## Number Theory Handy & Ramanujan (920) Thm let $V(x) = *\{P \in P : P \mid x\}$ . If $f(n) \to \infty \approx n \to \infty$ then

#  $\omega \in [n]$ :  $V(\omega)$ -lnln  $n > f(n)\sqrt{\ln n}$  = o(n)It i.e. "almost all " n have "very close to" lnln prime factors

Let ([n], p) be a p-space with  $P(\omega) = \frac{1}{n} V \omega \in [n]$ .

Xp(w) = { | if Plw X = Zper xp and Ln = fin , Inhin

EX = \( \frac{1}{n} \) \( \frac{1}{p} \) \( \frac{1}{p} \) = \( \ln n n + A + O(\frac{1}{p} n) \) (see Apostol: Introduction to Analytic Number theory p. 90. Thm 4.12)

Theory p. 90. Thm 4.12)

TO(n) = | P = n|

 $\sum_{\substack{p \neq 8 \\ p \neq 8}} Cov(X_p, X_g) = \sum_{\substack{p \neq 8 \\ p \neq 8}} \left(\frac{\frac{n}{p-1}}{n} - \frac{\frac{n}{p-1}}{n} \frac{\frac{n}{p-1}}{n}\right) \leq \frac{1}{n} \sum_{\substack{p \neq 8 \\ p \neq 8}} \left(\frac{1}{p} + \frac{1}{8}\right) = \frac{2\pi c(n)}{n} \sum_{\substack{p \neq 8 \\ p \neq 8}} \frac{1}{p} = o(1)$   $P_1 g \in \mathbb{R}_{\leq n}$   $P_2 \in \mathbb{R}_{\leq n}$   $P_3 \in \mathbb{R}_{\leq n}$   $P_4 \in \mathbb{R}_{\leq n}$   $P_8 \in \mathbb{R}_{\leq n}$   $P_8 \in \mathbb{R}_{\leq n}$   $P_8 \in \mathbb{R}_{\leq n}$ 

 $LHS/n = P(|X-lnhn| > L_n) \leq P(|X-EX| + |EX-lnhn| > L_n)$ 

< Pr(|X-EX|>= Ln) as n sufficiently : EX=lulin+A+o(1) and Ln -> oo as n -> oo

 $\leq 4 \frac{\text{Var} X}{L_n^2} \leq 4 \frac{\mathcal{E}(X_p) + \sum_{i} cov(X_p, X_g)}{L_n^2} = 4 \frac{\mathcal{E}X + o(i)}{L_n^2} \leq 4 \frac{\text{Inlinn} + A + o(i)}{L_n^2} = o(i)$ 

The second moment method is an effective tool in number theory. Let  $\nu(n)$  denote the number of primes p dividing n. (We do not count multiplicity though it would make little difference.) The following result says, roughly, that "almost all" n have "very close to"  $\ln \ln n$  prime factors. This was first shown by Hardy and Ramanujan in 1920 by a quite complicated argument. We give a remarkably simple proof of Paul Turan [1934], a proof that played a key role in the development of probabilistic methods in number theory.

Theorem 2.1 Let  $\omega(n) \to \infty$  arbitrarily slowly. Then number of x in  $\{1,\ldots,n\}$  such that

$$|\nu(x) - \ln \ln n| > \omega(n) \sqrt{\ln \ln n}$$

is o(n).

**Proof.** Let x be randomly chosen from  $\{1, \ldots, n\}$ . For p prime set

$$X_p = \left\{ \begin{array}{ll} 1 & \text{if } p|x \\ 0 & \text{otherwise} \end{array} \right.$$

Set  $M=n^{1/10}$  and set  $X=\sum X_p$ , the summation over all primes  $p\leq M$ . As no  $x\leq n$  can have more than ten prime factors larger than M we have  $\nu(x)-10\leq X(x)\leq \nu(x)$  so that large deviation bounds on X will translate into asymptotically similar bounds for  $\nu$ . (Here 10 could be any large constant.) Now

$$E[X_p] = \frac{\lfloor n/p \rfloor}{n}$$

As  $y - 1 < \lfloor y \rfloor \le y$ 

$$E[X_p] = 1/p + O(1/n)$$

By linearity of expectation

$$E[X] = \sum_{p \le M} \frac{1}{p} + O(\frac{1}{n}) = \ln \ln n + O(1)$$

Now we find an asymptotic expression for  $Var[X] = \sum_{p \leq M} Var[X_p] + \sum_{p \neq q} Cov[X_p, X_q]$ . As  $Var[X_p] = \frac{1}{p}(1 - \frac{1}{p}) + O(\frac{1}{n})$ ,

$$\sum_{p \le M} Var[X_p] = \sum_{p \le M} \frac{1}{p} + O(1) = \ln \ln n + O(1)$$

