

PERIODIC OSCILLATIONS OF THE VARIANCE OF TRIE STATISTICS AND RELATED STRUCTURES

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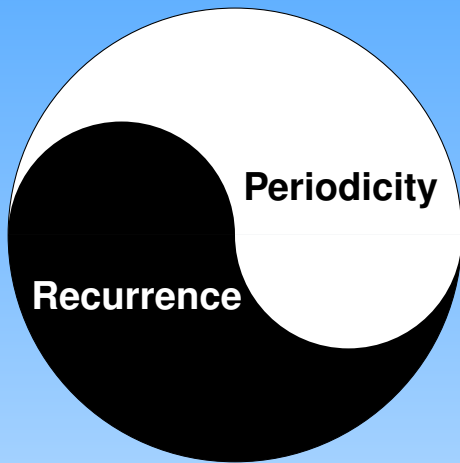
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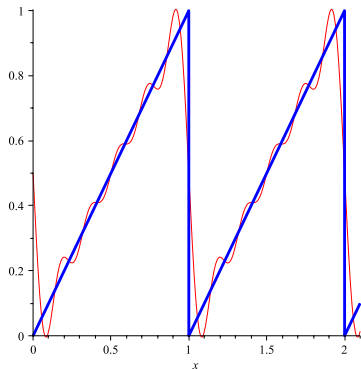


PERIODICITY EVERYWHERE IN ALGORITHMS



FLOOR & SAW-TOOTH FUNCTIONS

Fractional part



Fourier series

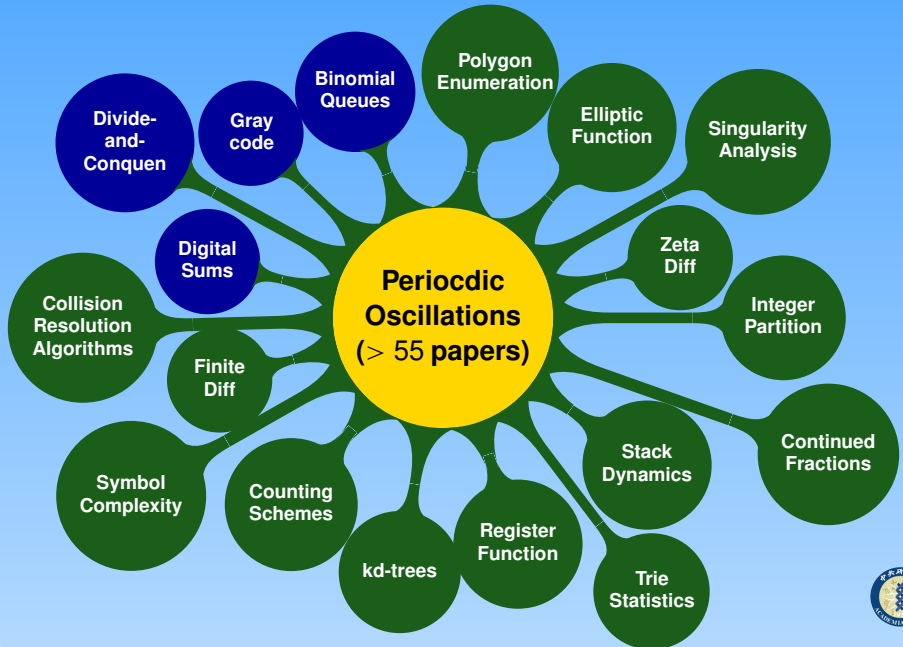
$$\begin{aligned}x - \lfloor x \rfloor \\ &= \frac{1}{2} - \frac{1}{\pi} \sum_{k \geq 1} \frac{\sin(2k\pi x)}{k}\end{aligned}$$

(\Leftarrow use five terms)

$$f_n = f_{\lfloor n/2 \rfloor} + f_{\lceil n/2 \rceil} + g_n$$



PERIODIC OSCILLATIONS IN PF'S WORKS



PERIODIC OSCILLATIONS EVERYWHERE

Two common types of periodic functions

- C^0 but nowhere differentiable:

$$\begin{aligned} \frac{1}{n} \sum_{0 \leq k < n} \#(\mathbf{1}\text{'s in } k\text{'s binary expansion}) \\ = \frac{1}{2} \log_2 n + \underbrace{F(\log_2 n)}_{\text{fractal}} \quad (\text{identity!}) \end{aligned}$$

- C^∞ with very small amplitude:

E(internal nodes in a random trie of n keys)

$$= \frac{n}{\log 2} + \underbrace{G(\log_2 n)}_{|\cdot| \leq 10^{-6}} n + O(1)$$

How to compute the Fourier expansion?



TWO MAJOR ANALYTIC APPROACHES

Mellin transform: \mathbf{E} (size of tries)

$$\begin{aligned}\tilde{f}(z) &= 2\tilde{f}(z/2) + 1 - (1+z)e^{-z} \\ &= -\frac{1}{2\pi i} \int_{(-\frac{3}{2})} \frac{(s+1)\Gamma(s)}{1-2^{s+1}} z^{-s} ds \\ \implies a_n &= \frac{n}{\log 2} \left(1 - \sum_{k \neq 0} \chi_k \Gamma(-1 + \chi_k) n^{-\chi_k} \right) - 1 + \dots\end{aligned}$$

