



## EXPECTED NUMBER OF LEVEL CROSSINGS OF A RANDOM ORTHOGONAL POLYNOMIAL

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### ABSTRACT

This paper provides number of zeros of a class of orthogonal polynomial which is a sequence of mutually independent, normally distributed random variables with mean zero and variance unity then the average number of zeros of the random orthogonal polynomial is asymptotic to  $\sqrt{n/3}$ .

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### INTRODUCTION

Let  $y = \sum_{k=0}^n y_k(w) b_k \Phi_k(t)$  be random polynomial such that  $\{y_k(w)\}_{k=0}^n$  is a sequence of mutually independent, normally distributed random variables with mean zero and variance unity and  $\{\Phi_k(t)\}_{k=0}^n$  is a sequence of classical Gegenbauer polynomials such that  $\{b_k \Phi_k(t)\}_{k=0}^n$  is a sequence of normalized orthogonal polynomials. Then, it is proved that the average number of zeros of the random polynomial is asymptotic to  $\sqrt{n/3}$ .

Let  $y = \sum_{k=0}^n b_k y_k(w) \Phi_k(t)$ ,  $0 < w < 1$  be random polynomial, where  $\{y_0(w), y_1(w), \dots, y_n(w)\}$  is a sequence of mutually independent, normally distributed random variables with mathematical expectation zero and variance one. Let  $\{\Phi_0(t), \dots, \Phi_1(t)\}$  is a sequence of real valued polynomials (functions) and  $(b_0, b_1, \dots)$  is a sequence of real constants. J.E. Littlewoods and A.C. Offord<sup>[4,5,6]</sup> showed that, when  $b_k=1$  and  $\Phi_k(t)=t^k$ , most of the equation of the form  $y=0$ , have at most  $25(\log n)^2$  real zeros for large  $n$ . When  $b_0=0$ ,  $b_k=1$  for  $k \neq 0$ , and  $\Phi_0(t) = \cos k(\cos^{-1}t)$ , J.E.A., Dunnage<sup>[3]</sup> estimated the average number of zeros of the family of equations  $y=0$  to be asymptotic to  $2n/\sqrt{3}$  in the interval  $(-1,1)$ .

It is interesting to observe that while  $t^k$ 's are a set of functions monotonic in  $(-\infty, 0)$  and  $[0, \infty]$ ,  $\cos k(\cos^{-1}t)$ , for each  $k$ , oscillates  $k$  times between  $-1$  and  $1$ . The fact that the average number of zeros of  $y=0$  when  $\Phi_k(t) = \cos k(\cos^{-1}t)$  is proportional to the number of individual oscillations of  $\Phi_k(t)$  about the  $t$ -axis, draws attention to the equation as to how far the oscillatory nature of  $\Phi_k(t)$  decisively affects the zeros of  $y=0$ . Although the answer remains still inconclusive, we attempt to show that for large  $n$ , the above equation may be expected to have c.n., ( $C > 0$ ) number of real roots when  $\Phi_k(t)$  happens to be the ultra spherical classical orthogonal polynomial (Gegenbauer polynomial). In other words the oscillatory property of  $\Phi_k(t)$  is also shared by  $\sum_{k=0}^n b_k y_k(w) \Phi_k(t)$

Now  $\Phi_k(t)$  is associated with a weight function  $u(t) = (1-t^2)^{-1/2}$ ,  $\lambda > 1/2$  corresponding to the interval  $(-1,1)$  over which the integral of  $u(t)$ ,  $\Phi_k(t)$  is a positive number  $h_k$ . We take  $b_k = h^{-1/2}_k$ . Then the integral of  $\Psi^2_k(t) = b^2_k \Phi^2_k(t)$  over the given interval is unity, so that each of the terms of the polynomial  $\sum_{k=0}^n b_k y_k(w) \Psi_k(t) = \sum_{k=0}^n b_k y_k(w) \Phi_k(t)$  has same weightage in the same sense.

Thus, in what follows, we find the average number of zeros of the equation

$$\sum_{k=0}^n b_k y_k(w) \Psi_k(t) \quad (1.1)$$

We denote by  $EN_n(f; \alpha, \beta)$  the expected number of real zeros of (1.1) in the interval  $(\alpha, \beta)$ . Das<sup>[2]</sup> was first to find  $EN_n(f; \alpha, \beta)$  for a random orthogonal polynomial, although  $\Psi_k(t)$

considered by him was a normalized Legendre polynomial, which is a special case of the polynomial considered by us.

