Superexponential decay for the GEM process

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Abstract We show that the GEM process has strong ordering properties: the probability that the k-th largest element in the GEM sequence is beyond the first ck elements (c > 1) decays super-exponentially in k.

Let $\{U_i\}_{i=1}^{\infty}$ denote a sequence of [0, 1] valued i.i.d. random variables, with common law μ possessing a density $p_{\theta}(x) = \theta x^{\theta-1}$. Here, $\theta > 0$ is a fixed known parameter, and throughout we use $\overline{U}_i = 1 - U_i$.

Define the random sequence (GEM process) $A_1 = U_1$ and

$$A_i = U_i \prod_{j=1}^{i-1} \overline{U}_j , i \ge 2.$$

For references and background on the GEM process and its properties, see [2]. Note that stochastically, A_i dominates A_{i+1} , but of course it is still possible that $A_i < A_{i+1}$. Our goal here is to estimate how unlikely is really this reverse inequality. More precisely, let $\{X_i\}$ denote the reordered sequence of $\{A_i\}$. That is, for each i there is a j = j(i) such that $X_i = A_j$ and $X_{i+1} < X_i$. For c > 1, define the event

$$\Omega_{k,c} = \{X_k \text{ is not among } A_i, i < ck\},\$$

and let $P_{\theta,c,k} = \text{Prob}(\Omega_{k,c})$. Our goal is to prove the

Theorem 1.

$$\lim_{k\to\infty} \frac{\log P_{\theta,c,k}}{k\log k} = -\theta(c-1).$$

Proof: We begin by quickly demonstrating a lower bound (which, incidentally, captures the correct order of magnitude but does not exhibit necessarily the most likely event).

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Fix $\alpha > 0$ independent of k, and denote by $\Omega'_{k,c}$ the event

$$\Omega'_{k,c} = \{U_{ck} > \frac{1}{2}, U_j < \frac{\alpha}{(c-1)k}, j=k, k+1, \dots, ck-1\}.$$

Because, in the event $\Omega'_{k,c}$,

$$A_{ck} \ge \frac{1}{2} \left(1 - \frac{\alpha}{(c-1)k} \right)^{(c-1)k} \prod_{i=1}^{k-1} \overline{U}_i \ge \frac{1}{2} e^{-2\alpha} \prod_{i=1}^{k-1} \overline{U}_i,$$

while

$$A_j \le \frac{2\alpha}{(c-1)k} \prod_{i=1}^{k-1} \overline{U}_i, \ j = k, \dots, ck-1,$$

it holds that for all k large enough, $A_{ck} \geq A_j$, $j = k, \ldots, ck-1$. Hence, for such k, $\Omega'_{k,c} \subset \Omega_{k,c}$. Thus, since for some constant $c_{\alpha,c}$ independent of k which may change from line to line,

$$Prob(U_1 < \frac{\alpha}{(c-1)k}) > c_{\alpha,c}k^{-\theta},$$

it holds that

$$P_{\theta,c,k} \ge P(U_{ck} > \frac{1}{2})c_{\alpha,c}^k k^{-\theta(c-1)k}$$

which is more than enough to imply the required lower bound.

We next turn to establish the (harder) complementary upper bound. Note first that

$$P_{\theta,c,k} = \operatorname{Prob}(\exists j \ge ck, \ A_j \ge X_k)$$

$$\le \sum_{j=ck}^{\infty} \operatorname{Prob}(A_j \ge X_k)$$

$$\le \sum_{j=ck}^{\infty} \operatorname{Prob}\left(\text{for some } I \in \mathcal{I}_{j,k}, A_j \ge A_i \ \forall i \in I\right),$$
(1)

where in the last inequality,

$$\mathcal{I}_{i,k} = \{ \text{all subsets of length } j - k \text{ of } \{1, \dots, j-1\} \}.$$

Note that the cardinality of $\mathcal{I}_{j,k}$ is $\binom{j}{k}$, while, from the definition of A_i and the i.i.d. assumption,

$$\max_{I \in \mathcal{I}_{j,k}} \operatorname{Prob} (A_j \ge A_i \ \forall i \in I) \le \operatorname{Prob} (A_j \ge A_i, i = k, \dots, j-1).$$