Singularity analysis: \mathbf{E} (size of m -ary search trees)

$$\begin{aligned}(1-z)^{m-1} f^{(m-1)}(z) - m! f(z) &= \frac{(m-1)!}{1-z} \\ f(z) &= \sum_{1 \leq j < m} c_j (1-z)^{-\lambda_j} \quad (\theta^{\overline{m-1}} - m! = \prod_{1 \leq j < m} (\theta - \lambda_j)) \\ \implies a_n &= \frac{n+1}{2(H_m - 1)} + \sum_{2 \leq j < m} c_j \binom{n + \lambda_j - 1}{n}\end{aligned}$$



BINOMIAL SPLITTING PROCESS



BINOMIAL SPLITTING PROCESS

Structure
of size n

$$|\cdot| = L_n$$

$$L_n \approx \text{Binomial}$$
$$L_n + R_n \approx n$$

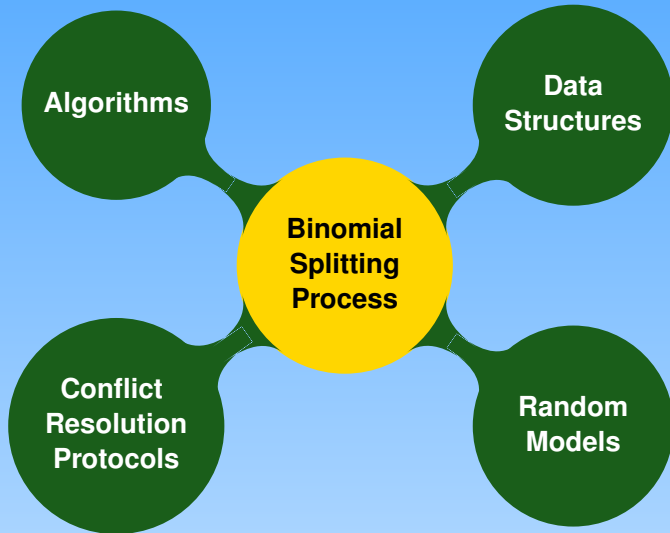
$$|\cdot| = R_n$$

$$X_n \stackrel{d}{=} X_{L_n} + X_{R_n}^* + T_n$$





A HUGE NUMBER OF EXAMPLES

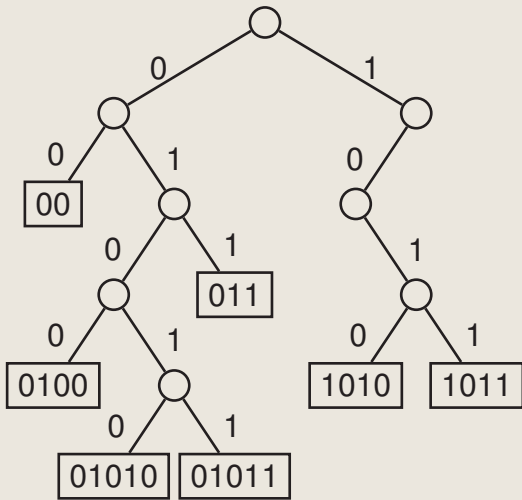




TRIES



TRIE (0 → LEFT, 1 → RIGHT)



USEFULNESS OF TRIES

Simple, efficient data structures

Widely used in diverse applications: data compression, suffix trees, internet IP addresses lookup, automatically correcting words in texts, taxonomies of regular language, ...

Fundamental, prototype data structures

- **have a large number of variations, extensions**
- **closely connected to several *splitting procedures using coin-flipping*: collision resolution in multi-access (or broadcast) communication models, loser selection or leader election, etc.**
- **have direct *combinatorial interpretations* in terms of words, urn models, extendible hashing, etc.**



TRIES: LOG-SHAPE CHARACTERISTICS

- **Depth**: Devroye (1982, 1992, 1999), Mendelson (1982), Pittel (1986), Jacquet & Régnier (1986), Szpankowski (1987, 1988), Kirschenhofer & Prodinger (1988), Jacquet & Szpankowski (1994), Louchard (1994), Schachinger (2001), Fayolle & Ward (2005), Christohpi & Mahmoud (2008), Janson (2012).
- **Height**: Mendelson (1982), Flajolet & Steyaert (1982), Flajolet (1983), Devroye (1984), Pittel (1985, 1986), Jacquet & Régnier (1986), Szpankowski (1991), Devroye (1992, 1999, 2002), Clément, Flajolet, Vallée (2001), Broutin & Devroye (2008), Devroye, Lugosi, Park & Szpankowski (2009), Janson (2012).
- **Horton-Strahler # & stack-size**: Devroye & Kruszewski (1996), Nebel (2000, 2002), Bourdon, Nebel & Vallée (2001).
- **One-sided height (or leader election or loser selection)**: Prodinger (1995), Fill, Mahmoud, Szpankowski (1996), Janson & Szpankowski (1997), Ward & Szpankowski (2004, 2005), Louchard & Prodinger (2006), Mahmoud & Ward (2008).



TRIES: LINEAR SHAPE CHARACTERISTICS

- **Sum(internal nodes)** : Jacquet & Régnier (1988), Régnier & Jacquet (1989), Kirschenhofer & Prodinger (1991), Jacquet & Szpankowski (1994), Rachev & Rüschemdorf (1995), Schachinger (1995), Knuth (1998), Clément, Flajolet, Vallée (2001), Schachinger (2001, 2004), Neininger & Rüschemdorf (2004), Wagner (2009).
- **Node sorts** : Tsybakov & Mikhailov (1978), Capetanakis (1979), Mendelson (1982), Flajolet (1983), Mathys & Flajolet (1985), Kaplan & Gulko (1985), Szpankowski (1988), Kirschenhofer & Prodinger (1988, 1991), Janssen & de Jong (2000), Schachinger (2001), Nguyễn-Thé (2003), Mahmoud & Ward (2008), Wagner (2010), Janson (2012), Gaither, Homma, Sellke & Ward (2012).
- **External path length** : Devroye (1984), Szpankowski (1987), Kirschenhofer, Prodinger, Szpankowski (1989), Schachinger (1995, 2001, 2004), Clément, Flajolet, Vallée (2001), Nguyễn-Thé (2003), Neininger & Rüschemdorf (2004).



