_m - \ln(\ln(m))}{\sqrt{\ln(\ln(m))}} \Rightarrow \mathcal{N}(0,1)$

BECAUSE:  $\ln(d_m) = \ln(m) / \ln(\ln(m))$

$\ln(\ln(d_m)) = \ln(\ln(m)) - \ln(\ln(\ln(m)))$

THUS  $T_m \leq Z_m$  AND

$E(Z_m - T_m) \leq \sum_{d_m < P \leq m} \frac{1}{P} \approx \ln(\ln(\ln(m))) + O(1)$

BY CLAIM FROM PAGE 129

THUS  $\frac{Z_n - T_n}{\sqrt{\ln(\ln(n))}} \xrightarrow{A} 0$  BY MARKOV'S INEQ.

HENCE IF WE PROVE  $(*)$  THEN  $(\ddot{\circ})$  WILL FOLLOW.

LET  $(X_{Y,P})_{P \in \mathcal{P}}$  BE INDEP.,  $X_{Y,P} \stackrel{B}{\sim} \text{BER}\left(\frac{1}{P}\right)$

LET  $S_n := \sum_{\substack{P \in \mathcal{P} \\ P \leq d_n}} X_{Y,P}$

LEMMA:  $\frac{S_n - \ln(\ln(n))}{\sqrt{\ln(\ln(n))}} \xrightarrow{D} \mathcal{N}(0, 1)$

PROOF: ENOUGH TO SHOW:

$\frac{S_n - \ln(\ln(d_n))}{\sqrt{\ln(\ln(d_n))}} \xrightarrow{E} \mathcal{N}(0, 1) \leftarrow (\ddot{\circ})$

SINCE  $\ln(\ln(d_n)) \stackrel{F}{=} \ln(\ln(n)) - \ln(\ln(\ln(n)))$

WE WILL USE LINDBERBERG TO PROVE  $(\ddot{\circ})$ .

$$E(S_n) \stackrel{G}{=} \sum_{\substack{P \in \mathcal{P} \\ P \leq d_n}} \frac{1}{P} \stackrel{H}{=} \ln(\ln(d_n)) + O(1)$$

$$\text{Var}(S_n) \stackrel{I}{=} \sum_{\substack{P \in \mathcal{P} \\ P \leq d_n}} \left(\frac{1}{P} - \frac{1}{P^2}\right) \stackrel{J}{=} \ln(\ln(d_n)) + O(1)$$

LINDBERGF'S CONDITION:

LET  $\tilde{X}_{Y_P} := X_{Y_P} - E(X_{Y_P})$  THEN  $N \left[ \tilde{X}_{Y_P} \right] \leq 1$

GIVEN SOME  $\epsilon > 0$ :

$$\lim_{n \rightarrow \infty} \frac{1}{\ln(\ln(d_n))} \sum_{\substack{P \in P \\ P \leq d_n}} E \left( |\tilde{X}_{Y_P}|^2 \cdot \mathbb{1} \left[ |\tilde{X}_{Y_P}| > \epsilon \cdot \sqrt{\ln(\ln(d_n))} \right] \right) = 0$$

SINCE  $\epsilon \cdot \sqrt{\ln(\ln(d_n))} > 1$  IF  $n$  IS LARGE ENOUGH

LET  $S_n^* := \frac{S_n - \ln(\ln(n))}{\sqrt{\ln(\ln(n))}}$        $T_n^* = \frac{T_n - \ln(\ln(n))}{\sqrt{\ln(\ln(n))}}$

KNOW:  $S_n^* \Rightarrow N(0,1)$

WANT:  $T_n^* \Rightarrow N(0,1)$

ACTUALLY WE KNOW MORE (HW 9.1):

$\forall z \in \mathbb{N} : \lim_{n \rightarrow \infty} E \left( (S_n^*)^z \right) = E(X^z), \quad X \sim N(0,1)$

WE WILL NOW SHOW:

$\forall z \in \mathbb{N} : \lim_{n \rightarrow \infty} E \left( (T_n^*)^z \right) = E(X^z), \quad -||-$

