

Regularity of the last-passage percolation time constant on complete directed acyclic graphs

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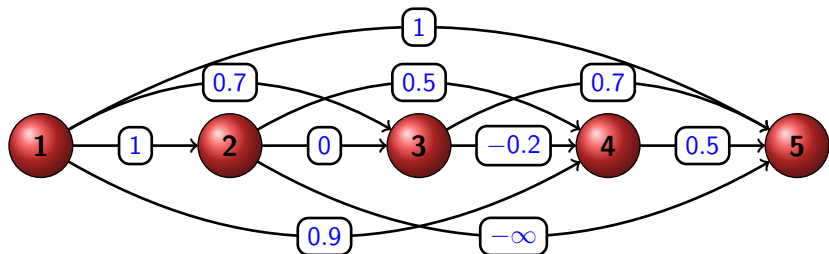
June 20, 2023

Overview

- 1 The last-passage percolation model on complete graphs
- 2 A motivating example: parallel computing with precedence constraints
- 3 Main results
- 4 Some elements of proof

Last-passage percolation on complete graphs

- Set of vertices: $V_n = \{1, \dots, n\}$
- Set of **directed edges**: $E_n = \{(i, j) \mid 1 \leq i < j \leq n\}$
- ν is a probability distribution on $\{-\infty\} \cup \mathbb{R}$
- **Random weight** $X_{i,j}$ on each edge $(i, j) \in E_n$, where $(X_{i,j})_{i < j}$ are i.i.d. with distribution ν



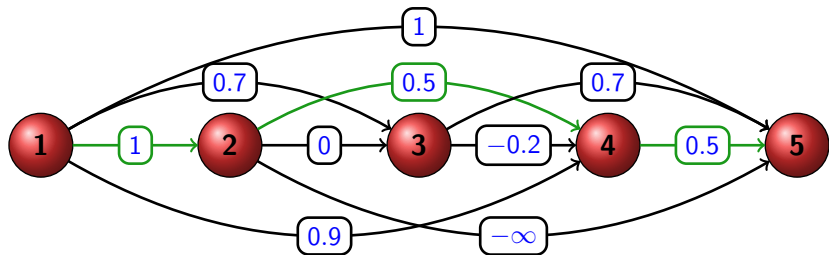
Last-passage percolation on complete graphs

- We consider directed paths in the graph.
- The weight of a path is the sum of the weights of its edges.
- W_n the **maximal weight of paths starting from 1 and ending at n** .

Property

There exists a deterministic constant $C(\nu) \in [0, +\infty]$ called **time constant** such that

$$\frac{W_n}{n} \xrightarrow[n \rightarrow \infty]{a.s., L^1} C(\nu).$$

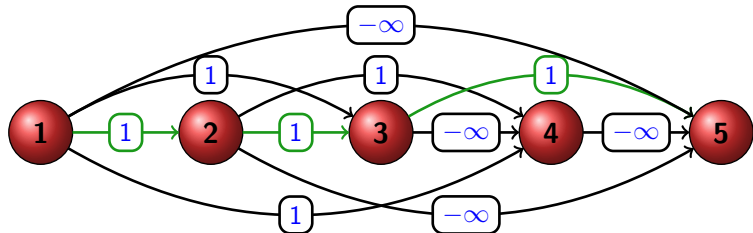


A particular case: the longest path in Barak-Erdős graphs

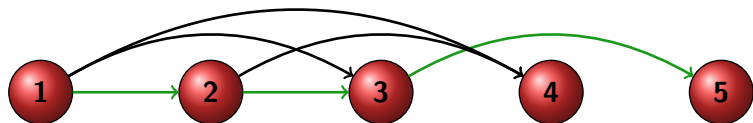
Remark

In terms of a heaviest path, having $X_{i,j} = -\infty$ is equivalent to removing the edge (i,j) from the graph.