RANDOM TRIES

The simplest Bernoulli model

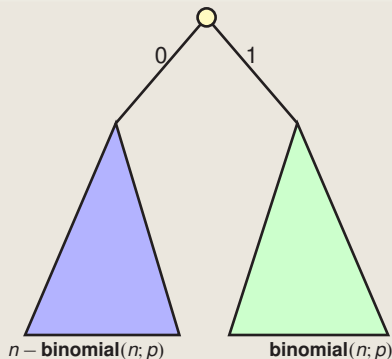
Input = $\{Y_1, Y_2, \dots, Y_n\}$ iid

$$Y_i = \{Y_{i,j}\}_{j \geq 1},$$

$$\begin{cases} \mathbf{P}(Y_{i,j} = 1) = p \\ \mathbf{P}(Y_{i,j} = 0) = q = 1 - p. \end{cases}$$

Random trie: the trie constructed from this sequence of RVs.

$$X_n \stackrel{d}{=} X_{L_n} + X_{R_n}^* + T_n$$



Many other general models in the literature

SIZE OF TRIES



SIZE OF RANDOM TRIES (BERNOULLI MODEL)

$$X_n \stackrel{d}{=} X_{\text{Binom}(n;p)} + X_{n-\text{Binom}(n;p)} + 1$$

$\mu_n := \mathbf{E}(X_n)$ satisfies $\mu_0 = \mu_1 = 0$ and

$$\mu_n = \sum_{0 \leq k \leq n} \underbrace{\binom{n}{k} p^k q^{n-k}}_{\pi_{n,k}} (\mu_k + \mu_{n-k}) + 1 \quad (n \geq 2).$$

Poisson generating function $\tilde{f}(z) := e^{-z} \sum_n \mu_n z^n / n!$

$$\tilde{f}(z) = \tilde{f}(pz) + \tilde{f}(qz) + 1 - (1+z)e^{-z}.$$

Mellin transform ($G_1(s) := -(s+1)\Gamma(s)$)

$$\tilde{f}(z) = \frac{1}{2\pi i} \int_{(-\frac{3}{2})} \frac{G_1(s)}{1 - p^{-s} - q^{-s}} z^{-s} ds$$



EXPECTED SIZE OF RANDOM TRIES

Growth order + calculus of residues
($h := -p \log p - q \log q$)

$$\frac{\tilde{f}(z)}{z} \sim \frac{1}{h} + \begin{cases} \frac{1}{h} \overbrace{\sum_{k \neq 0} G_1(-1 + \chi_k) z^{-\chi_k}}^{\text{Periodic}}, & \text{if } \frac{\log p}{\log q} = \frac{r}{\ell}, (r, \ell) = 1 \\ 0, & \text{if } \frac{\log p}{\log q} \notin \mathbf{Q} \end{cases}$$

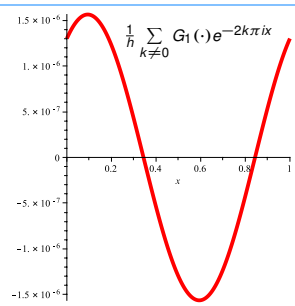
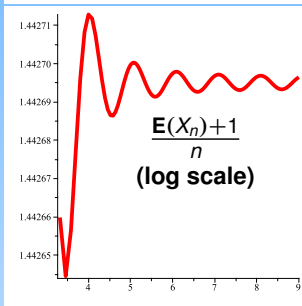
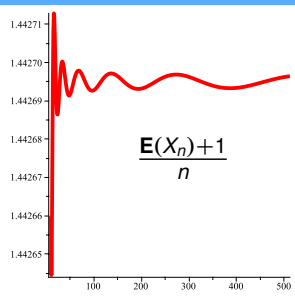
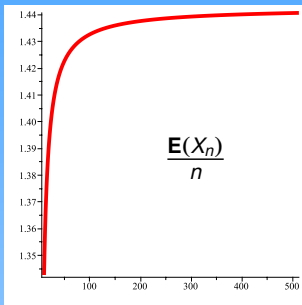
Uniformly for $\Re(z) > 0$ ($\chi_k := 2kr\pi i / \log p$)

Analytic de-Poissonization (Jacquet-Szpankowski)

