

Deviation inequalities and moderate deviation principle for Bifurcating Markov Chains.

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joint work with
(H. Djellout and A. Guillin)

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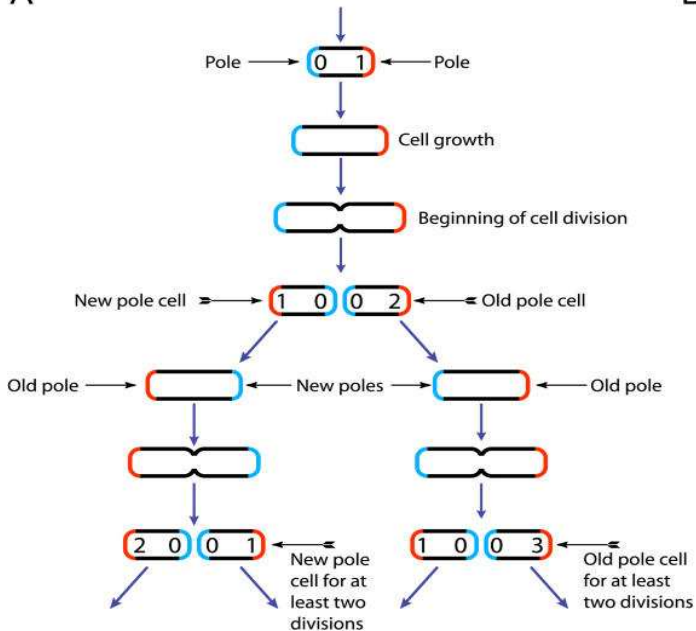
Colloque Jeunes Probabilistes Statisticiens, CIRM 2012

Outline

- 1 Introduction**
 - Motivation
 - The model of bifurcating Markov chain
- 2 Goals
- 3 Framework for the results
- 4 Results
- 5 Application

- Guyon, J. *Limit theorems for bifurcating markov chains. Application to the detection of cellular aging.* Ann. Appl. Probab., (2007), Vol. 17, No. 5-6, pp. 1538-1569.
- Escherichia Coli (E.Coli)

A



B

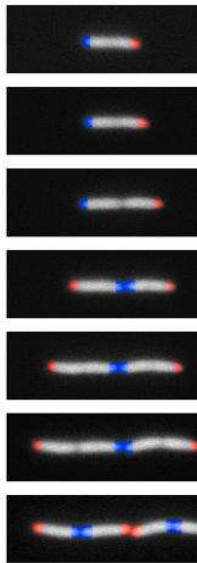
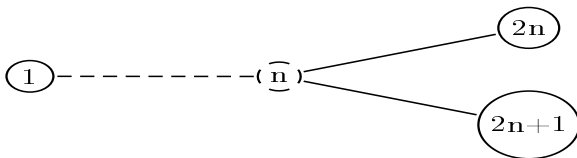


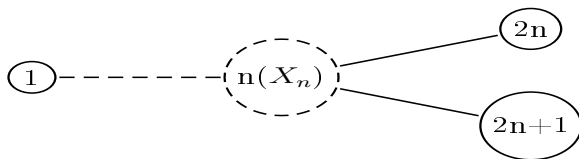
Figure: Cell division from E. J. Stewart and al.

- Guyon, J. Bize, A. Paul, G. Stewart, E.J. Delmas, J.F. Taddéi, F. *Statistical study of cellular aging*. CEMRACS 2004 Proceedings, ESAIM Proceedings, (2005), 14, pp. 100-114.

