

An Lower Estimate for the Number of Level Crossing of a Random Algebraic Curve when the Coefficients follow Semi-Stable Distribution

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Abstract- Let $\xi_1(\omega), \xi_2(\omega), \xi_3(\omega) \dots \dots \dots \xi_n(\omega)$ be a sequence of mutually independent, identically distributed random variables following semi-stable distribution with characteristic function $\exp(-C + \cos \log |t|) |t|^\alpha$, $C > 1, 1 < \alpha < 2$. In this work, we obtain the lower bound of the number of real zeros of the random algebraic equation $\sum_{r=0}^n \xi_r(\omega)x^r$. $N_n(\omega)$ denote the number of real roots must $> \left(\frac{\mu \log n}{\log \log n}\right)$ for all $n > n_0$, except for a set of measure at most $\frac{\mu' \log \log n}{\log n}$.

Index Terms- Random variables, Joint distribution, Characteristic function, Semi-stable distribution, Multiple roots, Random algebraic equations.

I. INTRODUCTION

Let $N_n(\omega)$ be the number of real zeros of the n th degree polynomial $\sum_{r=0}^n \xi_r x^r$ where the coefficients ξ_r 's are identically distributed independent random variables. Ibragimov and Maslova first studied the average number of real roots when the coefficients belong to the domain of attraction of normal law and established that $EN_n \sim \frac{1}{\pi} \log n$. Samal and Mishra [6], [7], [8] have studied the lower bound for the number of real roots of the equation $\sum_{r=0}^n \xi_r x^r = 0$, where the coefficients ξ_r 's are symmetric stable random variables with infinite variance i.e, with characteristic function $\exp(-C |t|^\alpha), C > 0$ and $1 < \alpha < 2$. In this paper we have studied the lower bound for the number of real roots where ξ_r 's follow semi-stable distribution. Thus we have extended the results of Samal and Mishra to wider class of random variables. Thus our result is a generalization of earlier works in this respect

Theorem -1

Let $f(x) = \sum_{r=0}^n \xi_r(\omega)x^r$ be an algebraic polynomial, where ξ_r 's are random variables following semi-stable distribution with common characteristics function $\varphi(t) = \exp(-C + \cos \log |t|) |t|^\alpha$ for constant $C > 1, 1 < \alpha < 2$. Then there exist $n_0 \in N$ such that for $n > n_0$, N_n the number of level crossing satisfy of the random algebraic polynomial $f(x) = 0$. $N_n(\omega) > \left(\frac{\mu \log n}{\log \log n}\right)$ for all $n > n_0$, out side a set of measure at most $\frac{\mu' \log \log n}{\log n}$.

II. PRELIMINARIES

We need the following lemmas for the proof of the theorem. Choose constants A and B such that

$$0 < B < 1, A > 1 \quad \text{and} \quad \text{let } \lambda = \log n$$

$$\text{Let } M \text{ be the integer determined by } M = \left\lceil \frac{2^{\alpha+1} A e \lambda^\alpha}{B} \right\rceil + 1$$

$$\text{Let } k \text{ be an integer determined by } M^{2k} \leq n \leq M^{2k+2} \tag{2.2.1}$$

Since the number of roots in $(-1,1)$ is twice in $[0,1]$ it will be four times in $(-\infty, \infty)$ as transformation $x = \frac{1}{t}$ will reduce $(0, \infty)$ to $[0,1]$. There fore it will be sufficient to consider the points x_m given by

$$x_m = \left(1 - \frac{1}{M^{2m}}\right)^{\frac{1}{\alpha}} \tag{2.2.2}$$

Where $m \in S = \left\{ \left[\frac{k}{2} \right] + 1, \left[\frac{k}{2} \right] + 2, \left[\frac{k}{2} \right] + 3, \dots \dots k \right\}$

There are $\left[\frac{k}{2} \right]$ points if k is even and $\left(\frac{k+1}{2} \right)$ points if k is odd.