Heaviest path



Longest path



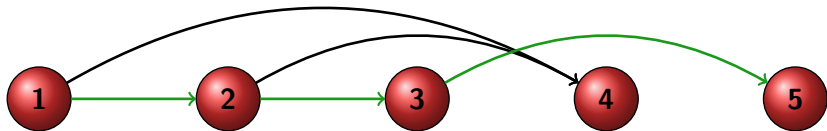
A motivating example for Barak-Erdős graphs

Processing time for parallel computing with precedence constraints

- n tasks to process, infinite number of processors
- processing time 1 for each task
- **Precedence constraints** that must be satisfied during processing given by a **task graph**: task i must be processed before task $j \iff (i, j)$ is in the task graph.

A motivating example for Barak-Erdős graphs

Example of task graph with $n = 5$ vertices:

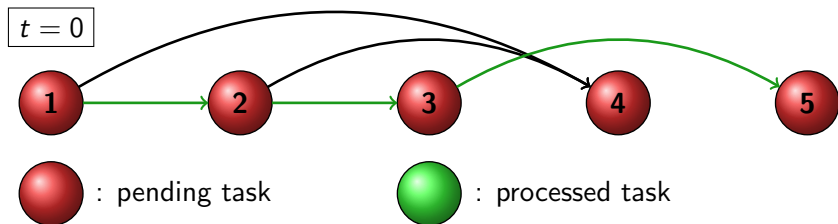


Processing time = number of vertices in a path of maximal length

What if we consider a random task graph?

A motivating example for Barak-Erdős graphs

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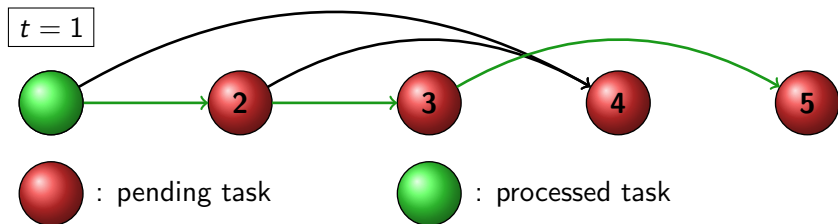


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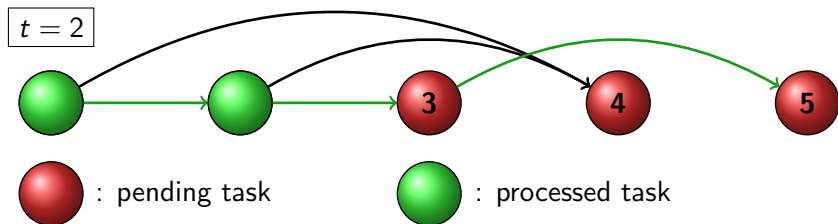


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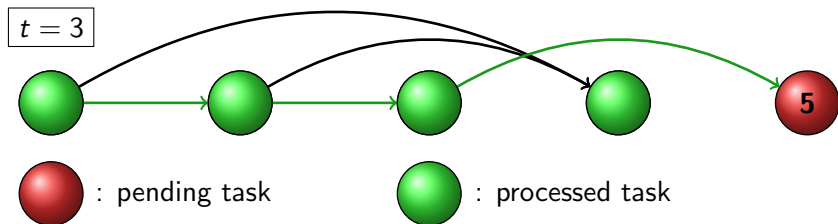


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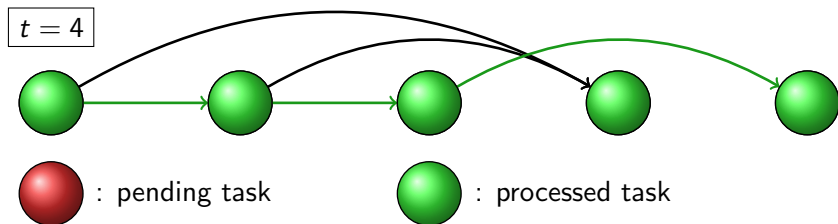


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A motivating example for Barak-Erdős graphs

Example of task graph with $n = 5$ vertices:



Processing time = number of vertices in a path of maximal length

What if we consider a random task graph?