First order bifurcating autoregressive process BAR(1)



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$$\mathcal{L}(X_1) = \nu, \quad \text{and} \quad \forall n \geq 1, \quad \begin{cases} X_{2n} = \alpha_0 X_n + \beta_0 + \varepsilon_{2n} \\ X_{2n+1} = \alpha_1 X_n + \beta_1 + \varepsilon_{2n+1}, \end{cases} \quad (1)$$

First order bifurcating autoregressive process BAR(1)

where ν is a distribution probability on \mathbb{R} , $\alpha_0, \alpha_1 \in (-1, 1)$; $\beta_0, \beta_1 \in \mathbb{R}$ and $((\varepsilon_{2n}, \varepsilon_{2n+1}), n \geq 1)$ forms a sequence of centered i.i.d bivariate random variables with covariance matrix

$$\Gamma = \sigma^2 \begin{pmatrix} 1 & \rho \\ \rho & 1 \end{pmatrix}, \quad \sigma^2 > 0, \quad \rho \in (-1, 1).$$

$\theta = (\alpha_0, \beta_0, \alpha_1, \beta_1), \sigma$ and ρ .

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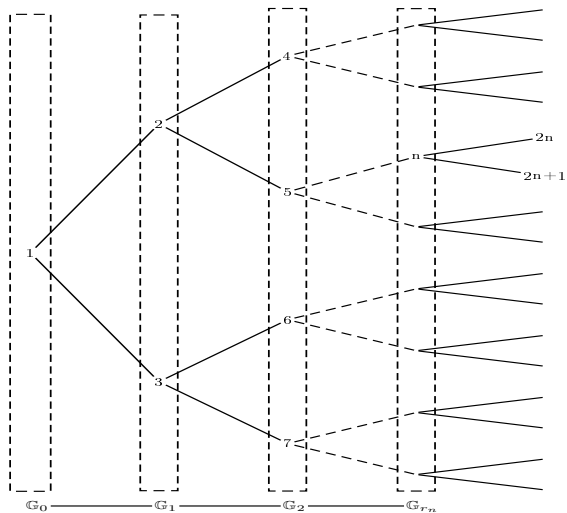
$H_0 = \{(\alpha_0, \beta_0) = (\alpha_1, \beta_1)\}$ against $H_1 = \{(\alpha_0, \beta_0) \neq (\alpha_1, \beta_1)\}$.

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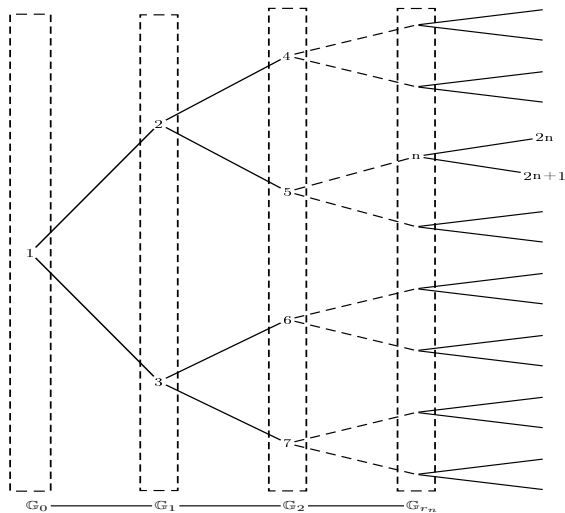
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The binary tree \mathbb{T}



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$$\mathbb{G}_q = \{2^q, 2^q + 1, \dots, 2^{q+1} - 1\}, \mathbb{T}_r = \bigcup_{q=0}^r \mathbb{G}_q, r_n = \lceil \log_2 n \rceil.$$



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- $P : \mathcal{S} \times \mathcal{S}^2 \rightarrow [0, 1]$ such that
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- $P_0(\cdot, \cdot) = P(\cdot, \cdot \times S)$, $P_1(\cdot, \cdot) = P(\cdot, S \times \cdot)$ and $Q = \frac{P_0 + P_1}{2}$.