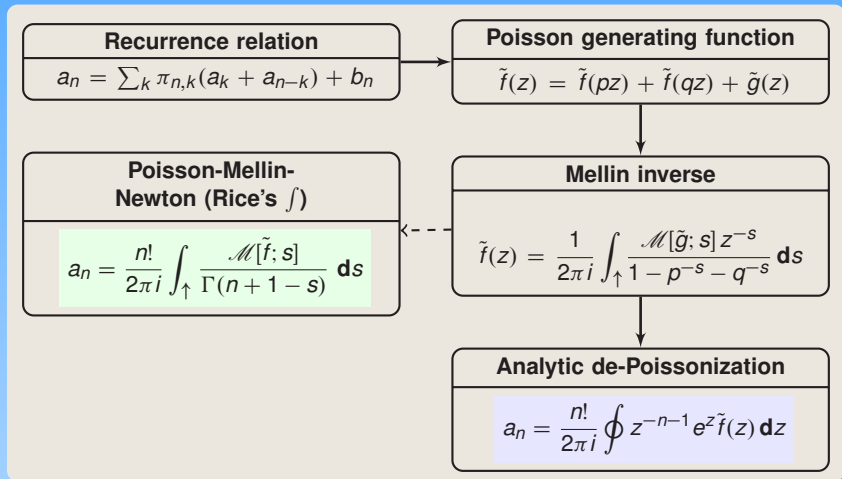
$$\frac{\mu_n}{n} \sim \frac{1}{h} + \begin{cases} \frac{1}{h} \overbrace{\sum_{k \neq 0} G_1(-1 + \chi_k) n^{-\chi_k}}^{\text{Periodic}}, & \text{if } \frac{\log p}{\log q} = \frac{r}{\ell}, (r, \ell) = 1 \\ 0, & \text{if } \frac{\log p}{\log q} \notin \mathbf{Q} \end{cases}$$



PERIODIC OSCILLATIONS: $\rho = \frac{1}{2}$



THE MELLIN-DE-POISSONIZATION APPROACH



More systematic tools



ASYMPTOTICS OF $V(X_n)$

New result: $\rho = \frac{1}{2}$ ($\chi_k := \frac{2k\pi i}{\log 2}$)

$$\frac{V(X_n)}{n} = \frac{1}{\log 2} \sum_{k \in \mathbb{Z}} G(-1 + \chi_k) n^{-\chi_k} + o(1)$$

$$G(s) = (s+1)\Gamma(s) \left(1 - \frac{s^2 + 4s + 8}{2^{s+3}} \right) + 2 \sum_{j \geq 1} \frac{(-1)^j j(j(j+s+1)-1)\Gamma(j+s+1)}{(j+1)!(2^j-1)}$$

$(\Re(s) > -2)$



ASYMPTOTICS OF $V(X_n)$

Average value of the periodic function

$$\frac{G(-1)}{\log 2} = \frac{1}{\log 2} \left(\frac{1}{4} + 2 \sum_{j \geq 1} \frac{(-1)^j (j-1)}{2^j - 1} \right) \approx 0.845858623076001 \dots$$

Régnier & Jacquet (1988), Kirschenhofer & Prodinger (1991)

$$\frac{G(-1)}{\log 2} = \frac{1}{2 \log 2} - \frac{1}{\log^2 2} - \frac{2}{\log 2} \sum_{j \geq 1} \frac{(-1)^j}{2^j - 1} - \frac{4\pi^2}{\log^3 2} \sum_{j \geq 1} \frac{j}{\sinh \frac{2j\pi^2}{\log 2}}$$

Mahmoud (1992)

$$\frac{G(-1)}{\log 2} = \frac{1}{\log 2} \left(\frac{1}{2} + 2 \sum_{j \geq 1} \frac{1}{2^j + 1} \right) - \frac{1}{\log^2 2} - \frac{4\pi^2}{\log^3 2} \sum_{j \geq 1} \frac{j}{\sinh \frac{2j\pi^2}{\log 2}}$$

An identity (*direct proof by residue calculus*)

$$\sum_{j \geq 1} \frac{(-1)^j j}{2^j - 1} = \frac{1}{8} - \frac{1}{2 \log 2} - \underbrace{\frac{2\pi^2}{\log^2 2} \sum_{j \geq 1} \frac{j}{\sinh \frac{2j\pi^2}{\log 2}}}_{\leq 1.1 \times 10^{-10}}$$

$$|\cdot| \leq 1.1 \times 10^{-10}$$



ASYMPTOTICS OF $V(X_n)$

Fourier coefficients: $k \neq 0$

$$G(-1 + \chi_k) = -\frac{\chi_k \Gamma(-1 + \chi_k) (1 + \chi_k)^2}{4} + 2 \sum_{j \geq 1} \frac{(-1)^j j(j + \chi_k - 1) \Gamma(j + \chi_k)}{(j + 1)! (2^j - 1)}$$

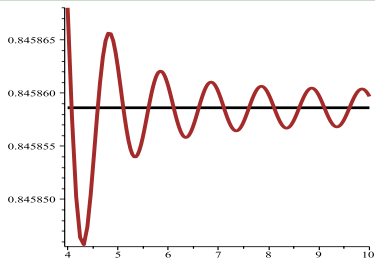
Kirschenhofer & Prodinger (1991)

$$\begin{aligned} G(-1 + \chi_k) &= -3\chi_k \Gamma(-1 + \chi_k) + \frac{1}{\log 2} \sum_{\substack{j+m=k \\ j,m \neq 0}} \chi_j \Gamma(-1 + \chi_j) \chi_m \Gamma(1 + \chi_m) \\ &\quad - (1 - \chi_k)(2 - \chi_k) \Gamma(\chi_k) \left(\frac{1}{2} - \sum_{j \geq 1} \frac{(\chi_k + j) \binom{-\chi_k}{j-1}}{(j+1)(2^j - 1)} \right) \\ &\quad - 2\Gamma(1 + \chi_k) \left(\frac{5 - \chi_k}{4(1 - \chi_k)} - \sum_{j \geq 1} \frac{(\chi_k + j + 1) \binom{-\chi_k - 1}{j-1}}{(j+1)(2^j - 1)} \right) \\ &\quad - \frac{\chi_k \Gamma(1 + \chi_k)}{\log 2} \end{aligned}$$