Clearly when $m \rightarrow \infty, k \rightarrow \infty$ and $x_m \rightarrow 1$ and

When $m \rightarrow 0 (i.e k \rightarrow 0), x_m \rightarrow 0$. So that x_m covers all the points in $[0,1]$.

Ste $f(x_m) = A_m + R_m$, where $A_m = \sum_1 \xi_r(\omega) x_m^r = \sum_{r=M^{2m-1}+1}^{M^{2m+1}} \xi_r(\omega) x_m^r$

And $R_m = \sum_2 \xi_r(\omega) x_m^r + \sum_3 \xi_r(\omega) x_m^r$

$R_m = \sum_{r=0}^{M^{2m-1}} \xi_r(\omega) x_m^r + \sum_{r=M^{2m+1}+1}^n \xi_r(\omega) x_m^r$

r ranging from $M^{2m-1} + 1$ to M^{2m+1} in $\sum_1 \xi_r(\omega) x_m^r$,

from 0 to M^{2m-1} in $\sum_2 \xi_r(\omega) x_m^r$ and from $M^{2m+1} + 1$ to n in $\sum_3 \xi_r(\omega) x_m^r$ respectively.

Obviously A_m, A_{m+1} are independent random variables. We defined normalizing constants V_m by

$$V_m^\alpha = \sum_1 x_m^{\alpha r} = \sum_{r=M^{2m-1}+1}^{M^{2m+1}} x_m^{\alpha r}.$$

Lemma 2.1

If $\varphi(t)$ is characteristic function of a random variables $\xi(\omega)$ which follow semi-normal distribution, then

$$P(\omega: |\xi(\omega)| > \epsilon) \leq \frac{c+1}{(\alpha+1)\epsilon^\alpha}$$

Proof of the lemma

By Loeve [3], we have

$$P\{\omega: |\xi(\omega)| > \epsilon\} < 7\epsilon \int_0^{\frac{1}{\epsilon}} (1 - \varphi(t)) dt = 7\epsilon(c+1) \int_0^{\frac{1}{\epsilon}} t^\alpha dt$$

$$\therefore P\{\omega: |\xi(\omega)| > \epsilon\} \leq \frac{7(c+1)}{(\alpha+1)\epsilon^\alpha}$$

Lemma 2.2

$|\sum_{r=M^{2m+1}+1}^n \xi_r(\omega) x_m^r| < \frac{1}{2} V_m$, except for a set of measure at most $\frac{2^\alpha}{\alpha+1} \frac{Ae}{B} (c+1) e^{-M}$

Proof of the lemma

The characteristic function of $\sum_3 \xi_r(\omega) x_m^r$ is $\exp\{-(c + \cos \log |t|) |t|^\alpha \cdot \sum_{r=M^{2m+1}+1}^n x_m^{\alpha r}\}$ for $1 < \alpha < 2$

Hence by lemma (2.1), we have

$$\begin{aligned} P_1 &= P \left\{ \left| \sum_{r=M^{2m+1}+1}^n \xi_r(\omega) x_m^r \right| \geq \frac{1}{2} V_m \right\} \\ &\leq \frac{(c+1)}{(\alpha+1) \left(\frac{V_m}{2}\right)^\alpha} \cdot \sum_{r=M^{2m+1}+1}^n x_m^{\alpha r} \end{aligned} \tag{2.2.3}$$

$$\therefore \sum_{r=M^{2m+1}+1}^n x_m^{\alpha r} < M^{2m} e^{-M}$$

$$\Rightarrow P_1 = P \left\{ \left| \sum_{r=M^{2m+1}+1}^n \xi_r(\omega) x_m^r \right| \geq \frac{1}{2} V_m \right\} < \frac{(c+1)2^\alpha}{\alpha+1} \frac{Ae}{B} e^{-M}$$

Lemma 2.3

$$|\sum_{r=0}^{M^{2m-1}} \xi_r(\omega) x_m^r| < \lambda (\sum_{r=0}^{M^{2m-1}} x_m^{ar})^{\frac{1}{\alpha}},$$

except for a set of measure at most $\frac{(C+1)}{(\alpha+1)} \frac{1}{\lambda^\alpha}$.