LEMMA:  $|E(S_m^z) - E(T_m^z)| \stackrel{A}{\leq} \frac{1}{n} \cdot (d_m)^z$

PROOF: LET  $N := \sum_{\substack{P \in \mathcal{P} \\ P \leq d_m}} 1$  B

NOTE:  
 $N \leq d_m$  C

$(X_1 + \dots + X_N)^z \stackrel{D}{=} \sum_{l=1}^N \sum_{\substack{z_1, \dots, z_l \geq 1 \\ z_1 + \dots + z_l = z}} \sum_{1 \leq m_1 < m_2 < \dots < m_l \leq N} C(b; z_1, \dots, z_l) \cdot X_{m_1}^{z_1} \dots X_{m_l}^{z_l} \geq 0$

THUS:  $E(S_m^z) \stackrel{E}{=} \dots \cdot E(X_{m_1}^{z_1} \dots X_{m_l}^{z_l})$

NOW IF  $1 < p_1 < p_2 < \dots < p_l \leq d_m$  THEN

$E(X_{p_1}^{z_1} \dots X_{p_l}^{z_l}) \stackrel{F}{=} E(X_{p_1} \dots X_{p_l}) \stackrel{G}{=} \frac{1}{p_1 \cdot p_2 \cdot \dots \cdot p_l}$

$E(X_{m_1 p_1}^{z_1} \dots X_{m_l p_l}^{z_l}) \stackrel{H}{=} E(X_{m_1 p_1} \dots X_{m_l p_l}) \stackrel{I}{=} \frac{1}{n} \cdot \left[ \frac{n}{p_1 \cdot p_2 \cdot \dots \cdot p_l} \right]$

THUS:  $0 \leq E(S_m^z) - E(T_m^z) \stackrel{J}{\leq} \dots \stackrel{K}{\leq}$

$\sum_{l=1}^N \sum_{z_1, \dots, z_l} \sum_{\dots} C(b; z_1, \dots, z_l) \cdot \frac{1}{n} \stackrel{L}{=}$

$= \frac{1}{n} \cdot (1 + 1 + \dots + 1)^z = \frac{1}{n} \cdot N^z \stackrel{M}{\leq} \frac{1}{n} \cdot (d_m)^z$

COROLLARY:

$$\forall k \in \mathbb{N}: \left| \mathbb{E}[(S_m^*)^k] - \mathbb{E}[(T_m^*)^k] \right| \xrightarrow[\infty]{n} 0 \quad \text{A}$$

PROOF:  $\mathbb{E}[(S_m^*)^k] \stackrel{\text{B}}{=} \frac{1}{\ln(\ln(m))^{k/2}} \mathbb{E}[(S_m - \ln(\ln(m)))^k]$

$$\stackrel{\text{C}}{=} \frac{1}{\ln(\ln(m))^{k/2}} \sum_{l=0}^k \binom{k}{l} \cdot \mathbb{E}(S_m^l) \cdot (-\ln(\ln(m)))^{k-l}$$

THUS  $\left| \mathbb{E}[(S_m^*)^k] - \mathbb{E}[(T_m^*)^k] \right| \stackrel{\text{D}}{=} \left| \frac{1}{\ln(\ln(m))^{k/2}} \sum_{l=0}^k \binom{k}{l} \cdot (\mathbb{E}(S_m^l) - \mathbb{E}(T_m^l)) \cdot (-\ln(\ln(m)))^{k-l} \right|$

$$\leq \sum_{l=0}^k \binom{k}{l} \cdot \frac{1}{n} \cdot d_m^l \cdot \ln(\ln(m))^{k-l} \stackrel{\text{E}}{=} \sum_{l=0}^k \binom{k}{l} \cdot \frac{1}{n} \cdot d_m^l \cdot \ln(\ln(m))^{k-l} \stackrel{\text{F}}{=} \frac{1}{n} \cdot (d_m \cdot \ln(\ln(m)))^k \xrightarrow[\infty]{n} 0 \quad \text{G}$$

BECAUSE  $d_m = m^{1/\ln(\ln(m))}$ , THUS  $\forall \epsilon > 0$

$d_m \leq m^\epsilon$  IF  $m$  IS LARGE ENOUGH.