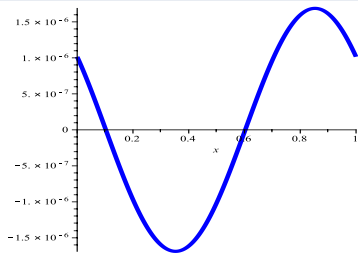


PERIODIC OSCILLATIONS: $\rho = \frac{1}{2}$

$$\frac{V(X_n)}{n}$$



$$\frac{1}{\log 2} \sum_{k \in \mathbb{Z}} G(-1 + \chi_k) n^{-\chi_k}$$



$$\frac{1}{\log 2} \sum_{k \neq 0} |G(-1 + \chi_k)| \leq 1.7 \times 10^{-6}$$



ASYMPTOTICS OF $V(X_n)$

New result: $\frac{\log p}{\log q} \notin \mathbb{Q}$

(implicit in Jacquet & Régnier, 1989)

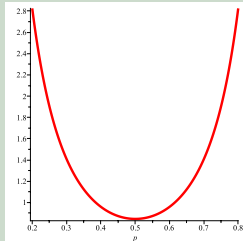
$$\frac{V(X_n)}{n} = \frac{1}{h} \left(\frac{1}{2} - \frac{1}{h} + 2 \sum_{j \geq 2} \frac{(-1)^j (p^j + q^j)}{1 - p^j - q^j} \right) + o(1)$$

($h := -p \log p - q \log q$)

As $p \rightarrow \frac{1}{2}$

$$\frac{1}{h} \left(\frac{1}{2} - \frac{1}{h} + 2 \sum_{j \geq 2} \frac{(-1)^j (p^j + q^j)}{1 - p^j - q^j} \right)$$

$$\rightarrow \frac{1}{\log 2} \left(\frac{1}{2} - \frac{1}{\log 2} - 2 \sum_{j \geq 1} \frac{(-1)^j}{2^j - 1} \right)$$



ASYMPTOTICS OF $V(X_n)$

New result: $\frac{\log p}{\log q} = \frac{r}{\ell}, (r, \ell) = 1$ ($\chi_k := \frac{2kr\pi i}{\log p}$)

$$\frac{V(X_n)}{n} = \frac{1}{h} \sum_{k \in \mathbb{Z}} G(-1 + \chi_k) n^{-\chi_k} + o(1)$$

same as irrational case

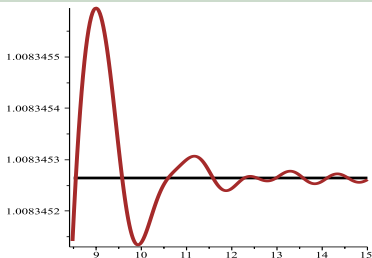
$$G(-1) = \frac{1}{2} - \frac{1}{h} + 2 \sum_{j \geq 2} \frac{(-1)^j (p^j + q^j)}{1 - p^j - q^j} - \frac{1}{h \log p} \sum_{j \geq 1} \frac{4rj\pi^2}{\sinh \frac{2rj\pi^2}{\log p}}$$

$$\begin{aligned} G(-1 + \chi_k) &= \chi_k \Gamma(-1 + \chi_k) \left(1 - \frac{\chi_k + 3}{2^{1+\chi_k}} \right) \\ &\quad - \frac{1}{h} \sum_{j \in \mathbb{Z}} \Gamma(\chi_j + 1) \Gamma(\chi_{k-j} + 1) \\ &\quad - 2 \sum_{j \geq 1} \frac{(-1)^j (j + 1 + \chi_k) \Gamma(j + \chi_k) (p^{j+1} + q^{j+1})}{(j-1)! (j+1) (1 - p^{j+1} - q^{j+1})} \end{aligned}$$

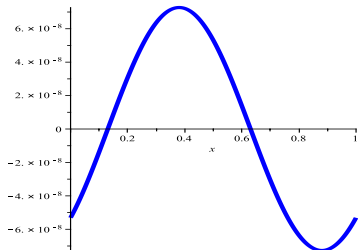


PERIODIC OSCILLATIONS: $q = p^2$ ($p = \frac{\sqrt{5}-1}{2}$)

$$\frac{V(X_n)}{n}$$



$$\frac{1}{h} \sum_{k \in \mathbb{Z}} G(-1 + \chi_k) n^{-\chi_k}$$



$$\frac{1}{h} \sum_{k \neq 0} |G(-1 + \chi_k)| \leq 7.3 \times 10^{-8}$$



METHOD OF PROOF



POISSONIZED VARIANCE

$$\tilde{f}_m(z) := e^{-z} \sum_n \frac{\mathbf{E}(X_n^m)}{n!} z^n$$

- Mean: $\mathbf{E}(X_n) \sim \tilde{f}_1(n)$
- Second moment: $\mathbf{E}(X_n^2) \sim \tilde{f}_2(n)$
- Variance $\mathbf{V}(X_n) = \mathbf{E}(X_n^2) - (\mathbf{E}(X_n))^2$

$$\tilde{D}(n) := \tilde{f}_2(n) - \tilde{f}_1(n)^2 \stackrel{??}{\approx} \mathbf{V}(X_n)?$$

Yes, but not good enough in certain cases.