It thus follows from (1) that

$$P_{\theta,c,k} \leq \sum_{j=ck}^{\infty} {j \choose k} \operatorname{Prob}(A_j \geq A_i, i = k, \dots, j-1)$$

$$\leq \sum_{j=ck}^{\infty} {j \choose k} \operatorname{Prob}(A_{j-k} \geq A_i, i = 1, \dots, j-k-1)$$

$$\leq \sum_{j=ck}^{\infty} {j \choose k} \operatorname{Prob}(U_{j-k} \prod_{\ell=i+1}^{j-k-1} \overline{U}_{\ell} \geq \frac{U_i}{\overline{U}_i}, i = 1, \dots, j-k-2) \stackrel{\triangle}{=} \sum_{j=ck}^{\infty} {j \choose k} P_{j,k}$$

$$(2)$$

Since for $j \ge ck$ there exists a $c_{\alpha,c}$ independent of j,k such that $\binom{j}{k} \le e^{c_{\alpha,c}k\log(j/k)}$, the proof is completed by the following lemma:

Lemma 1. There exists a constant $c_{\theta,c}$, independent of k, j, such that for j > ck,

$$P_{j,k} \le c_{\theta,c} e^{-\theta(j-k)\log k} \,. \tag{3}$$

Proof of Lemma 1: Throughout this proof, we use c_{α} to denote constants, whose values may change from line to line, which are independent of k, j but may depend on θ, c . Let n = j - k. Then

$$P_{j,k} \le \operatorname{Prob}\left(\forall \ 2 \le \ell \le n \,, \ \sum_{j=1}^{\ell-1} \log \overline{U}_j \ge \log V_\ell\right) \stackrel{\triangle}{=} P_n \,,$$

where $V_{\ell} = U_{\ell}/\overline{U}_{\ell}$.

For simplicity in notations, we assume below that both $\log n$ and $n/\log n$ are integers, the general case posing no new difficulties. Define

$$A_{i} = \frac{\sum_{j=i \log n}^{(i+1) \log n - 2} \log \overline{U}_{j}}{\log n}, \quad Z_{i} = \log V_{(i+1) \log n - 1},$$

and let $\mathbf{x} \in \mathbb{R}^{n/\log n}$ have components x_i . Further, let

$$\mathcal{A}_n = \{ \mathbf{x} \in \mathbb{R}^{n/\log n} : 0 > x_i > -n(\log n + 1), x_i = -jn^{-2}, \text{ some integer } j \}.$$

Note that the cardinality of A_n is bounded by $(n^2(n+1)\log n)^{n/\log n} \leq e^{c_{\alpha}n}$. Then,

$$P_{n} \leq \operatorname{Prob}(\sum_{i=1}^{J} A_{i} \geq Z_{j}/\log n, j = 1, \dots, n/\log n)$$

$$\leq \frac{n}{\log n} \operatorname{Prob}(Z_{1} < -n\log^{2} n/2)$$

$$+ \sum_{\mathbf{x} \in A_{n}} \operatorname{Prob}(A_{i} \in [x_{i}, x_{i} + n^{-2}], \sum_{\ell=1}^{i} x_{\ell} + in^{-2} \geq \frac{Z_{i}}{\log n}, i = 1, \dots, n/\log n).$$

Since $\operatorname{Prob}(Z_1 < -n\log^2 n/2) \leq e^{-c_{\alpha}n\log^2 n}$, the bound on the cardinality of \mathcal{A}_n and the independence of the $\{A_i\}$ and $\{Z_i\}$ reveals that for some C_n, C'_n with

$$\log(C_n)/n\log n \to -\infty, \log(C_n')/n\log n \to 0, \tag{4}$$

$$P_n \le C_n + C'_n \max_{\mathbf{x} \in A_n} \prod_{i=1}^{n/\log n} \text{Prob}(x_i + n^{-2} \ge A_i \ge x_i) \prod_{i=1}^{n/\log n} \text{Prob}(Z_i \le \log n \sum_{j=1}^i x_j + \frac{1}{n}).$$
 (5)