Proof of the lemma

The characteristic function of $\sum_{r=0}^{M^{2m-1}} \xi_r(\omega) x_m^r$ is $\exp\{-(c + \cos \log |t|) |t|^\alpha \cdot \sum_{r=0}^{M^{2m-1}} x_m^{ar}\}$ for $1 < \alpha < 2$

$$\text{Let } P_2 = P \left\{ \left| \sum_{r=0}^{M^{2m-1}} \xi_r(\omega) x_m^r \right| \geq \lambda (\sum_{r=0}^{M^{2m-1}} x_m^{ar})^{\frac{1}{\alpha}} \right\}$$

Let $D_m^\alpha = \sum_{r=0}^{M^{2m-1}} x_m^{ar}$ and the characteristic function of $\sum_{r=0}^{M^{2m-1}} \xi_r(\omega) x_m^r$ be $\varphi(t)$

$$\text{So } P_2 = P \left\{ \left| \sum_{r=0}^{M^{2m-1}} \xi_r(\omega) x_m^r \right| \geq \lambda (\sum_{r=0}^{M^{2m-1}} x_m^{ar})^{\frac{1}{\alpha}} \right\} \leq \frac{(C+1)}{(\alpha+1)} \frac{1}{\lambda^\alpha}$$

$$\text{So } P_2 \leq \frac{(C+1)}{(\alpha+1)} \frac{1}{\lambda^\alpha}$$

Now by using Lemmas (2.2) and (2.3), we have for any given m

$$|R_m| < \frac{1}{2} V_m + \lambda D_m, \text{ except for a set of measure at most}$$

$$\frac{(C+1)2^\alpha A e}{\alpha+1} \frac{1}{B} e^{-M} + \frac{(C+1)}{(\alpha+1)} \frac{1}{\lambda^\alpha}, \text{ but } \lambda (\sum_{r=0}^{M^{2m-1}} x_m^{ar})^{\frac{1}{\alpha}} < \lambda V_m \left(\frac{2Ae}{MB} \right)^{\frac{1}{\alpha}}$$

Hence $R_m < \frac{1}{2} V_m + \frac{1}{2} V_m = V_m$, for $m \in S$ except for a set of measure less than equal to

$$= \frac{(C+1)}{(\alpha+1)} \left(2^\alpha \frac{Ae}{B} e^{-M} + \frac{1}{\lambda^\alpha} \right) \tag{2.2.4}$$

We define events E_m and F_m by

$$E_m = \{\omega: A_{2m}(\omega) > V_{2m}, A_{2m+1}(\omega) < -V_{2m+1}\}, \quad F_m = \{\omega: A_{2m}(\omega) < -V_{2m}, A_{2m+1}(\omega) > V_{2m+1}\}$$

Lemma 2.4

Union of events E_m and F_m defined by

$$E_m = \{\omega: A_{2m}(\omega) > V_{2m}\} \cap \{\omega: A_{2m+1}(\omega) < -V_{2m+1}\}$$

$$F_m = \{\omega: A_{2m}(\omega) < -V_{2m}\} \cap \{\omega: A_{2m+1}(\omega) > V_{2m+1}\}$$

occur with positive probability.

Proof:

Let $g_m(u)$ and $G_m(x)$ denote respectively characteristic function and distribution function of the random variables $\left(\frac{A_m(\omega)}{V_m}\right)$. Then

$$g_m(u) = \exp \left\{ -(c + \cos \log |t|) |t|^\alpha \frac{1}{V_m^\alpha} \sum_{r=0}^{M^{2m+1}} x_m^{ar} L \left(x_m^r \frac{u}{V_m} \right) \right\}.$$

obviously $V_m \rightarrow \infty$ as $m \rightarrow \infty$ and as such

$$x_m^\alpha \frac{u}{V_m} \rightarrow 0 \text{ and } x_m^r \frac{\theta}{V_m} \rightarrow 0 \text{ as } m \rightarrow \infty.$$