POISSONIZED VARIANCE

$$\tilde{f}_2(z) = \tilde{D}(z) + \tilde{f}_1(z)^2$$

$$\begin{aligned}\sigma_n^2 &= \mathbf{E}(X_n^2) - \mu_n^2 \\ &= \sum_{j \geq 0} \frac{\tilde{f}_2^{(j)}(n)}{j!} \tau_j(n) - \left(\sum_{j \geq 0} \frac{\tilde{f}_1^{(j)}(n)}{j!} \tau_j(n) \right)^2 \\ &= \tilde{D}(n) - \boxed{n\tilde{f}_1'(n)^2} - n\tilde{f}_1(n)\tilde{f}_1''(n) + \mathbf{s.o.t.}\end{aligned}$$

Take $\tilde{f}_1(z) \asymp z \log z$

$$n\tilde{f}_1'(n)^2 \asymp n(\log n)^2, \quad n\tilde{f}_1(n)\tilde{f}_1''(n) \asymp n \log n.$$

In most such case, $\sigma_n^2 \asymp n \log n$ or $\sigma_n^2 \asymp n$.



OUR APPROACH TO $V(T_n)$

The crucial step

$$\tilde{V}(z) := \tilde{f}_2(z) - \tilde{f}_1(z)^2 - z\tilde{f}'_1(z)^2$$

Then (with $\tilde{f}_2(n) = \tilde{V}(n) + \tilde{f}_1(n)^2 + n\tilde{f}'_1(n)^2$)

$$\begin{aligned}\sigma_n^2 &= \sum_{j \geq 0} \frac{\tilde{f}_2^{(j)}(n)}{j!} \tau_j(n) - \left(\sum_{j \geq 0} \frac{\tilde{f}_1^{(j)}(n)}{j!} \tau_j(n) \right)^2 \\ &= \tilde{V}(n) - \underbrace{\frac{n}{2} \tilde{V}'''(n) - \frac{n^2}{2} \tilde{f}_1''(n)^2}_{=O(1)} + o(1).\end{aligned}$$

$z\tilde{f}'_1(z)^2 =$ **the right correction term when $\sigma_n^2 \approx n(\log n)^\beta$**



ANALYTIC DE-POISSONIZATION



Poisson-Charlier expansion

$$\tilde{f}(z) := e^{-z} \sum_{j \geq 0} \frac{a_j}{j!} z^j \text{ entire}$$

$$\implies a_n = \sum_{j \geq 0} \frac{\tilde{f}^{(j)}(n)}{j!} \tau_j(n)$$

$$\tau_j(n) = \sum_{0 \leq \ell \leq j} \binom{j}{\ell} (-1)^{j-\ell} \frac{n \cdots (n - \ell + 1)}{n^\ell}$$

First few τ_j

$$\tau_0(n) = 1$$

$$\tau_1(n) = 0$$

$$\tau_2(n) = -n$$

$$\tau_3(n) = 2n$$

$$\tau_4(n) = 3n(n-2)$$

$$\tau_5(n) = -4n(5n-6)$$

...

$\deg(\tau_j(n)) = \lfloor j/2 \rfloor$
(Charlier or Laguerre)



POISSON-CHARLIER EXPANSION

An example: $\tilde{f}(z) = e^{-2z}$

$$(-1)^n = e^{-2n} \sum_{j \geq 0} \frac{(-2)^j}{j!} \tau_j(n).$$

But $(-1)^n \not\sim e^{-2n}$.

Another example: $\tilde{f}(z) = e^z$

$$2^n = e^n \sum_{j \geq 0} \frac{\tau_j(n)}{j!}.$$

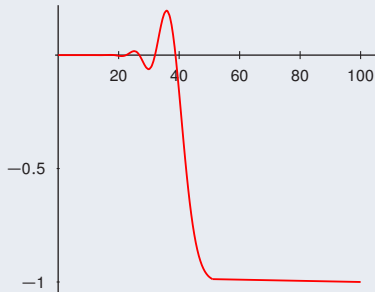
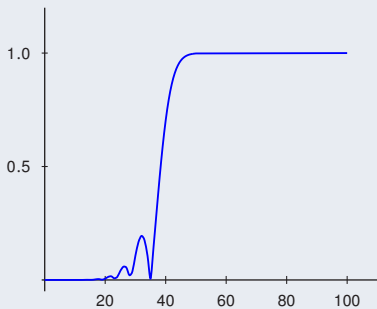
But $2^n \not\sim e^n$.

Major difficulty: prove the asymptotic nature

$$a_n = \sum_{0 \leq j \leq k} \frac{\tilde{f}^{(j)}(n)}{j!} \tau_j(n) + O(\tilde{f}^{(k)}(n) \tau_k(n))$$



$$e^{-2n} \sum_{j \geq 0} \frac{(-2)^j}{j!} \tau_j(n) \longrightarrow (-1)^n \quad (n = 10, 11)$$



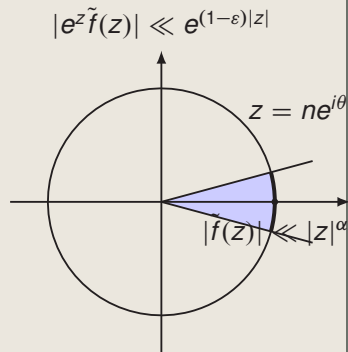
THE MISSING STEPS

Justification of de-Poissonization

Justify

$$- \mu_n = \sum_{j \geq 0} \frac{\tilde{f}_1^{(j)}(n)}{j!} \tau_j(n);$$

$$- \mathbf{E}(X_n^2) = \sum_{j \geq 0} \frac{\tilde{f}_2^{(j)}(n)}{j!} \tau_j(n)$$