With p, q distinct primes,  $X_p X_q = 1$  if and only if p|x and q|x which occurs if and only if pq|x. Hence

$$\begin{aligned} &Cov[X_p, X_q] = E[X_p]E[X_q] - E[X_pX_q] \\ &= \frac{\lfloor n/pq \rfloor}{n} - \frac{\lfloor n/p \rfloor}{n} \frac{\lfloor n/q \rfloor}{n} \\ &\leq \frac{1}{pq} - (\frac{1}{p} - \frac{1}{n})(\frac{1}{q} - \frac{1}{n}) \\ &\leq \frac{1}{n}(\frac{1}{p} + \frac{1}{q}) \end{aligned}$$

Thus

$$\sum_{p \neq q} Cov[X_p, X_q] \le \frac{1}{n} \sum_{p \neq q} \frac{1}{p} + \frac{1}{q} \le \frac{2M}{n} \sum_{p \neq q} \frac{1}{p}$$

Thus

$$\sum_{p \neq q} Cov[X_p, X_q] = O(n^{-9/10} \ln \ln n) = o(1)$$

That is, the covariances do not affect the variance,  $Var[X] = \ln \ln n + O(1)$  and Chebyschev's Inequality actually gives

$$\Pr[|X - \ln \ln n| > \lambda \sqrt{\ln \ln n}] < \lambda^{-2} + o(1)$$

for any constant  $\lambda$ . As  $|X - \nu| \le 10$  the same holds for  $\nu$ .  $\square$ 

In a classic paper Paul Erdős and Marc Kac [1940] showed, essentially, that  $\nu$  does behave like a normal distribution with mean and variance  $\ln \ln n$ . Here is their precise result.

Theorem 2.2. Let  $\lambda$  be fixed, positive, negative or zero. Then

$$\lim_{n\to\infty}\frac{1}{n}|\{x:1\leq x\leq n,\nu(x)\geq \ln\ln n+\lambda\sqrt{\ln\ln n}\}|=\int_{\lambda}^{\infty}\frac{1}{\sqrt{2\pi}}e^{-t^2/2}dt$$

We outline the argument, emphasizing the similarities to Turan's proof. Fix a function s(n) with  $s(n) \to \infty$  and  $s(n) = o((\ln \ln n)^{1/2})$  - e. g.  $s(n) = \ln \ln \ln n$ . Set  $M = n^{1/s(n)}$ . Set  $X = \sum X_p$ , the summation over all primes  $p \le M$ . As no  $x \le n$  can have more than s(n) prime factors greater than M we have  $\nu(x) - s(n) \le X(x) \le \nu(x)$  so that it suffices to show Theorem 2.2 with  $\nu$  replaced by X. Let  $Y_p$  be independent random variables with  $\Pr[Y_p = 1] = p^{-1}$ ,  $\Pr[Y_p = 0] = 1 - p^{-1}$  and set  $Y = \sum Y_p$ , the summation over all primes  $p \le M$ . This Y represents an idealized version of X. Set

$$\mu = E[Y] = \sum_{p \le M} p^{-1} = \ln \ln n + o((\ln \ln n)^{1/2})$$

and

$$\sigma^2 = Var[Y] = \sum_{p \le M} p^{-1} (1 - p^{-1}) \sim \ln \ln n$$

and define the normalized  $\tilde{Y} = (Y - \mu)/\sigma$ . From the Central Limit Theorem (well, an appropriately powerful form of it!)  $\tilde{Y}$  approaches the standard normal N and  $E[\tilde{Y}^k] \to E[N^k]$  for every positive integer k. Set  $\tilde{X} = (X - \mu)/\sigma$ . We compare  $\tilde{X}, \tilde{Y}$ .

For any distinct primes  $p_1, \ldots, p_s \leq M$ 

$$E[X_{p_1}\cdots X_{p_s}] - E[Y_{p_1}\cdots Y_{p_s}] = \frac{\lfloor \frac{n}{p_1\cdots p_s}\rfloor}{n} - \frac{1}{p_1\cdots p_s} = O(n^{-1})$$

We let k be an arbitrary fixed positive integer and compare  $E[\tilde{X}^k]$  and  $E[\tilde{Y}^k]$ . Expanding,  $\tilde{X}^k$  is a polynomial in X with coefficients  $n^{o(1)}$ . Further expanding each  $X^j = (\sum X_p)^j$  - always reducing  $X_p^a$  to  $X_p$  when  $a \geq 2$  -gives the sum of  $O(M^k) = n^{o(1)}$  terms of the form  $X_{p_1} \cdots X_{p_s}$ . The same expansion applies to  $\tilde{Y}$ . As the corresponding terms have expectations within  $O(n^{-1})$  the total difference

$$E[\tilde{X}^k] - E[\tilde{Y}^k] = n^{-1 + o(1)} = o(1)$$

Hence each moment of  $\tilde{X}$  approach that of the standard normal N. A standard, though nontrivial, theorem in probability theorem gives that  $\tilde{X}$  must therefore approach N in distribution.  $\square$ 

We recall the famous quotation of G. H. Hardy:

317 is a prime, not because we think so, or because our minds are shaped in one way rather than another, but because it is so, because mathematical reality is built that way.

How ironic - though not contradictory - that the methods of probability theory can lead to a greater understanding of the prime factorization of integers.