We use lemma 6.3 here taking

$$\lim_{u \rightarrow 0} C(x) = C \text{ and get}$$

$$\frac{L\left(\left|\frac{x_m^r t}{V_m}\right|\right)}{L\left(\left|\frac{x_m^r \theta}{V_m}\right|\right)} = \frac{c\left(\left|\frac{x_m^r t}{V_m}\right|\right) \exp\left\{-\int_a^{\left(\left|\frac{x_m^r t}{V_m}\right|\right)} -(c + \cos \log \left|\frac{\epsilon(u)}{u}\right|) \left|\frac{\epsilon(u)}{u}\right|^\alpha du\right\}}{c\left(\left|\frac{x_m^r \theta}{V_m}\right|\right) \exp\left\{-\int_a^{\left(\left|\frac{x_m^r \theta}{V_m}\right|\right)} -(c + \cos \log \left|\frac{\epsilon(u)}{u}\right|) \left|\frac{\epsilon(u)}{u}\right|^\alpha du\right\}}$$

Since $\lim_{u \rightarrow 0} C(x) = C$

and

since $\frac{x_m^r t}{V_m} \rightarrow 0$ and $\frac{x_m^r \theta}{V_m} \rightarrow 0$ as $m \rightarrow \infty$, we have

$$C\left(\left|\frac{x_m^r t}{V_m}\right|\right) = (1 + o(1))C \text{ as } m \rightarrow \infty,$$

And

$$C\left(\left|\frac{x_m^r \theta}{V_m}\right|\right) = (1 + o(1))C \text{ as } m \rightarrow \infty.$$

Also since $\lim_{u \rightarrow 0} \epsilon(u) = 0$, $\epsilon > 0$ there exists $t_o > 0$ such that for $u < t_o$, $|\epsilon(u)| < \epsilon$

so, $\epsilon(u) = (1 + o(1))\epsilon$ as $u \rightarrow 0$.

we have

$$\begin{aligned} \frac{L\left(\left|\frac{x_m^r t}{V_m}\right|\right)}{L\left(\left|\frac{x_m^r \theta}{V_m}\right|\right)} &= \frac{c\left(\left|\frac{x_m^r t}{V_m}\right|\right) \exp\left\{-\int_a^{\left(\left|\frac{x_m^r t}{V_m}\right|\right)} -(c + \cos \log \left|\frac{\epsilon(u)}{u}\right|) \left|\frac{\epsilon(u)}{u}\right|^\alpha du\right\}}{c\left(\left|\frac{x_m^r \theta}{V_m}\right|\right) \exp\left\{-\int_a^{\left(\left|\frac{x_m^r \theta}{V_m}\right|\right)} -(c + \cos \log \left|\frac{\epsilon(u)}{u}\right|) \left|\frac{\epsilon(u)}{u}\right|^\alpha du\right\}} \\ &= \frac{c\left(\left|\frac{x_m^r t}{V_m}\right|\right)}{c\left(\left|\frac{x_m^r \theta}{V_m}\right|\right)} \exp\left\{-\int_{\left(\left|\frac{x_m^r \theta}{V_m}\right|\right)}^{\left(\left|\frac{x_m^r t}{V_m}\right|\right)} -(c + \cos \log \left|\frac{\epsilon(u)}{u}\right|) \left|\frac{\epsilon(u)}{u}\right|^\alpha du\right\} \\ &= \frac{(1+o(1))C}{(1+o(1))C} \exp\left\{-\int_{\left(\left|\frac{x_m^r \theta}{V_m}\right|\right)}^{\left(\left|\frac{x_m^r t}{V_m}\right|\right)} \frac{\epsilon}{u} (1 + o(1)) du\right\} \text{ as } u \rightarrow 0 \\ &= \epsilon \exp\left\{-\int_{\left(\left|\frac{x_m^r \theta}{V_m}\right|\right)}^{\left(\left|\frac{x_m^r t}{V_m}\right|\right)} \frac{1}{u} du\right\} (1 + o(1)) \\ &= (1 + o(1)) e^{\log \left|\frac{\theta}{t}\right|^\epsilon} = \left|\frac{\theta}{t}\right|^{o(1)} \text{ as } u \rightarrow 0 \text{ making } \epsilon \rightarrow 0 \end{aligned}$$