The model of bifurcating Markov chain

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- For all $f \in \mathcal{B}(S^3)$,

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- $(X_n, n \in \mathbb{T}) : (\Omega, \mathcal{F}, (\mathcal{F}_r, r \in \mathbb{N}), \mathbb{P}) \rightarrow (\mathcal{S}, \mathcal{S})$

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- $(X_n, n \in \mathbb{T}) : (\Omega, \mathcal{F}, (\mathcal{F}_r, r \in \mathbb{N}), \mathbb{P}) \rightarrow (\mathcal{S}, \mathcal{S})$
- - (a) X_n is \mathcal{F}_{r_n} -measurable for all $n \in \mathbb{T}$,
 - (b) $\mathcal{L}(X_1) = \nu$,
 - (c) for all $r \in \mathbb{N}$ and for all family $(f_n, n \in \mathbb{G}_r) \subseteq \mathcal{B}_b(\mathcal{S}^3)$

$$\mathbb{E} \left[\prod_{n \in \mathbb{G}_r} f_n(X_n, X_{2n}, X_{2n+1}) / \mathcal{F}_r \right] = \prod_{n \in \mathbb{G}_r} P f_n(X_n).$$

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- 2 Goals**
- 3 Framework for the results
- 4 Results
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For all $i \in \mathbb{T}$, set $\Delta_i = (X_i, X_{2i}, X_{2i+1})$.

$$\bar{M}_{\mathbb{T}_r}(f) = \frac{1}{|\mathbb{T}_r|} \sum_{i \in \mathbb{T}_r} f(X_i) \quad \text{if } f \in \mathcal{B}(\mathcal{S})$$

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Non asymptotic behavior for $\bar{M}_{\mathbb{T}_r}(f)$ ($f \in \mathcal{B}(S)$ or $\mathcal{B}(S^3)$)

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Non asymptotic behavior for $\bar{M}_{\mathbb{T}_r}(f)$ ($f \in \mathcal{B}(\mathcal{S})$ or $\mathcal{B}(\mathcal{S}^3)$)

A moderate deviation principle for $\frac{M_{\mathbb{T}_r}(f)}{b_{|\mathbb{T}_r|}}$ (for $f \in \mathcal{B}(\mathcal{S}^3)$ such that $Pf = 0$) where

$$M_{\mathbb{T}_r}(f) = \sum_{i \in \mathbb{T}_r} f(\Delta_i) \quad \text{if } f \in \mathcal{B}(\mathcal{S}^3)$$

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- 1 Introduction
- 2 Goals
- 3 Framework for the results**
 - Functional space
 - Hypothesis
- 4 Results
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We will work with the subspace F of $\mathcal{B}(S)$ which verifies

- (i) F contains the constants,
- (ii) $F^2 \subset F$,
- (iii) $F \otimes F \subset L^1(P(x, \cdot))$ for all $x \in S$, and $P(F \otimes F) \subset F$,
- (iv) there exists a probability μ on (S, \mathcal{S}) such that $F \subset L^1(\mu)$ and $\lim_{r \rightarrow \infty} Q^r f(x) = (\mu, f)$ for all $x \in S$ and $f \in F$,
- (v) for all $f \in F$, there exists $g \in F$ such that for all $r \in \mathbb{N}$,
 $|Q^r f| \leq g$,
- (vi) $F \subset L^1(\nu)$

Two cases for the results

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(H1) Geometric ergodicity of Q : $\forall f \in F$ such that $(\mu, f) = 0$,
 $\exists g \in F$ such that $\forall r \in \mathbb{N}$ and $\forall x \in S$, $|Q^r f(x)| \leq \alpha^r g(x)$ for
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some $\alpha \in (0, 1)$.