Previous results on Barak-Erdős graphs

Let ν_p be the distribution of a random variable equal to:

- 1 with probability p ,
- $-\infty$ with probability $1 - p$.

For i.i.d. weights $(X_{i,j})_{i < j}$ with distribution ν_p ,

$$\frac{W_n}{n} \xrightarrow[n \rightarrow \infty]{\text{a.s., } L^1} C(\nu_p).$$

Theorem (Mallein, Ramassamy; 19', 21')

- $p \mapsto C(\nu_p)$ is analytic on $(0, 1]$.
- The Taylor expansion of $C(\nu_p)$ at $p = 1$ has integer coefficients.
- $C(\nu_p) = ep + \frac{\pi^2 ep}{2 \log(p)^2} (1 + o(1))$ as $p \rightarrow 0$.

Main results

- Fix k real numbers $-\infty \leq a_k < \cdots < a_1$
- For any positive real numbers (p_1, \dots, p_k) such that $p_1 + \cdots + p_k = 1$, consider the probability distribution

$$\nu_{(p_1, \dots, p_k)} := \sum_{i=1}^k p_i \delta_{a_i}.$$

Theorem (T. 23')

The map $(p_1, \dots, p_k) \mapsto C(\nu_{p_1, \dots, p_k})$ is **analytic** on the set $\{(p_1, \dots, p_k) \in [0, 1]^k \mid p_1 + \cdots + p_k = 1, p_1 > 0\}$.

Main results

- For ν a probability distribution with upper-bounded support, set $M_\nu := \inf\{t \in \mathbb{R} \mid \nu([t, +\infty]) = 0\}$ the **essential supremum** of ν .
- For two probability distributions ν and ν' , set

$$d(\nu, \nu') = \max(d_{LP}(\nu, \nu'), |M_\nu - M_{\nu'}|),$$

where d_{LP} is the Lévy-Prokhorov metric.

Theorem (T. 23')

$\nu \mapsto C(\nu)$ is **continuous** for the metric d on the set of probability measures ν with upper-bounded support.

Main results

Let $\nu_{p,m}$ be the distribution of a random variable equal to:

- 1 with probability p ,
- m with probability $1 - p$.

Theorem (T. 23')

For any real number $m > 0$, $p \mapsto C(\nu_{p,m})$ is a **rational function** on $[0, 1]$.

Main results

- Consider ν_1 and ν_2 two probability distributions.
- ν_1 is **stochastically dominated** by ν_2 when for all $t \in \mathbb{R}$,

$$\nu_1((t, +\infty)) \leq \nu_2((t, +\infty)).$$

Theorem (T. 23')

$\nu \mapsto C(\nu)$ is **strictly increasing** for the stochastic order on the set of distributions with positive finite essential supremum.

Elements of proof

Let μ be a probability measure on $[-\infty, 1)$.

Let $\nu_{p,\mu} = p\delta_1 + (1-p)\mu$ be the distribution of a random variable equal to:

- 1 with probability p ,
- a random variable with distribution μ with probability $1-p$.

Theorem (T. 23')

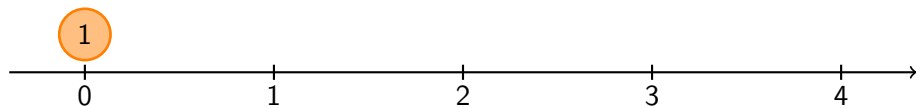
$p \mapsto C(\nu_{p,\mu})$ is **analytic** on $(0, 1]$.