THUS

$$\forall r \in \mathbb{N} : \lim_{n \rightarrow \infty} \mathbb{E} \left( (T_n^*)^r \right) = \mathbb{E} \left( X^r \right), \quad X \sim \mathcal{N}(0,1)$$

IT REMAINS TO SHOW THAT THIS

IMPLIES  $T_n^* \Rightarrow \mathcal{N}(0,1)$

THE MOMENT PROBLEM:

$$\text{IF } \forall r \in \mathbb{N} : \mathbb{E}(V^r) = \mathbb{E}(W^r) < +\infty$$

DOES THIS IMPLY THAT  $V \sim W$ ?

UNFORTUNATELY NOT, E.G. THE LOG-NORMAL DISTRIBUTION IS NOT DETERMINED BY ITS MOMENTS (HW 9.2). HOWEVER:

LEMMA: IF  $M_r = \mathbb{E}(V^r) = \mathbb{E}(W^r)$  AND

$$\limsup_{n \rightarrow \infty} \left( \frac{|M_n|}{n!} \right)^{1/n} =: R^{-1} < +\infty \quad \text{THEN } V \sim W.$$

PROOF: SEE NEXT PAGE

PROOF:  $\Psi(t) \stackrel{\text{A}}{=} \mathbb{E}(e^{it \cdot V}) \stackrel{\text{B}}{=} \sum_{k=0}^{\infty} \frac{M_k}{k!} i^k \cdot t^k$

THIS POWER SERIES WILL HAVE A RADIUS OF CONVERGENCE  $R > 0$ ,

THUS  $\Psi(t)$  UNIQUELY EXTENDS AS AN ANALYTIC FUNCTION TO THE STRIP  $\{t \in \mathbb{C} : \text{Im}(t) \in (-R, R)\}$ , SEE PAGE 85.

THUS  $\Psi_V(t) \stackrel{\text{C}}{=} \Psi_W(t)$ , THUS  $V \stackrel{\text{D}}{\sim} W$

EX:  $N(0, 1)$ :  $M(\lambda) \stackrel{\text{E}}{=} e^{\lambda^2/2} \stackrel{\text{F}}{=} \sum_{l=0}^{\infty} \frac{(\lambda^2/2)^l}{l!}$

THUS  $\sum_{k=0}^{\infty} \frac{M_k}{k!} \lambda^k \stackrel{\text{G}}{=} \sum_{l=0}^{\infty} \frac{\lambda^{2l}}{2^l \cdot l!}$  | THUS

$M_k = 0$  IF  $k$  IS ODD AND

IF  $k = 2l$  THEN  $M_k = M_{2l} \stackrel{\text{H}}{=} \frac{(2l)!}{2^l \cdot l!}$

RADIUS OF CONVERGENCE:  
 $R = +\infty$



THM: LET  $M_{n,r} \stackrel{A}{=} \mathbb{E}((V_n)^r) < +\infty$

IF  $\forall r \in \mathbb{N}$ :  $\lim_{n \rightarrow \infty} M_{n,r} \stackrel{B}{=} M_r$  AND

THE SEQUENCE OF MOMENTS  $(M_r)_{r=0}^{\infty}$

UNIQUELY DETERMINES A DISTRIBUTION

(RANDOM VARIABLE):  $V$ , THEN  $V_n \stackrel{C}{\Rightarrow} V$

PROOF:  $(V_n)_{n=1}^{\infty}$  IS A TIGHT SEQUENCE:

$$\mathbb{P}(|V_n| \geq k) \stackrel{D}{=} \mathbb{P}(V_n^2 \geq k^2) \stackrel{E}{\leq} \frac{M_{n,2}}{k^2}$$

MARKOV

THUS GIVEN SOME  $\varepsilon > 0$  WE HAVE

$$\sup_n \mathbb{P}(|V_n| \geq k) \stackrel{F}{\leq} \varepsilon \quad \text{IF } k \stackrel{G}{=} \sqrt{\frac{\sup_n M_{n,2}}{\varepsilon}}$$