Define next

$$\Lambda_{\theta}(\lambda) = \log \left(\int (1-x)^{\lambda} p_{\theta}(x) dx \right) \,,$$

and its Fenchel-Legendre transform

$$\Lambda_{\theta}^{*}(x) = \sup_{\lambda \in \mathbb{R}} (\lambda x - \Lambda(\lambda)).$$

Finally, let $\overline{\Lambda_{\theta}^*(x)} = \min_{y \in [x, x+n^{-2}]} \Lambda_{\theta}^*(y)$. By Cramér's theorem (see, e.g., [1, pg. 27]),

$$\operatorname{Prob}(x_i + n^{-2} \ge A_i \ge x_i) < 2e^{-\log n\overline{\Lambda_{\theta}^*(x_i)}}.$$

On the other hand,

Prob
$$(Z_i \le \log n \sum_{i=1}^i x_j + n^{-1}) \le c_{\alpha} e^{\theta \log n \sum_{j=1}^i x_j}$$
.

Combining the above, and still using C_n , C'_n to denote (possibly different) constants still satisfying (4), one obtains

$$P_{n} \leq C_{n} + C'_{n} \max_{\mathbf{x} \in A_{n}} \exp \left(-\log n \left(\sum_{i=1}^{n/\log n} \overline{\Lambda_{\theta}^{*}}(x_{i}) - \theta \sum_{i=1}^{n/\log n} \sum_{j=1}^{i} x_{j}\right)\right)$$

$$\leq C_{n} + C'_{n} \max_{\phi \in \mathcal{A}} \exp -n \left(\int_{0}^{1} \left(\overline{\Lambda_{\theta}^{*}}(\dot{\phi}_{s}) - \theta \frac{n}{\log n} \phi_{s}\right) ds\right)$$

$$= C_{n} + C'_{n} \max_{\phi \in \mathcal{A}} \exp -n \left(\int_{0}^{1} \left(\overline{\Lambda_{\theta}^{*}}(\frac{\log n \dot{\phi}_{s}}{n}) - \theta \phi_{s}\right) ds\right)$$

$$\leq C_{n} + C'_{n} \max_{\phi \in \mathcal{A}} \exp -n \left(\int_{0}^{1} \left(\Lambda_{\theta}^{*}(\frac{\log n \dot{\phi}_{s}}{n}) - \theta \phi_{s}\right) ds\right) \stackrel{\triangle}{=} C_{n} + C'_{n} \max_{\phi \in \mathcal{A}} \exp -n I_{n}(\phi), \tag{6}$$

where

$$\mathcal{A} = \{ \phi \text{ absolutely continuous, nonincreasing, } \phi_0 = 0 \},$$

the second inequality is obtained by noting that polygonal decreasing functions (at steps of size $\log n/n$) form a subset of \mathcal{A} , and the last one by the continuity of Λ_{θ}^* away from 0 and a change in the value of C'_n .

Let next $\eta \in (0,1)$ be arbitrary. Using the convexity of Λ_{θ}^* , one notes that for $\phi \in \mathcal{A}$,

$$I_n(\phi) \geq \eta \Lambda_{\theta}^*(\frac{\phi_{\eta} \log n}{n\eta}) + (1 - \eta) \Lambda_{\theta}^*(\frac{(\phi_1 - \phi_{\eta}) \log n}{(1 - \eta)n}) - \theta \int_0^1 \phi_s ds.$$

Fixing η and ϕ_{η} , recalling that $\Lambda_{\theta}^* \geq 0$ and that ϕ is non-increasing,

$$\min_{\phi \in \mathcal{A}} I_n(\phi) \ge \min_{\phi_{\eta} < 0} \left(\eta \Lambda_{\theta}^* (\frac{\phi_{\eta} \log n}{n\eta}) - \theta (1 - \eta) \phi_{\eta} \right).$$

In Lemma 2 below, we collect some properties of $\Lambda_{\theta}^*(\cdot)$. In particular, it holds that $\Lambda_{\theta}^*(x) \geq -\theta \log x (1+o(1))$ for x small. A direct optimization over ϕ_{η} reveals then that there exit negative constants $c_1(\eta), c_2(\eta)$ independent of n such that

$$\min_{\phi_{\eta} < 0} \left(\eta \Lambda_{\theta}^*(\frac{\phi_{\eta} \log n}{\eta n}) - \theta(1 - \eta)\phi_{\eta} \right) = \min_{c_2 \log n < \phi_{\eta} < c_1} \left(\eta \Lambda_{\theta}^*(\frac{\phi_{\eta} \log n}{\eta n}) - \theta(1 - \eta)\phi_{\eta} \right) \ge \theta \eta \log n (1 + o(1)) .$$