**1.2. Formula for ENn (f: a,b)** Following the procedure of Kac,<sup>[7]</sup> we obtain

$$EN_n(f: a, b) = \frac{1}{\pi} \int_a^b \frac{(\{X_n(t)\}\{Z_n(t)\} - \{Y_n(t)\}^2)^{\frac{1}{2}}}{X_n(t)} dt \quad (1.2)$$

$$\text{where } X_n(t) = X \sum_{k=0}^n \Psi_k(t)^2$$

$$Y_n(t) = Y \sum_{k=0}^n [\Psi'_k(t)][\Psi'_k(t)]$$

$$Z_n(t) = Z \sum_{k=0}^n [\Psi'_k(t)]^2$$

provided that  $X_n Z_n - Y_n^2 > 0$ .

The last inequality holds good by Cauchy's inequality.

Let us put  $\mu_n = I_n h_n^{-1} r^{-1}_{n+1}$  where  $r_n$  is the coefficient of  $t^n$  in  $\Phi_n(t)$ . The famous Crammer and Leadbetter<sup>[1]</sup> formula of the theory of orthogonal functions reads as follows.

$$\sum_{k=0}^n h_k^{-1} \Phi_k(\mu) \Phi_k(t) = \mu_n \frac{\{\Phi_{n+1}(\mu) \Phi_n(t) - \Phi_n(\mu) \Phi_{n+1}(t)\}}{\mu - t} \quad (1.3)$$

Putting  $\mu = t + \gamma$  in the formula (1.3), we obtain by Taylor's expansion that

$$\sum_{k=0}^n h_k^{-1} \Phi_k(\mu) \Phi_k(t + \gamma) = \mu_n \frac{\{\Phi_{n+1}(\mu) \Phi_{n+1}(t) - \Phi_n(\mu) \Phi_{n+1}(t + \gamma) \Phi_{n+1}(t)\}}{t + \gamma - t}$$

Now

$$\begin{aligned} LHS (=) & \sum_{k=0}^n h_k^{-1} (t) \Phi_k(t) + \gamma \Phi'_k(t) + \frac{\gamma^2}{2!} \Phi''_k(t) + \dots \\ & = \sum_{k=0}^n h_k^{-1} \Phi_k^2(t) + \gamma \sum_{k=0}^n h_k^{-1} \Phi_k(t) + \Phi'_k(t) + \frac{\gamma^2}{2!} \sum_{k=0}^n h_k^{-1} \Phi_k(t) + \Phi''_k(t) + \dots \\ RHS_n (=) & \mu \left[ \Phi_n(t) \left\{ \Phi_{n+1}(t) + \gamma \Phi'_{n+1}(t) + \gamma \Phi'_k(t) + \frac{\gamma^2}{2!} \Phi''_{n+1}(t) + \frac{\gamma^3}{3!} \Phi'''_{n+1}(t) \right\} \right] \\ & - \Phi_{n+1}(t) \left\{ \Phi_n(t) + \gamma \Phi'_n(t) + \frac{\gamma^2}{2!} \Phi''_n(t) + \frac{\gamma^3}{3!} \Phi'''_n(t) \right\} \\ & = \mu \left\{ \Phi_n(t) + \Phi'_{n+1}(t) + \frac{1}{2} \Phi_n(t) + \Phi''_{n+1}(t) \gamma + \frac{1}{6} \Phi_n(t) \Phi''_{n+1}(t) \gamma^2 \right\} \\ & - \Phi_{n+1}(t) \Phi'_n(t) - \frac{1}{2} \Phi_{n+1}(t) \Phi''_n(t) \gamma - \frac{1}{6} \Phi_{n+1}(t) \Phi'''_n(t) \gamma^2 \dots \end{aligned}$$

Now, equating coefficients of like powers of  $\gamma$  on both the sides, we obtain

$$\sum_{k=0}^n h_k^{-1} [\Phi_k(t)] = \mu_n [\Phi'_{n+1}(t)\Phi_n(t) - \Phi_{n+1}(t)\Phi'_n(t)] \quad (1.4)$$

$$\sum_{k=0}^n h_k^{-1} [\Phi_k(t)\Phi'_k(t)] = \frac{\mu_n}{2} [\Phi''_{n+1}(t)\Phi_n(t) - \Phi_{n+1}(t)\Phi''_n(t)] \quad (1.5)$$

$$\sum_{k=0}^n h_k^{-1} [\Phi_k(t)\Phi'_k(t)] = \frac{\mu_n}{3} [\Phi'''_{n+1}(t)\Phi_n(t) - \Phi_{n+1}(t)\Phi'''_n(t)] \quad (1.6)$$