which implies

$$L\left(\left|\frac{x_m^r t}{V_m}\right|\right) = \left|\frac{\theta}{t}\right|^{o(1)} L\left(\left|\frac{x_m^r \theta}{V_m}\right|\right) \text{ as } u \rightarrow 0.$$

Hence

$$g_m(u) \rightarrow \exp\left\{-\left(c + \cos \log |u|\right) |u|^\alpha \frac{1}{V_m^\alpha} \left|\frac{\theta}{t}\right|^{o(1)} \sum_{M^{2m-1}+1}^{M^{2m+1}} x_m^{\alpha r} L\left(\frac{x_m^r \theta}{V_m}\right)\right\}$$

$$= \exp\left\{-\left(c + \cos\log|u|\right)|u|^{\alpha-o(1)}\theta^{o(1)}\frac{1}{V_m^\alpha}\sum_{M^{2m-1}+1}^{M^{2m+1}}x_m^{\alpha r}L\left(\frac{x_m^r\theta}{V_m}\right)\right\}$$

$$= \exp\left\{-\left(c + \cos\log|u|\right)|u|^{\alpha-o(1)}\theta^{o(1)}\right\}$$

(by definition of V_m).

So as $m \rightarrow \infty$, $g_m(u) \rightarrow \exp\{-\left(c + \cos\log|u|\right)|u|^\alpha\}$ in any bounded interval of u – values. Since $\exp\{-|u|^\alpha\}$ is continuous for all u , $G_m(x)$ will converge to a distribution $F(x)$ (say) corresponding to the characteristic function $g_m(u)$ (cf Gnedenko and Kolmagorov [17] p.83).

so

$$\sup_x |G_m(x) - F(x)| = o(1) \tag{2.2.5}$$

So for $\epsilon > 0$, $|G_{2m}(-1) - F(-1)| < \epsilon$

$$\Rightarrow F(-1) - \epsilon < G_{2m}(-1) < F(-1) + \epsilon$$

and $|G_{2m+1}(-1) - F(-1)| < \epsilon$

implies $F(-1) - \epsilon < G_{2m+1}(-1) < F(-1) + \epsilon$.

Then $P(A_{2m} < -V_{2m}) = P\left(\frac{A_{2m}}{V_{2m}} < -1\right)$

$G_{2m}(-1) > F(-1) - \epsilon$ as $m \rightarrow \infty$

and similarly

$P(A_{2m+1} < -V_{2m+1}) > F(-1) - \epsilon$

also

$|G_{2m}(1) - F(1)| < \epsilon$

$$\Rightarrow F(1) - \epsilon < G_{2m}(1) < F(1) + \epsilon$$

$-(F(1) - \epsilon) > -G_{2m}(1)$

$1 - (F(1) - \epsilon) > 1 - G_{2m}(1)$

or

$$1 - G_{2m}(1) < 1 - F(1) + \epsilon \tag{2.2.6}$$

So,

$P(A_{2m} > V_{2m}) = 1 - P(A_{2m} < V_{2m})$.

$= 1 - G_{2m}(1)$

$> 1 - F(1) - \epsilon \tag{2.2.7}$

Similarly $P(A_{2m+1} > V_{2m+1}) > 1 - F(1) - \epsilon. \tag{2.2.8}$

Hence it follows from above that

$P(E_m \cup F_m) > 2(F(-1) - \epsilon)(1 - F(-1) - \epsilon)$ for $\epsilon > 0$.

Hence as $m \rightarrow \infty$, $P(E_m \cup F_m) \geq 2F(-1)(1 - F(-1)) > 0$.

Hence the proof.