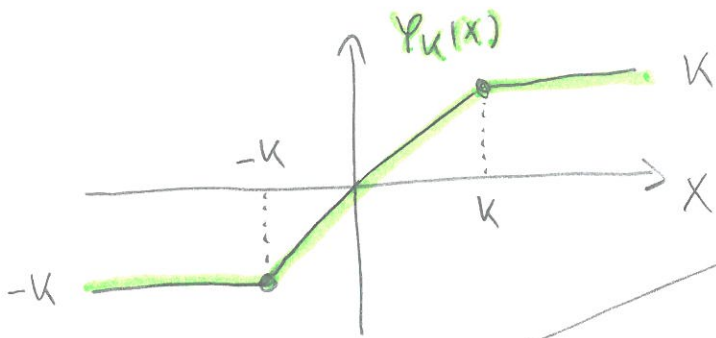
NOW WE SHOW THAT IF  $V_{n_i}$  IS A SUBSEQUENCE WHICH CONVERGES

IN DISTRIBUTION  $(V_{n_i} \stackrel{H}{\Rightarrow} \tilde{V})$  THEN

$$\mathbb{E}(\tilde{V}^r) \stackrel{I}{=} M_r, \quad r \in \mathbb{N}$$

INDEED: LET US DEFINE

$$\varphi_k(x) = x \cdot \mathbb{1}[|x| \leq k] + \text{sign}(x) \cdot k \cdot \mathbb{1}[|x| > k]$$



IF  $\alpha$  IS EVEN:  
BY MONOTONE CONV.

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IF  $\alpha$  IS ODD:  
DOMINATED CONV.

$$\mathbb{E}(\tilde{V}^\alpha) = \lim_{k \rightarrow \infty} \mathbb{E}(\varphi_k(\tilde{V})^\alpha)$$

SINCE  $V_{n'} \Rightarrow \tilde{V}$   
 $\varphi_k$  IS BOUNDED CONTINUOUS

$$= \lim_{k \rightarrow \infty} \lim_{n' \rightarrow \infty} \mathbb{E}(\varphi_k(V_{n'})^\alpha)$$

$$= \lim_{k \rightarrow \infty} \lim_{n' \rightarrow \infty} (\mathbb{E}(V_{n'}^\alpha) - \mathbb{E}(V_{n'}^\alpha - \varphi_k(V_{n'})^\alpha))$$

$$= \lim_{n' \rightarrow \infty} M_{n', \alpha} - \lim_{k \rightarrow \infty} \lim_{n' \rightarrow \infty} \mathbb{E}(V_{n'}^\alpha - \varphi_k(V_{n'})^\alpha)$$

$M_\alpha$

WE WILL SHOW THAT THIS IS ZERO

$$|\mathbb{E}(V_n^\alpha - \varphi_k(V_n)^\alpha)| \leq \mathbb{E}(|V_n|^\alpha \cdot \mathbb{1}[|V_n| \geq k])$$

$$\leq \sqrt{\mathbb{E}(V_n^{2\alpha})} \cdot \sqrt{\mathbb{E}(\mathbb{1}[|V_n| \geq k]^2)}$$

$$\sqrt{M_{n, 2\alpha}} \cdot \sqrt{P(|V_n| \geq k)}$$

CAUCHY-SCHWARZ

$$\textcircled{\text{smiley}} \leq \sqrt{M_{n,2r}} \cdot \sqrt{M_{n,2}/K^2} \leq \frac{\sqrt{\sup_n M_{n,2r} \cdot M_{n,2}}}{K}$$

↑ A
B

MARKOV INEQ

THUS  $\textcircled{\text{frowny}} \leq \lim_{K \rightarrow \infty} \frac{\sqrt{\sup_n M_{n,2r} \cdot M_{n,2}}}{K} = 0 \quad \checkmark$

C D

THUS  $E(\tilde{V}^r) = M_r, \forall r \in \mathbb{N}$ , THUS  $\tilde{V} \stackrel{\sim}{\sim} V$

E F

TO SHOW THAT THE WHOLE SEQUENCE  $V_n$  CONVERGES IN DISTRIBUTION TO  $V$ , JUST REPEAT THE ARGUMENT FROM PAGE 110-112.