Elements of proof: coupling with a particle system

- Coupling last-passage percolation with a particle system called the **max-growth system** (Foss, Konstantopoulos, Mallein, Ramassamy, 23')
- We construct the graph iteratively adding one vertex at the time:

Maximal weight of a path starting at 1 and ending at n
=
Position of the new particle at time n

Elements of proof: coupling with a particle system

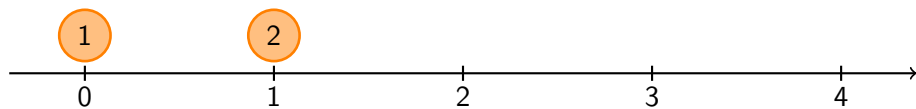
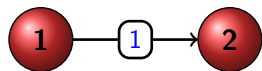


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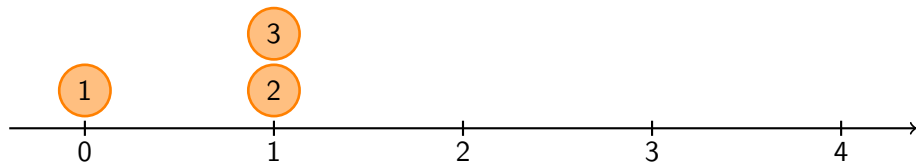
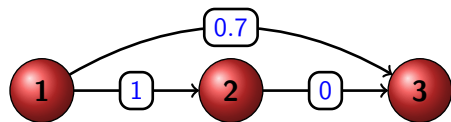


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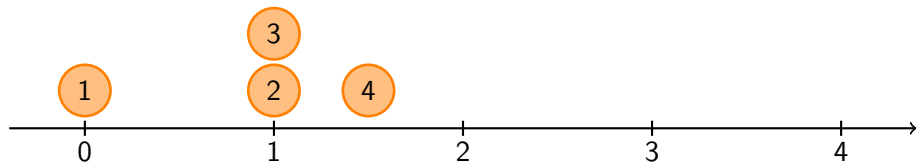
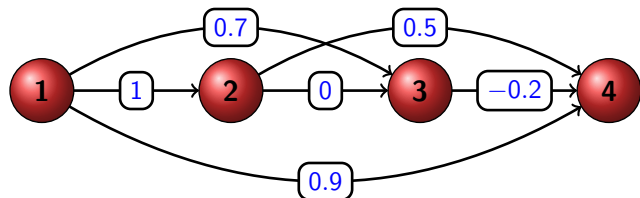


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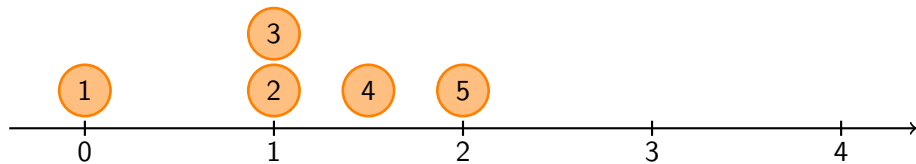
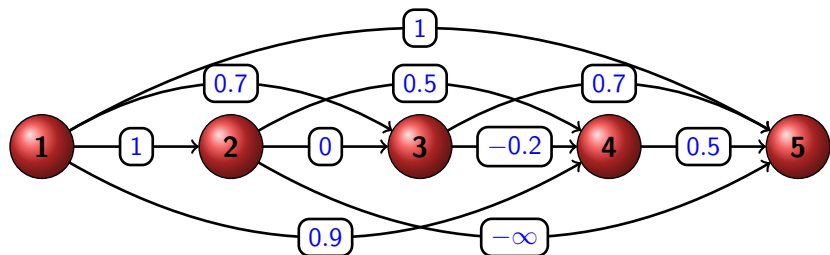


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Elements of proof: coupling with a particle system