Differentiating (1.5), we get

$$\sum_{k=0}^n h_k^{-1} [\Phi'_k(t) + \Phi''_k(t)] = \frac{\mu_n}{2} [\Phi'''_{n+1}(t)\Phi_n(t) + \Phi'_n(t)\Phi''_{n+1}(t) - \Phi'_{n+1}(t)\Phi''_n(t) - \Phi''_n(t)\Phi_{n+1}(t)]$$

$$\sum_{k=0}^n h_k^{-1} [\Phi'_k(t) + \Phi''_k(t)] = \frac{\mu_n}{2} [\Phi'''_{n+1}(t)\Phi_n(t) + \Phi'_n(t)\Phi''_{n+1}(t) - \Phi'_{n+1}(t)\Phi''_n(t) - \Phi''_n(t)\Phi_{n+1}(t)]$$

or

$$\sum_{k=0}^n h_k^{-1} [\Phi'_k(t)]^2 = \frac{\mu_n}{2} [\Phi''_{n+1}(t)\Phi'_n(t) - \Phi'_{n+1}(t)\Phi''_n(t)] \quad \text{Hence, from (1.4), (1.5) and (1.7), it is evident that}$$

$$+ \frac{\mu_n}{6} [\Phi'''_{n+1}(t)\Phi_n(t) - \Phi_{n+1}(t)\Phi'''_n(t)] \quad (1.6)$$

$$\sum_{k=0}^n h_k^{-1} [\Phi_k(t)]^2 = \sum_{k=0}^n h_k^{-1/2} [\Phi_k(t)]^2$$

$$= \sum_{k=0}^n (\Psi_k(t))^2 = X_n$$

and

$$\sum_{k=0}^n h_k^{-1} [\Phi_k(t)]^2 = \sum_{k=0}^n h_k^{-1/2} [\Phi_k(t)]^2$$

$$= \sum_{k=0}^n (\Psi'_k(t))^2 = Y_n$$

and

$$\begin{aligned} \sum_{k=0}^n h_k^{-1} [\Phi_k(t)]^2 &= \sum_{k=0}^n h_k^{-1/2} [\Phi_k(t)]^2 \\ &= \sum_{k=0}^n (\Psi_k(t))^2 = Z_n \end{aligned}$$

Now, making use of (1.4), (1.5) and (1.7), together with the fact that  $\mu_n \neq 0$ , we obtain

$$\frac{X_n Z_n - Y_n^2}{X_n^2} = \frac{Z_n}{X_n} - \left( \frac{Y_n}{X_n} \right)^2$$

$$= \left\{ \frac{\frac{\mu_n}{2} [\Phi_{n+1}''(t)\Phi_n'(t) - \Phi_{n+1}'(t)\Phi_n''(t)] + \frac{\mu_n}{6} [\Phi_{n+1}'''(t)\Phi_n(t) - \Phi_{n+1}(t)\Phi_n'''(t)]}{\mu_n [\Phi_{n+1}'(t)\Phi_n(t) - \Phi_{n+1}(t)\Phi_n'(t)]} \right\}^2 \text{ i.e. } EN_n(f : a, b) = \frac{1}{\pi} \int_a^b g_n(t) dt. \quad (1.8)$$

where

$$g_n^2(t) = \left[ \frac{W_n(t) + V_n(t)}{R_n(t)} - \frac{U_n^2(t)}{4R_n^2(t)} \right],$$

$$R_n(t) = R_n = \Phi_{n+1}'(t)\Phi_n(t) - \Phi_{n+1}(t)\Phi_n'(t),$$

$$U_n(t) = U_n = \Phi_{n+1}''(t)\Phi_n(t) - \Phi_{n+1}(t)\Phi_n''(t),$$

$$V_n(t) = V_n = \frac{1}{2} [\Phi_{n+1}'''(t)\Phi_n'(t) - \Phi_{n+1}'(t)\Phi_n'''(t)]$$

and

$$W_n(t) = W_n = \frac{1}{6} [\Phi_{n+1}''''(t)\Phi_n(t) - \Phi_{n+1}(t)\Phi_n''''(t)]$$

### 1.3. Proof of the theorem

To prove the theorem, we divide the interval  $(-1, 1)$  into three subintervals;

(i)  $(-1, \varepsilon, 1 + \varepsilon)$ ,

(ii)  $(-1, -1 + \varepsilon)$ , (iii).  $(1 - \varepsilon, 1)$ .

We choose  $\varepsilon = n^{-\frac{1}{4+\delta}}$ .

In the above section we find out the average number of zeros in the interval  $(-1, \varepsilon, 1 + \varepsilon)$ .

In the section 1.4 we prove that the number of zeros in the intervals (ii) and (iii) are negligible in comparison to those in the interval (i).