Lemma 2.5

Out side a set of measure at most $\frac{c'}{k}$ the event $(E_m \cup F_m)$ occurs at least for C_k values of m . i.e out side a set of measure $\frac{c'}{k}$. $A_{2m} > V_{2m}$ and $A_{2m+1} < -V_{2m+1}$.

Proof of the lemma

$(E_m \cup F_m)$ occurs with positive probability. (by Lemma 2.4)

Suppose $P(E_m \cup F_m) = \zeta_m > 0$

Then there exist an absolute constant $\eta > 0$, such that $P(E_m \cup F_m) = \zeta_m > \zeta$.

Let $\zeta = \sum_S \zeta_m$, where summation \sum_S means that the summation is taken over all's pairs.

Applying Chebyshev's inequality, we have for $0 < \epsilon < \zeta_m$.

$P(|\zeta - E(\zeta)| \geq S\epsilon) \leq \frac{\mu_1}{k}$, there fore outside a set of measure $\frac{\mu_1}{k}$.

$|\zeta - E(\zeta)| < S\epsilon$, there fore, it follows that out side a set of measure $\frac{\mu_1}{k}$, $E_m \cup F_m$ occurs i.e either $A_{2m} > V_{2m}$ or $A_{2m+1} < -V_{2m+1}$

III. MAIN RESULTS

Proof of the theorem

Define $\xi_m = \begin{cases} 0 & \text{if } |R_{2m}| < V_{2m} \text{ and } |R_{2m+1}| < V_{2m+1} \\ 1 & \text{otherwise} \end{cases}$

We have $f(x_{2m}) = A_{2m} + R_{2m}$ and $f(x_{2m+1}) = A_{2m+1} + R_{2m+1}$

$\zeta_m - \zeta_m \xi_m = 1$ only if $\zeta_m = 1$ and $\xi_m = 0$ which implies the occurrence of one of the events.

$$E(\sum_{m=m_0}^k \zeta_m \xi_m) \leq (k+1) \left\{ C_1 e^{-M} + \left(\frac{C+1}{\alpha+1} \right) \frac{1}{\lambda^\alpha} \right\}$$

Hence for $1 < \beta < 2$

$$P\left(\sum_{m=m_0}^k \zeta_m \xi_m > (k+1) \left(C_1 e^{-M} + \left(\frac{C+1}{\alpha+1} \right) \frac{1}{\lambda^\alpha} \right) \lambda^\beta\right) < \frac{C_4 k}{\lambda^{\alpha-\beta}} \tag{3.3.1}$$

except for a set of measure $\leq \frac{1}{\lambda^\beta}$.

From (2.2.1), it follows that $M^{2k} \leq n \leq M^{2k+2}$

After some calculation

$$\mu_1 \frac{\log n}{\log M} \leq k \leq \mu_2 \frac{\log n}{\log M}, \text{ we have } M = \left\lceil \frac{2^{\alpha+1} A e \lambda^\alpha}{B} \right\rceil + 1$$

It follows that

$$\mu_5 \frac{\log n}{\log \log n} \leq k \leq \mu_6 \frac{\log n}{\log \log n} \tag{3.3.2}$$

Earlier we have to shown that

$$\sum_{m=m_0}^k \zeta_m > \mu' k \text{ out side a set of measure } \frac{\mu'}{k}.$$

Now $N_n > \sum_{m=m_0}^k (\zeta_m - \zeta_m \xi_m) > \mu k$, out side a set of measure at most.

$$\text{but } \frac{\mu'}{k} + \frac{1}{\lambda^\beta} < \mu' \frac{\log n}{\log \log n} + \frac{1}{(\log n)^\beta} \leq \mu' \frac{\log n}{\log \log n}, \text{ since } (\log n)^\beta > \frac{\log n}{\log \log n} \text{ as } (1 < \beta < 2)$$

$$\text{Hence } N_n > \mu \frac{\log n}{\log \log n}$$

Similarly

$$\text{except for a set of measure at most } \mu' \frac{\log \log n}{\log n}.$$

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