(H2) Uniform geometric ergodicity of Q : $\exists c > 0$ such that

$$|Q^r f(x)| \leq c\alpha^r \quad \text{for some } \alpha \in (0, 1) \text{ and for all } x \in S,$$

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- S.Valère. Bitseki Penda, Hacène. Djellout and Arnaud. Guillin. *Deviation inequalities, Moderate deviations and some limit theorems for bifurcating Markov chains with application.* arXiv:1111.7303 (2011)

Theorem (Deviation inequalities I)

Let $f \in F$ such that $(\mu, f) = 0$. We assume hypothesis **(H1)**.
Then for all $r \in \mathbb{N}$

$$\mathbb{P}\left(|\overline{M}_{T_r}(f)| > \delta\right) \leq \begin{cases} \frac{c'}{\delta^4} \left(\frac{1}{4}\right)^{r+1} & \text{if } \alpha^2 < \frac{1}{2} \\ \frac{c'}{\delta^4} r^2 \left(\frac{1}{4}\right)^{r+1} & \text{if } \alpha^2 = \frac{1}{2} \\ \frac{c'}{\delta^4} \alpha^{4r+4} & \text{if } \alpha^2 > \frac{1}{2} \end{cases} \quad (3)$$

where the positive constant c' depends on α and f .

When f depends on the mother-daughters triangle (Δ_i)

Theorem (Deviation inequalities II)

We assume that **(H1)** is fulfilled. Let $f \in \mathcal{B}(S^3)$ such that Pf and Pf^2 exists and belong to F and $(\mu, Pf) = 0$. Then for all $\delta > 0$ and all $r \in \mathbb{N}$

$$\mathbb{P}\left(|\overline{M}_{T_r}(f)| > \delta\right) \leq \begin{cases} \frac{c'}{\delta^2} \left(\frac{1}{2}\right)^{r+1} & \text{if } \alpha^2 < \frac{1}{2}; \\ \frac{c'}{\delta^2} r \left(\frac{1}{2}\right)^{r+1} & \text{if } \alpha^2 = \frac{1}{2}; \\ \frac{c'}{\delta^2} \alpha^{2(r+1)} & \text{if } \alpha^2 > \frac{1}{2}, \end{cases} \quad (4)$$

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where the positive constant c' depends on f and α .

If $Pf = 0$,

$$\mathbb{P}\left(|\overline{M}_{\mathbb{T}_r}(f)| > \delta\right) \leq \frac{c'}{\delta^4} \left(\frac{1}{4}\right)^{r+1} \quad (5)$$

Ideas for the proofs

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Markov inequality

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+ control of second and fourth order moment of $\overline{M}_{\mathbb{T}_r}(f)$ using **(H1)** and hypothesis (i)-(vi) on F .

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Markov inequality

+ control of second and fourth order moment of $\overline{M}_{\mathbb{T}_r}(f)$ using **(H1)** and hypothesis (i)-(vi) on F .

- Notice that the dichotomy around the value $\alpha^2 = \frac{1}{2}$ naturally appears in the calculus.

Under the stronger assumption of uniform geometric ergodicity of Q , we have the following more sharp estimations for the above empirical mean.

Theorem: Exponential probability inequalities

Let $f \in \mathcal{B}_b(\mathcal{S})$ such that $(\mu, f) = 0$. Assume that **(H2)** is satisfied. Then for all $\delta > 0$ we have

$$\mathbb{P}\left(\overline{M}_{\mathbb{T}_r}(f) > \delta\right) \leq \begin{cases} \exp(c''\delta) \exp(-c'\delta^2|\mathbb{T}_r|), \forall r \in \mathbb{N}, \text{ if } \alpha < \frac{1}{2} \\ \exp(2c'\delta(r+1)) \exp(-c'\delta^2|\mathbb{T}_r|), \forall r \in \mathbb{N}, \text{ if } \alpha = \frac{1}{2} \\ \exp(-c'\delta^2|\mathbb{T}_r|), \forall r > \frac{\log(\delta/c_0)}{\log \alpha}, \text{ if } \frac{1}{2} < \alpha < \frac{\sqrt{2}}{2} \\ \exp\left(-c'\delta^2 \frac{|\mathbb{T}_r|}{r+1}\right), \forall r > \frac{\log(c_0/\delta)}{\log \sqrt{2}}, \text{ if } \alpha = \frac{\sqrt{2}}{2}, \\ \exp\left(-c'\delta^2 \left(\frac{1}{\alpha^2}\right)^{r+1}\right), \forall r > \frac{\log(\delta/c_0)}{\log \alpha}, \text{ if } \alpha > \frac{\sqrt{2}}{2} \end{cases} \quad (6)$$

where c_0 , c' and c'' depend on α , $\|f\|_\infty$ and c .