### 1.4. Expected Number of Zeros in The Interval $(-1, \varepsilon, 1 + \varepsilon)$

In order to evaluate  $EN_n(f : -1 + \varepsilon, 1 + \varepsilon)$ , we use the formula derived in 1.2.

From above, we have

$$(1-t^2)\Phi_n''(t) = (2\lambda+1)t\Phi_n'(t) + n(n+2\lambda)\Phi_n(t) \quad (1.9)$$

and

$$(1-t^2)\Phi_{n+1}''(t) - (2\lambda+1)t\Phi_{n+1}'(t) - (n+1)(n+1+2\lambda)\Phi_{n+1}(t) \quad (1.10)$$

Multiplying (1.9) by  $\Phi_{n+1}''(t)$  and (1.10) by  $\Phi_n''(t)$ , we obtain

$$(1-t^2)\Phi_n''(t)\Phi_{n+1}'(t) = (2\lambda+1)t\Phi_n'(t)\Phi_{n+1}'(t) - n(n+2\lambda)\Phi_n'(t)\Phi_{n+1}'(t) \quad (1.11)$$

and

$$(1-t^2)\Phi_{n+1}''(t)\Phi_n'(t) = (2\lambda+1)t\Phi_{n+1}'(t)\Phi_n'(t) - (n+1)(n+1+2\lambda)\Phi_{n+1}'(t)\Phi_n'(t) \quad (1.12)$$

Subtracting (1.11) from (1.12), we have

$$(1-t^2)\Phi_n''(t)\Phi_{n+1}'(t) = (2\lambda+1)t\Phi_n'(t)\Phi_{n+1}'(t) - n(n+2\lambda)\Phi_n'(t)\Phi_{n+1}'(t) \quad (1.13)$$

Multiplying (1.9) by  $\Phi_n''(t)$  and (1.10) by  $\Phi_n''(t)$ , we have

$$(1-t^2)\Phi_n''(t)\Phi_{n+1}'(t) = (2\lambda+1)t\Phi_n'(t)\Phi_{n+1}'(t) - n(n+2\lambda)\Phi_n'(t)\Phi_{n+1}'(t) \quad (1.14)$$

and

$$(1-t^2)\Phi_n''(t)\Phi_{n+1}'(t) = (2\lambda+1)t\Phi_n'(t)\Phi_{n+1}'(t) - n(n+2\lambda)\Phi_n'(t)\Phi_{n+1}'(t) \quad (1.15)$$

Subtracting (1.15) from (1.14), we have

$$\begin{aligned} & (1-t^2)\Phi_{n+1}''(t)\Phi_n'(t) - \Phi_n''(t)\Phi_{n+1}'(t) \\ &= (2\lambda+1)t\left(\Phi_{n+1}'(t)\Phi_n'(t) - \Phi_n'(t)\Phi_{n+1}'(t)\right) \\ & - (2n+1+2\lambda)\Phi_{n+1}'(t)\Phi_n'(t) \quad (1.16) \end{aligned}$$

Differentiating (1.9) and (1.10), we obtain

$$-2t\Phi_n''(t) + (1-t^2)\Phi_n'''(t) = (2\lambda+1)\Phi_n'(t) + (2\lambda+1)t\Phi_n''(t) - n(n+2\lambda)\Phi_n'(t) \quad (1.17)$$

and

$$\begin{aligned}
& -2t\Phi''_n(t) + (1-t^2)\Phi''''_n(t) = (2\lambda+1)\Phi'_{n+1}(t) \\
& = (2\lambda+1)t\Phi'_{n+1}(t) + (2\lambda+1)\Phi''''_{n+1}(t) \\
& - (n+1)(n+12\lambda)\Phi'_{n+1}(t)
\end{aligned} \tag{1.18}$$

Multiplying (1.9) by  $\Phi''_n(t)$  and (1.10) by  $\Phi_n(t)$ , and subtracting, we have