Taking now $\eta \to 1$ yields

$$\min_{\phi \in \mathcal{A}} I_n(\phi) \ge \theta \log n (1 + o(1)).$$

Substituting back in (6), this concludes the proof of the Lemma 1 and hence of Theorem 1.

The following lemma was used in the course of the proof of Lemma 1.

Lemma 2. Λ_{θ}^* is strictly convex, $\Lambda_{\theta}^*(x) = \infty$ for $x \ge 0$, $\lim_{x \to -\infty} \Lambda_{\theta}^*(x) = \infty$, and $\Lambda_{\theta}^*(y) = 0$ if and only if $y = \int \log(1-x)p_{\theta}(x)dx$. Finally, $\Lambda_{\theta}^*(x) \ge -\theta \log(x)(1+o(1))$ for x small.

Proof of Lemma 2: The first part of the lemma is a trivial consequence of the fact that $\Lambda_{\theta}(\lambda) < \infty$ for all λ with $|\lambda| < \lambda_0(\theta)$ (c.f. [1, pg. 28]). To see the second part, note first that for $\theta = 1$ and x < 0, U_1 is uniformly distributed and a straight forward computation reveals that $\Lambda_{\theta}(\lambda) = -\log(\lambda + 1)$ for $\lambda > -1$ and $\Lambda_{\theta}^*(x) = -1 - x - \log(-x)$. We use below c_{θ} to denote various constants, whose value may change from line to line but which are independent of λ . To see the claim for $0 < \theta < 1$, simply note that for $\lambda > 1$,

$$\int_{0}^{1} y^{\lambda} (1-y)^{\theta-1} dy = \int_{0}^{1-\lambda^{-1}} y^{\lambda} (1-y)^{\theta-1} dy + \int_{1-\lambda^{-1}}^{1} y^{\lambda} (1-y)^{\theta-1} dy
\leq (1-\lambda^{-1})^{\lambda} \lambda^{1-\theta} \left(\theta \frac{1-\lambda^{-1}}{1+\lambda} + \lambda^{-1} \right) \leq c_{\theta} \lambda^{-\theta} ,$$

whereas for $\theta > 1$ and $\lambda > 0$,

$$\int_{0}^{1} y^{\lambda} (1-y)^{\theta-1} dy \leq \sum_{k=0}^{\infty} \int_{1-(k+1)\lambda^{-1}}^{1-k\lambda^{-1}} y^{\lambda} (1-y)^{\theta-1} dy
\leq \sum_{k=0}^{\infty} \lambda^{-\theta} k^{\theta-1} e^{-k} \leq c_{\theta} \lambda^{-\theta}.$$

Hence, for any $\theta > 0$ and $\lambda > 1$,

$$\Lambda_{\theta}(\lambda) \leq c_{\theta} - \theta \log \lambda$$
.

It follows that, with the choice $\lambda = -x^{-1}$,

$$\Lambda_{\theta}^*(x) \ge -1 - c_{\theta} - \theta \log(-x),$$

as claimed. \Box

Remark: In fact, the exact form of p_{θ} was never used. In order to get Theorem 1, all that is needed is that the common law μ of the (0,1) valued i.i.d. random variables U_i possesses a density near 0, 1 such that, for some positive constants θ, α_{θ} ,

$$\lim_{x \to \infty} \frac{1}{x} \log \operatorname{Prob} \left(\log(U_i / \overline{U}_i) < -x \right) = -\alpha_{\theta}, \tag{7}$$

$$\lim_{x \to 0^-} \frac{\Lambda^*(x)}{\log(-x)} = -\theta. \tag{8}$$

Here, $\Lambda^*(x)$ is the Fenchel-Legendre transform of

$$\Lambda(\theta) = \log \left(\int_0^1 (1-x)^{\lambda} \mu(dx) \right) .$$

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References

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