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where c_0 , c' and c'' depend on α , $\|f\|_\infty$ and c .

If $f \in \mathcal{B}_b(\mathcal{S}^3)$ such that $(\mu, Pf) = 0$ then we have the same conclusions for $\overline{M}_{\mathbb{T}_r}(f)$.

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Chernoff inequality

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Chernoff inequality + successive conditioning and successive applications of Azuma-Bennet-Hoeffding using **(H2)**.

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- Once again, notice that the dichotomy around $\alpha = \frac{1}{2}$ and $\alpha^2 = \frac{1}{2}$ in (6) naturally appears from the calculations.

Moderate deviation principle for $M_{\mathbb{T}_r}(f)$

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Let (b_n) be an increasing sequence of positive real numbers such that

- (I) $\frac{b_n}{\sqrt{n}} \rightarrow +\infty$,
- (II) if $\alpha^2 < \frac{1}{2}$, the sequence (b_n) is such that $\frac{b_n}{n} \rightarrow 0$,
- (III) if $\alpha^2 = \frac{1}{2}$, the sequence (b_n) is such that $\frac{b_n \log n}{n} \rightarrow 0$,
- (IV) if $\alpha^2 > \frac{1}{2}$, the sequence (b_n) is such that $\frac{b_n \alpha^{r_{n+1}}}{\sqrt{n}} \rightarrow 0$.

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- The conditions (II)-(IV) come from deviation inequalities (6).

Moderate deviation principle for $M_{\mathbb{T}_r}(f)$

Let $f \in \mathcal{B}_b(\mathbb{S}^3)$ such that $Pf = 0$. $(Z_r) = \left(\frac{1}{b_{|\mathbb{T}_r|}} M_{\mathbb{T}_r}(f) \right)$

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Theorem (Moderate deviation principle)

Assume that **(H2)** is satisfied. Let (b_n) be a sequence of real numbers satisfying the above assumptions (I)-(IV), then (Z_r) satisfies a MDP in \mathbb{R} with the speed $\frac{b_{|\mathbb{T}_r|}^2}{|\mathbb{T}_r|}$ and rate function

$$I(x) = \frac{x^2}{2(\mu, Pf^2)}.$$

Particularly, for all $\delta > 0$, we have

$$\lim_{r \rightarrow \infty} \frac{|\mathbb{T}_r|}{b_{|\mathbb{T}_r|}^2} \log \mathbb{P} \left(\frac{1}{b_{|\mathbb{T}_r|}} |M_{|\mathbb{T}_r|}(f)| > \delta \right) = -I(\delta). \quad (7)$$

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$$\mathbb{P} \left(\frac{1}{b_{\mathbb{T}_r}} |M_{\mathbb{T}_r}(f)| \geq \delta \right) \sim \exp \left(-\frac{b_{\mathbb{T}_r}^2}{\mathbb{T}_r} \frac{\delta^2}{2(\mu, Pf^2)} \right) \quad (7')$$

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We consider the BAR(1) process

$$\mathcal{L}(X_1) = \nu, \text{ and } \forall n \geq 1, \begin{cases} X_{2n} = \alpha_0 X_n + \beta_0 + \varepsilon_{2n} \\ X_{2n+1} = \alpha_1 X_n + \beta_1 + \varepsilon_{2n+1}, \end{cases}$$

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The noise values in a compact set.