$$\begin{aligned}
& (1-t^2)(\Phi''''_{n+1}(t)\Phi_n(t) - \Phi''''_n(t)\Phi_{n+1}(t)) \\
& = 2t(\Phi'_{n+1}(t)\Phi_n(t) - \Phi_{n+1}(t)\Phi''_n(t)) \\
& + (2\lambda+1)t(\Phi''_{n+1}(t)\Phi_n(t) - \Phi''_n(t)\Phi_{n+1}(t)) \\
& - n(n+2\lambda)(\Phi''_{n+1}(t)\Phi_n(t) - \Phi''_n(t)\Phi_{n+1}(t)) \\
& = (2\lambda+3)t(\Phi''_{n+1}(t)\Phi_n(t) - \Phi''_n(t)\Phi_{n+1}(t)) \\
& + (2\lambda+1) - n(n+2\lambda)(\Phi''_{n+1}(t)\Phi_n(t) - \Phi''_n(t)\Phi_{n+1}(t)) \\
& - (2n+1+2\lambda)(\Phi'_{n+1}(t)\Phi_n(t)) \\
& (2\lambda+3)t \left[ \frac{(2\lambda+1)}{(1-t^2)} (\Phi'_{n+1}(t)\Phi_n(t) - \Phi'_n(t)\Phi_{n+1}(t)) \right] \\
& - \left[ \frac{(2n+1+2\lambda)}{(1-t^2)} (\Phi'_{n+1}(t)\Phi_n(t)) \right] \\
& + \left[ (2\lambda+1) - n(n+2\lambda)(\Phi'_{n+1}(t)\Phi_n(t) - \Phi'_n(t)\Phi_{n+1}(t)) \right] \text{ (we have substituted)} \\
& + \left[ \frac{(2n+1+2\lambda)}{(1-t^2)} (\Phi_n(t)(n+1)t\Phi_{n+1}(t) - (2\lambda+1)\Phi_n(t)) \right] \\
& (t^2-1)(\Phi'_{n+1}(t) = (n+1)t\Phi_{n+1}(t) - (2\lambda+1)\Phi_n(t)) \\
& = \left[ \frac{(2\lambda+3)(2\lambda+1)t^2}{(1-t^2)} + (2\lambda+1) - n(n+2\lambda)(\Phi'_{n+1}(t) - \Phi'_n(t)\Phi_{n+1}(t)) \right] \\
& - \left[ \frac{(2n+1+2\lambda)(2\lambda+3)t}{(1-t^2)} - \frac{t(n+1)(2n+1+2\lambda)}{(1-t^2)} \right] \Phi_{n+1}(t)\Phi_n(t) \\
& - \left[ \frac{(2n+1+2\lambda)(2\lambda+1)\Phi_n^2(t)}{(1-t^2)} \right] \tag{1.19}
\end{aligned}$$

For large  $n$ , we shall use the asymptotic estimate of  $\Phi_n(t)$  as

$$\Phi_n(t) \sim \frac{2^\lambda}{(\pi n)^{1/2}} (1-t)^{1/2} (1+t)^{-1/2} \left[ \cos X + O\left(\frac{1}{n \sin \theta}\right) \right],$$

where  $X = (n\theta + \lambda\theta - \lambda\theta)$  and  $t = \cos\theta$ . (We have taken  $\alpha = \beta = \lambda - \frac{1}{2}$ ) From above, we get

$$(1-t^2)\Phi_n'(t) = (2\lambda - 1 + n)\Phi_{n-1}(t) - nt\Phi_n(t) \quad (1.20)$$

and

$$(1-t^2)\Phi_{n+1}'(t) = (2\lambda + n)\Phi_n(t) - (n+1)t\Phi_{n+1}(t) \quad (1.21)$$

From the two relations, we have

$$\begin{aligned} & (1-t^2)(\Phi_{n+1}'(t)\Phi_n(t) - \Phi_n'(t)\Phi_{n+1}(t)) \\ &= (1-t^2)R_n(t) \\ &= (\lambda+n)\Phi_n^2(t) - \Phi_{n+1}(t)\Phi_n(t) - (2\lambda+1+n)\Phi_{n-1}(t)\Phi_{n+1}(t). \end{aligned}$$

From the two relations, we have

$$\begin{aligned} & (1-t^2)(\Phi_{n+1}'(t)\Phi_n(t) - \Phi_n'(t)\Phi_{n+1}(t)) \\ &= (1-t^2)R_n(t) \quad (1.22) \\ &= (\lambda+n)\Phi_n^2(t) - \Phi_{n+1}(t)\Phi_n(t) - (2\lambda+1+n)\Phi_{n-1}(t)\Phi_{n+1}(t). \end{aligned}$$

Hence

$$\begin{aligned} & (1-t^2)R_n(t) \\ &= (\lambda+n)\frac{2^{2\lambda}}{\pi n}(1-t)^\lambda(1+t)^{-1} \left[ \cos^2 X + O\left(\frac{1}{n \sin n\theta}\right) \right] \\ &= t \frac{2^{2\lambda}}{\pi [n(n+1)]^{1/2}} (1-t)^\lambda(1+t)^{-1} \left[ \cos X \cos(X+\theta) + O\left(\frac{1}{n \sin n\theta}\right) \right] \\ &= (2\lambda-1+n) \frac{2^{2\lambda}}{\pi [n(n+1)]^{1/2}} (1-t)^\lambda(1+t)^{-1} \left[ \cos X \cos(X-\theta) + O\left(\frac{1}{n \sin n\theta}\right) \right] \\ &\sim \sigma^2 \left[ \cos^2 X + O\left(\frac{1}{n \sin n\theta}\right) \right] - \sigma^2 \left[ \cos X \cos(X-\theta) \cos(X+\theta) + O\left(\frac{1}{n \sin n\theta}\right) \right] \\ &= \sigma^2 \left[ \sin^2 \theta + O\left(\frac{1}{n \sin n\theta}\right) \right] \end{aligned}$$