ELSEVIER

Theoretical Computer Science 201 (1998) 1–62

Theoretical
Computer Science

Fundamental Study

Analytical depoissonization and its applications

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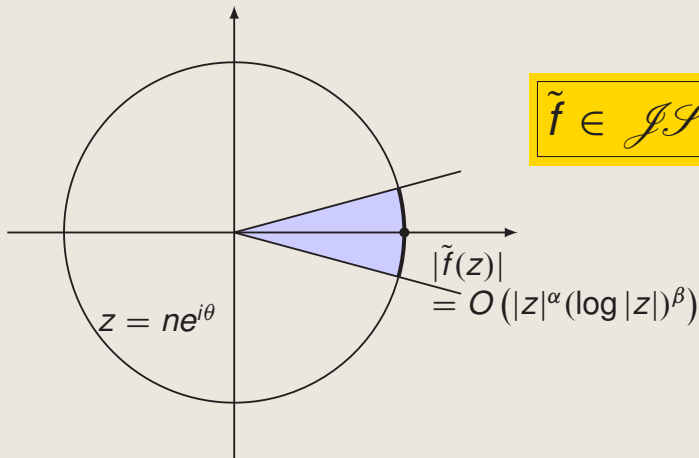
JS-ADMISSIBLE FUNCTIONS



JS-ADMISSIBILITY

\tilde{f} is JS-admissible (combining ideas from Hayman (1956) and Jacquet-Szpankowski (1998))

$$|e^z \tilde{f}(z)| = O(e^{(1-\varepsilon)|z|})$$



PROPERTIES OF JS-ADMISSIBLE FUNCTIONS

If $\tilde{f} \in \mathcal{JS}_{\alpha, \beta}$, then

$$a_n = \sum_{0 \leq j < 2k} \frac{\tilde{f}^{(j)}(n)}{j!} \tau_j(n) + O\left(n^{\alpha-k} \log^\beta n\right),$$

Closure properties: $m \geq 0, \alpha \in (0, 1)$

- (i) $z^m, e^{-\alpha z} \in \mathcal{JS}$.
- (ii) If $\tilde{f} \in \mathcal{JS}$, then $\tilde{P}\tilde{f} \in \mathcal{JS}$ for any polynomial $\tilde{P}(z)$.
- (iii) If $\tilde{f} \in \mathcal{JS}$, then $\tilde{f}(\alpha z) \in \mathcal{JS}$.
- (iv) If $\tilde{f}, \tilde{g} \in \mathcal{JS}$, then $\tilde{f} + \tilde{g} \in \mathcal{JS}$.
- (v) If $\tilde{f}, \tilde{g} \in \mathcal{JS}$, then $\tilde{f}(\alpha z)\tilde{g}((1-\alpha)z) \in \mathcal{JS}$.
- (vi) If $\tilde{f} \in \mathcal{JS}$, then $\tilde{f}^{(m)} \in \mathcal{JS}$.

TECHNICAL PROPERTIES

$$\tilde{f}(z) = \tilde{f}(pz) + \tilde{f}(qz) + \tilde{g}(z)$$

$$\tilde{f} \in \mathcal{J}\mathcal{S} \iff \tilde{g} \in \mathcal{J}\mathcal{S}$$

Closed under a Hadamard product:

$$\tilde{f}(z) := e^{-z} \sum_{n \geq 0} \frac{a_n}{n!} z^n, \tilde{g}(z) := e^{-z} \sum_{n \geq 0} \frac{b_n}{n!} z^n$$

$$\tilde{h}(z) := e^{-z} \sum_{n \geq 0} \frac{a_n b_n}{n!} z^n$$

$$\tilde{f} \in \mathcal{J}\mathcal{S}_{\alpha_1, \beta_1} \ \& \ \tilde{g} \in \mathcal{J}\mathcal{S}_{\alpha_2, \beta_2} \implies \tilde{h} \in \mathcal{J}\mathcal{S}_{\alpha_1 + \alpha_2, \beta_1 + \beta_2}$$

$$\tilde{h}(z) = \tilde{f}(z)\tilde{g}(z) + z\tilde{f}'(z)\tilde{g}'(z)$$

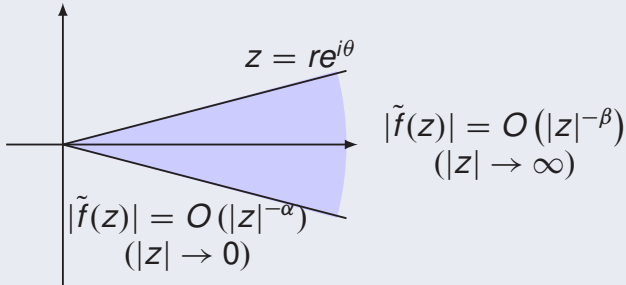
$$+ O(|z|^{\alpha_1 + \alpha_2 - 2} (\log_+ |z|)^{\beta_1 + \beta_2})$$

uniformly as $|z| \rightarrow \infty$ **and** $|\arg(z)| \leq \theta < \pi/2$



MELLIN TRANSFORM

Exponential Smallness Lemma (Flajolet et al., 1995)



$$\alpha < \Re(s) < \beta$$

$$\Rightarrow \left| \int_0^{\infty} e^{-zs} \tilde{f}(z) \, dz \right| = O\left(e^{-\theta |\Im(s)|}\right)$$