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The least square estimator $\hat{\theta}^r = (\hat{\alpha}_0^r, \hat{\beta}_0^r, \hat{\alpha}_1^r, \hat{\beta}_1^r)$ of

$\theta = (\alpha_0, \beta_0, \alpha_1, \beta_1)$ is given by, for $\eta \in \{0, 1\}$

$$\begin{cases} \hat{\alpha}_\eta^r = \frac{|\mathbb{T}_r|^{-1} \sum_{i \in \mathbb{T}_r} X_i X_{2i+\eta} - \left(|\mathbb{T}_r|^{-1} \sum_{i \in \mathbb{T}_r} X_i \right) \left(|\mathbb{T}_r|^{-1} \sum_{i \in \mathbb{T}_r} X_{2i+\eta} \right)}{|\mathbb{T}_r|^{-1} \sum_{i \in \mathbb{T}_r} X_i^2 - \left(|\mathbb{T}_r|^{-1} \sum_{i \in \mathbb{T}_r} X_i \right)^2} \\ \hat{\beta}_\eta^r = |\mathbb{T}_r|^{-1} \sum_{i \in \mathbb{T}_r} X_{2i+\eta} - \hat{\alpha}_\eta^r |\mathbb{T}_r|^{-1} \sum_{i \in \mathbb{T}_r} X_i. \end{cases} \quad (8)$$

$\mu_1 : \Theta \rightarrow \mathbb{R}$ and $\mu_2 : \Theta \times \mathbb{R}_+^* \rightarrow \mathbb{R}$ by writing

$$(\mu, \mathbf{x}) = \mu_1(\theta) \quad \text{and} \quad (\mu, \mathbf{x}^2) = \mu_2(\theta, \sigma^2), \quad (9)$$

where $\theta = (\alpha_0, \beta_0, \alpha_1, \beta_1) \in \Theta = (-1, 1) \times \mathbb{R} \times (-1, 1) \times \mathbb{R}$, and μ is the stationary distribution of Q .

Theorem

$\forall \delta > 0$ and $\forall \gamma < \min \left(\frac{c_1 b}{1+\delta}, \frac{c_1 b}{1+\sqrt{\delta}}, \frac{c_1 b}{1+\sqrt[4]{\delta}} \right)$, where $c_1 = c_1(\mu_1, \mu_2)$
 we have $\mathbb{P} \left(\left\| \hat{\theta}^r - \theta \right\| > \delta \right) \leq$

$$\left\{ \begin{array}{l} \exp (c''(\gamma \delta)^{1-p/2}) \exp (-c'(\gamma \delta)^{2-p}|\mathbb{T}_r|), \forall r \in \mathbb{N}, \text{ if } \alpha < \frac{1}{2} \\ \exp (c'(\gamma \delta(r+1))^{1-p/2}) \exp (-c'(\gamma \delta)^{2-p}|\mathbb{T}_r|), \forall r \in \mathbb{N}, \text{ if } \alpha = \frac{1}{2} \\ \exp (-c'(\gamma \delta)^{2-p}|\mathbb{T}_r|), \forall r > \frac{\log((\gamma \delta)^{1-p/2}/c_0)}{\log \alpha}, \text{ if } \frac{1}{2} < \alpha < \frac{\sqrt{2}}{2} \\ \exp \left(-c'(\gamma \delta)^{2-p} \frac{|\mathbb{T}_r|}{r+1} \right), \forall r > \frac{\log(c_0/(\gamma \delta)^{1-p/2})}{\log \sqrt{2}}, \text{ if } \alpha = \frac{\sqrt{2}}{2} \\ \exp \left(-c'(\gamma \delta)^{2-p} \left(\frac{1}{\alpha^2} \right)^{r+1} \right), \forall r > \frac{\log((\gamma \delta)^{1-p/2}/c_0)}{\log \alpha}, \text{ if } \alpha > \frac{\sqrt{2}}{2}, \end{array} \right. \quad (10)$$

where c' and c'' depend on α , $\|f\|_\infty$ and c , c_0 depends on α , $\|f\|_\infty$, c and γ , and $p \in \{0, 1, 3/2\}$.

End.