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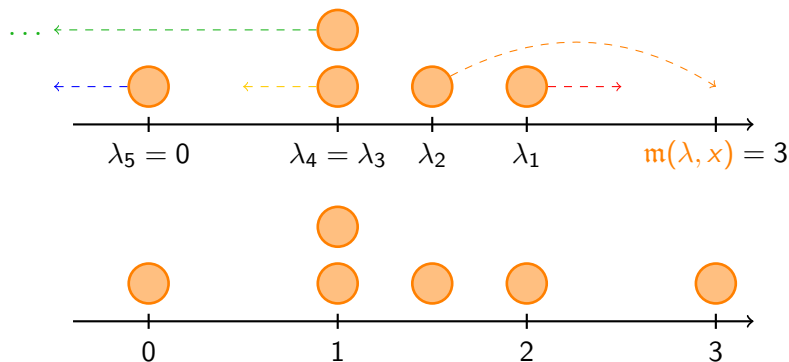
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Position of the new particle at time n

Elements of proof: dynamics of the Max Growth System

- A particle configuration: $\lambda = \sum_{i=1}^N \delta_{\lambda_i}$ where $\lambda_N \leq \dots \leq \lambda_1$
- A sequence of weights $X = (X_1, X_2, \dots, X_N) \in (\{-\infty\} \cup \mathbb{R})^N$

Illustration of the dynamics with $N = 5$ and $X = (0.5, 1.5, -0.5, -\infty, -1)$.



Position of the new particle: $m(\lambda, X) := \max_{1 \leq i \leq N} (\lambda_i + X_i)$.

Dynamics of the Max Growth System

- Consider $(X_i^{(n)})_{i,n \in \mathbb{N}^*}$ i.i.d. random variables with distribution ν .
- We start at time zero with a single particle at time zero: $\lambda^{(0)} = \delta_0$.
- For all $n \in \mathbb{N}^*$, we obtain the configuration at time n from the configuration at time $n - 1$ by using the sequence of i.i.d. weights $X^{(n)} = (X_i^{(n)})_{1 \leq i \leq n}$

Elements of proof: renovation events

- Assume that the support of ν is upper-bounded by 1.
- Notice that if $X_i^{(n)} = 1$, the position of the n -th particle does not depend on $(X_j^{(n)})_{j>i}$.