Mellin transform then becomes easy



POISSON GENERATING FUNCTIONS

The first two moments ($\tilde{V} := \tilde{f}_2 - \tilde{f}_1^2 - z(\tilde{f}'_1)^2$)

$$\tilde{f}_1(z) = \tilde{f}_1(pz) + \tilde{f}_1(qz) + 1 - (1+z)e^{-z}$$

$$\begin{aligned}\tilde{V}(z) = & \tilde{V}(pz) + \tilde{V}(qz) + pqz \left(\tilde{f}'_1(pz) - \tilde{f}'_1(qz) \right)^2 \\ & + e^{-z}(1+z - (1+2z+z^2+z^3)e^{-z}) \\ & + 2e^{-z}(1+z)(\tilde{f}_1(pz) + \tilde{f}_1(qz)) \\ & - 2z^2e^{-z} \left(p\tilde{f}'_1(pz) + q\tilde{f}'_1(qz) \right)\end{aligned}$$

Both Mellin-doable + JS-admissible

Advantages of the approach

- ***General, simple to use, easily-extendible***
- ***Simpler Fourier expansions: NOT of the form $F_1(x) - F_0^2(x)$.***



OTHER APPROACHES USED FOR VARIANCE

A few approaches used in the literature

- **Complex-analytic:** two-dimensional saddle-point method
- **Second moment:** compute the asymptotics of the 2nd moment, then handle the cancelation. (i) *more precise expansion* for μ_n is needed (ii) handle *analytically the convolution term* $\sum_j \pi_{n,j} \mu_j \mu_{n-j}$;
- **Poissonized variance:** relies on Poisson heuristic

Further manipulation of cancelations needed



GENERALITY

Other trie statistics

- External path length, internal path length, node sorts, imbalance measures, ...

	$p = \frac{1}{2}$	$\frac{\log p}{\log q} \in \mathbf{Q}$	$\frac{\log p}{\log q} \notin \mathbf{Q}$
$\frac{\mathbf{EPL}_n}{n} \sim$	Periodic	$c \log n + \mathbf{Periodic}$	$c \log n + c'$
$\frac{\mathbf{IPL}_n}{n} \sim$	Periodic $(\log n)^2$	(Periodic) $(\log n)^2$	$c(\log n)^2$

- Patricia tries: path length, patterns, ...

$$\begin{aligned} \mathcal{E}(z, y) &= \mathcal{E}(pe^y z, y) \mathcal{E}(qe^y z, y) + \mathcal{E}(pz, y) \\ &\quad + \mathcal{E}(qz, y) - \mathcal{E}(pe^y z, y) - \mathcal{E}(qe^y z, y) \end{aligned}$$

(Kirschenhofer et al., 1988)



GENERALITY

Other BSPs

- Radix sort (Mahmoud et al., 2000)

$$\mathcal{E}(z, y) = (1 - e^y)z + \mathcal{E}\left(\frac{e^y z}{b}, y\right)^b$$

- Conflict resolution algorithms (many papers)

$$\tilde{P}(z, y) = e^y \prod_{1 \leq j \leq r} \tilde{P}(p_j z, y) + (1 - e^y)(1 + z)e^{-z}$$

where $\sum_{1 \leq j \leq r} p_j = 1$, $\tilde{P}(z, y) = e^{-z} \sum_{n \geq 0} \frac{\mathbf{E}(e^{Xny})}{n!} z^n$.



CONFLICT RESOLUTION ALGORITHMS

Rational: $p_j = \rho^{\theta_j}$, $1 \leq j \leq r$; irrational otherwise

$$\frac{\mathbf{E}(X_n)}{n} = \frac{1}{h} \left(1 + \begin{cases} \sum_{k \neq 0} \chi_k \Gamma(-1 + \chi_k) n^{-\chi_k}, & \text{rational} \\ 0, & \text{irrational} \end{cases} \right) + o(1)$$

$$\frac{\mathbf{V}(X_n)}{n} = \frac{1}{h} \left(G(-1) + \begin{cases} \sum_{k \neq 0} G(-1 + \chi_k) n^{-\chi_k}, & \text{rational} \\ 0, & \text{irrational} \end{cases} \right) + o(1)$$

$$\begin{aligned} G(s) = & (s+1)\Gamma(s) \left(1 - \frac{s^2 + 4s + 8}{2^{s+3}} \right) \left(P(s) := \sum_j p_j^s \right) \\ & + 2 \sum_{j \geq 1} \frac{(-1)^j j(j(s+1) - 1) \Gamma(j+s+1) P(j+1)}{(j+1)!(1-P(j+1))} \\ & + \frac{\Gamma(3+s)}{2^{3+s}} - 2 \sum_{j \geq 1} \frac{(-1)^j \Gamma(j+3+s) P(j+1)}{(j-1)!(1-P(j+1))} \\ & - \begin{cases} \frac{1}{h} \sum_{j \in \mathbb{Z}} \Gamma(\chi_j + 1) \Gamma(3+s - \chi_j), & \text{rational} \\ 0, & \text{irrational} \end{cases} \end{aligned}$$



REMARKS

Features of binomial recurrences

- *O-estimate easier, but \sim -estimate harder*
- *mean and limit law easier but asymptotic variance harder*

Limit laws: at least five ways

- **Contraction method (Neininger, Rösler, Rüschemdorf)**
- **Martingale difference (Schacinger, 2000)**
- **Complex-analytic (Jacquet, Régnier, Szpankowski)**
- **Renewal theory (Janson, 2012)**
- **Method of moments (Fuchs, H.)**