where



$$\sigma = \frac{2^\lambda}{\pi^{1/2}} (1-t)^{-\lambda/2} (1+t)^{-\lambda/2}.$$

Hence

$$(1-t^2)R_n(t) = \frac{2^{2\lambda}}{\pi n} (1-t)^{-\lambda} (1+t)^{-1} \left[ (1-t^2) + O\left(\frac{1}{n \sin n\theta}\right) \right] \quad (1.23)$$

$$R_n(t) = \frac{2^{2\lambda}}{\pi(1-t^2)} (1-t)^{-\lambda} (1+t)^{-1} \left[ (1-t^2) + O\left(\frac{1}{n \sin n\theta}\right) \right]$$

Now

$$\Phi_n(t)\Phi_{n+1}(t) = \frac{2^{2\lambda}}{\pi n} (1-t)^{-\lambda} (1+t)^{-1} \left[ \cos X \cos(X+\theta) + O\left(\frac{1}{n \sin n\theta}\right) \right] \quad (1.24)$$

$$\leq \frac{K}{n} (1-t)^{-\lambda} (1+t)^{-1}$$

and

$$\Phi_n^2(t) \leq \frac{K}{n} (1-t)^{-\lambda} (1+t)^{-1} \quad (1.25)$$

Hence

$$\frac{\Phi_n(t)\Phi_{n+1}(t)}{\Phi'_{n+1}(t)\Phi_n(t) - \Phi'_n(t)\Phi_{n+1}(t)} \leq \frac{\frac{K}{n} (1-t)^{-\lambda} (1+t)^{-1}}{(1-t^2)^{-1} (1+t)^{-\lambda} \left\{ (1-t^2) + O\left(\frac{1}{n \sin n\theta}\right) \right\}}$$

So that

$$\frac{\Phi_{n+1}(t)\Phi_n(t)}{R_n(t)} = O\left(\frac{1}{n}\right) \quad (1.26)$$

and

$$\frac{\Phi_n^2(t)}{R_n(t)} = O\left(\frac{1}{n}\right) \quad (1.27)$$

Now

$$\begin{aligned} & (1-t^2)\Phi'_n(t)\Phi_{n+1}(t) \\ &= (2\lambda-1+n)\Phi'_{n-1}(t)\Phi_{n+1}(t) - n t \Phi_n(t)\Phi_{n+1}(t) \\ &= (2\lambda-1+n) \frac{\sigma^2}{n} \left[ \cos(X-\theta)\cos(X+\theta) + O\left(\frac{1}{n \sin n\theta}\right) \right] \\ & \quad - n t \frac{\sigma^2}{n} \left\{ \cos X + O\left(\frac{1}{n \sin n\theta}\right) \right\} \left\{ \cos(X+\theta) + O\left(\frac{1}{n \sin n\theta}\right) \right\}. \end{aligned}$$

Hence

$$\frac{\Phi'_n(t)\Phi_{n+1}(t)}{R_n(t)} = O\left(\frac{1}{(1-t^2)}\right) \quad (1.28)$$

Now

$$\begin{aligned} \frac{V_n(t)}{R_n(t)} &= \frac{\Phi''_{n+1}(t)\Phi'_n(t) - \Phi'_{n+1}(t)\Phi''_n(t)}{2(\Phi'_{n+1}(t)\Phi_n(t) - \Phi_{n+1}(t)\Phi'_n(t))} \\ &= \frac{\{n(n+2\lambda)(\Phi_n(t)\Phi'_{n+1}(t) - \Phi'_n(t)\Phi_{n+1}(t)) - (2n+1+2\lambda)\Phi_{n+1}(t)\Phi'_n(t)\}}{2(\Phi'_{n+1}(t)\Phi_n(t) - \Phi_{n+1}(t)\Phi'_n(t))} \\ &= \frac{n(n+2\lambda)}{2(1-t^2)} - \frac{\{2n+1+2\lambda\}}{2(1-t^2)} \frac{\{\Phi_{n+1}(t)\Phi'_n(t)\}}{(\Phi'_{n+1}(t)\Phi_n(t) - \Phi_{n+1}(t)\Phi'_n(t))} = \frac{(2\lambda+3)(2\lambda+1)t^2}{6(1-t^2)^2} + \frac{\{1+2\lambda\}}{6(1-t^2)} - \frac{n\{n+2\lambda\}}{6(1-t^2)} \\ &= \frac{n^2}{2(1-t^2)} + O\left(\frac{n}{2(1-t^2)}\right) \quad (1.29) \quad = \frac{(2n+1+2\lambda)(2\lambda+3)}{(1-t^2)^2} + \frac{\Phi_n^2(t)}{6(\Phi'_{n+1}(t)\Phi_n(t) - \Phi_{n+1}(t)\Phi'_n(t))} \\ &= \frac{n^2}{6(1-t^2)} + O\left(\frac{n}{(1-t^2)^2}\right) \quad (1.30), \end{aligned}$$