$$X^{(1)} = (1)$$

$$X^{(2)} = (-1, 1)$$

$$X^{(3)} = (-5, 0.5, 1)$$

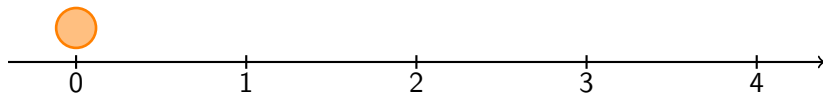
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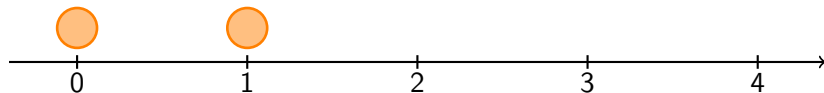
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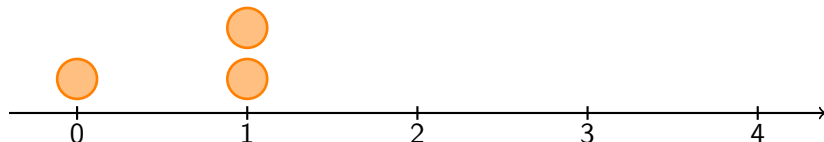
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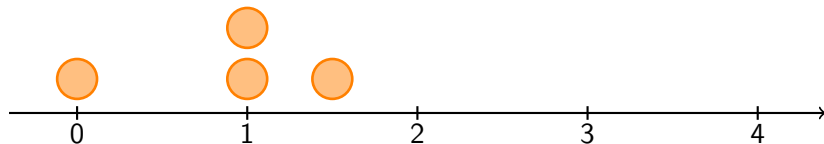
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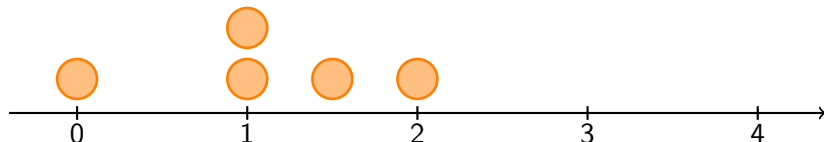
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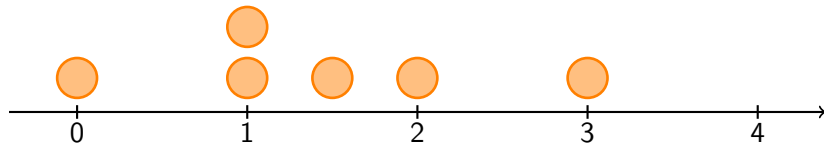
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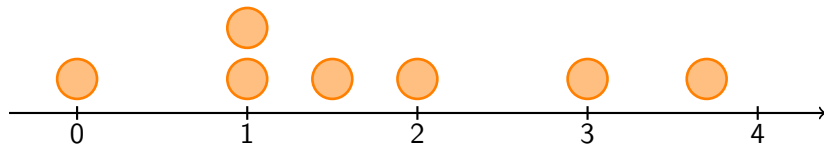
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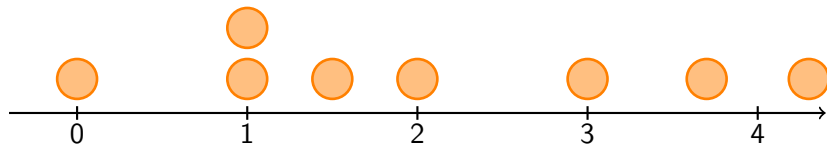
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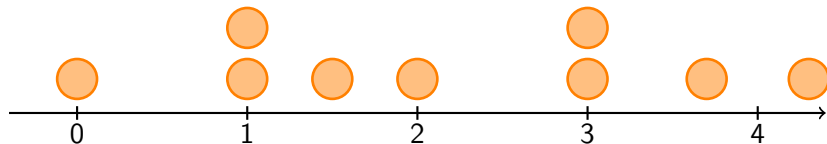
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Elements of proof: renovation events

- A **renovation time** is a time n at which the positions of all the future new particles in the system do not depend on the positions of the old particles.
- Sufficient condition for n to be a renovation time: there is at least a 1 in the first $k + 1$ weights of the sequence at each time $n + k$ for all $k \geq 0$

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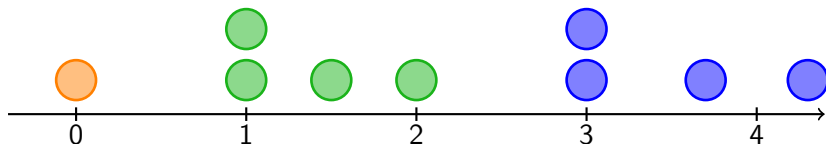
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Elements of proof: renovation events

- Set $\mathcal{T} = \{T_1 < T_2 < \dots\} \subseteq \mathbb{N}^*$ the set of all renovation times.
- By ergodicity:

$$\lim_{N \rightarrow \infty} \frac{W_n}{N} = C(\nu) = \frac{\mathbb{E}[W_{T_1, T_2}]}{\mathbb{E}[T_2 - T_1]}$$

- To prove the analyticity result, it suffices to prove that $\mathbb{E}[r^{T_2 - T_1}]$ is finite for some $r > 1$.

Open questions

- **Barak-Erdős case:** For $\nu = p\delta_1 + (1-p)\delta_{-\infty}$, the Taylor expansion of $C(\nu)$ in $q = 1-p$ at $q = 0$ is
(On-line Encyclopedia of Integer Sequences: A321309)

$$1 - q + q^2 - 3q^3 - 7q^4 + 15q^5 - 29q^6 + 54q^7 - 102q^8 \dots$$

Is there a combinatorial interpretation of those coefficients ?

- Same question for $\nu = p\delta_1 + (1-p)\delta_{-k}$ with $k \in \mathbb{N}^*$? Can we get its asymptotics at $p = 0$?

Thank you!