From (1.28) and (1.29), we have

$$\frac{W_n(t) + V_n(t)}{R_n(t)} = \frac{n^2}{3(1-t^2)} + O\left(\frac{n}{(1-t^2)^2}\right), \quad (1.31)$$

Also we have

$$\begin{aligned} \frac{U_n(t)}{2R_n(t)} &= \frac{\Phi'_{n+1}(t)\Phi_n(t) - \Phi_{n+1}(t)\Phi'_n(t)}{2(\Phi'_{n+1}(t)\Phi_n(t) - \Phi_{n+1}(t)\Phi'_n(t))} \\ &= \frac{\{(2\lambda+1)t(\Phi'_{n+1}(t)\Phi_n(t) - \Phi_{n+1}(t)\Phi'_n(t)) - (2n+1+2\lambda)\Phi_{n+1}(t)\Phi_n(t)\}}{2(1-t^2)(\Phi'_{n+1}(t)\Phi_n(t) - \Phi_{n+1}(t)\Phi'_n(t))} \\ &= \frac{(2\lambda+1)t}{2(1-t^2)^2} - \frac{\{2n+1+2\lambda\}}{2(1-t^2)} \frac{\Phi_{n+1}(t)\Phi_n(t)}{(\Phi'_{n+1}(t)\Phi_n(t) - \Phi_{n+1}(t)\Phi'_n(t))} \\ &= \frac{(2\lambda+3)(2\lambda+1)t^2}{6(1-t^2)^2} + \frac{\{1+2\lambda\}}{6(1-t^2)} - \frac{n\{n+2\lambda\}}{6(1-t^2)} \\ &= \frac{(2n+1+2\lambda)(2\lambda+3)}{(1-t^2)^2} + \frac{\Phi_n^2(t)}{6(\Phi'_{n+1}(t)\Phi_n(t) - \Phi_{n+1}(t)\Phi'_n(t))} \\ &= O\left(\frac{1}{(1-t^2)^2}\right) + O\left(\frac{1}{n(1-t^2)^2}\right). \end{aligned}$$

Hence

$$\frac{W_n(t) + V_n(t)}{R_n(t)} + \frac{U_n(t)}{2R_n(t)} = \frac{n^2}{3(1-t^2)} + O\left(\frac{n}{(1-t^2)^2}\right) \quad (1.32)$$

and

$$\frac{U_n(t)}{4R_n^2(t)} = O\left(\frac{1}{(1-t^2)^2}\right) \quad (1.33)$$

so that

$$g_n(t) = \sqrt{\frac{W_n(t)+V_n(t)}{R_n(t)} - \frac{U_n^2(t)}{4R_n^2(t)}} = \frac{n}{\sqrt{3}(1-t^2)^{1/2}} + O\left(\frac{1}{(1-t^2)^2}\right)^{1/2}$$

For the range  $(-1+\epsilon, 1-\epsilon)$ , we notice that

$$1-t^2 > 2\epsilon - \epsilon^2 = \frac{2n^{\frac{1}{4+\delta}} - 1}{n^{\frac{2}{4+\delta}}}, \text{ where } \epsilon = n^{\frac{1}{4+\delta}}, \text{ as previously specified.}$$

$$\text{Thus } (1-t^2)^{-1} = O\left(n^{\frac{1}{4+\delta}}\right).$$

This observation together with (1.33), shows that

$$\begin{aligned} g_n(t) &= \frac{n}{\sqrt{3}(1-t^2)^{1/2}} \left[ 1 + O\left(n^{\frac{2+\delta}{4+\delta}}\right) \right]^{1/2} \\ &= \frac{n}{\sqrt{3}(1-t^2)^{1/2}} \left[ 1 + O\left(n^{\frac{2+\delta}{4+\delta}}\right) \right] \end{aligned} \quad (1.34)$$

Thus from (1.8), we have

$EN_n(f: -1+\epsilon, 1-\epsilon)$ ,

$$\begin{aligned} &\int_{-1+\epsilon}^{1-\epsilon} \frac{n}{\sqrt{3}(1-t^2)^{1/2}} \left[ 1 + O\left(n^{\frac{2+\delta}{4+\delta}}\right) \right] \\ &= \frac{n}{\sqrt{3}} \left[ 1 + O\left(n^{\frac{2+\delta}{4+\delta}}\right) \right] \left[ \sin^{-1} t \right]_{-1+\epsilon}^{1-\epsilon} \\ &= \frac{n}{\pi\sqrt{3}} \left[ 1 + O\left(n^{\frac{2+\delta}{4+\delta}}\right) \right] \left[ \sin^{-1}(1-\epsilon) - \sin^{-1}(\epsilon-1) \right] \\ &= \frac{n}{\pi\sqrt{3}} \left[ 1 + O\left(n^{\frac{2+\delta}{4+\delta}}\right) \right] \left[ 2\sin^{-1}(1-\epsilon) \right] \\ &= \frac{n}{\sqrt{3}} \left[ 1 + O\left(n^{\frac{2+\delta}{4+\delta}}\right) \right] \end{aligned}$$

$$\text{(as } \sin^{-1}(1-\epsilon) \sim \pi/2 \text{)} \quad (1.35)$$

### 1.5. Number of Zeros in Subintervals (ii) and (iii).

Here we show that in the range  $(1-\varepsilon, 1)$  and  $(-1, -1+\varepsilon)$  the number of zeros of (1.1) is negligibly small in comparisons to  $EN_n(f: -1+\varepsilon, 1+\varepsilon)$ ,

$$\text{Let } f(z) = f(\vec{y}(w), z) = \sum_{k=0}^n y_k(w) \Psi_k(z) \quad (1.36)$$

where  $y(w)$  denotes the random vector  $(y_0(w), y_1(w), \dots, y_n(w))$ .

$$\text{Now } \vec{f}(\vec{y}(w), 1) = \sum_{k=0}^n y_k(w) \Psi_k(1),$$

is a random variable with mean zero and

$$\text{variance } \sigma^2 = \sum_{k=0}^n \Psi_k^2(1) \geq \Psi_0^2(1) \geq 0,$$

and hence has the distribution function

$$\frac{1}{\sqrt{2\pi\sigma^2}} \int_{-\infty}^t \exp\left(-\frac{v^2}{2\sigma^2}\right) dv.$$

Now

$$\begin{aligned} P(|f(1)| \leq e^{-2n\varepsilon}) &= \left(\frac{2}{\pi\sigma^2}\right)^{1/2} \int_0^{e^{-2n\varepsilon}} \exp\left(-\frac{V^2}{2\sigma^2}\right) \\ &= \left(\frac{2}{\pi\sigma^2}\right)^{1/2} e^{-2n\varepsilon} < e^{-n\varepsilon} \end{aligned} \quad (1.37)$$

Let

$$I_n = \max_{0 \leq k \leq n} (y_k(w)) \quad (1.38)$$

$$\begin{aligned}
P(I_n \leq n) &= P\left(\max_{0 \leq k \leq n} (y_k(w))\right) \\
&= P\left(\prod_{k=0}^n |y_k(w)| \leq n\right) \\
&= \left(\prod_{k=0}^n P(y_k(w) \leq n)\right) \\
&= \left(\prod_{k=0}^n |1 - P(y_k(w) > n)|\right) \\
&= \left(\prod_{k=0}^n \left[1 - \sqrt{\frac{2}{\pi}} \int_0^\infty e^{-v^2/2} dv\right]\right)
\end{aligned}$$

Then

$$\geq \left[1 - (n+1) \sqrt{\frac{2}{\pi}} \int_0^\infty e^{-v^2/2} dv\right] 1 - e^{-n^2/2} \quad (n > n_0). \quad (1.39)$$

$$\text{Let } T_n = \max_{0 \leq k \leq n} |\Psi_k(1 + 2 \in e^{i\theta})|$$

For the Gegenbauer polynomials,  $h_n$  is determined

$$\text{As } \frac{2^{1-2\lambda} \pi \Gamma(n+2\lambda)}{(n+\lambda)(\Gamma(\lambda))^2 \Gamma(n+1)} \quad \text{for } \lambda > \frac{1}{2}.$$

Hence

$$b_n = h_n^{-1/2} < \alpha_1 n^{1/2}$$

where  $\alpha_1$  is a constant.

For the integral representation of Gegenbauer polynomial, we have

$$\Phi_n(t) = \frac{2^{1-2\lambda} \pi \Gamma(n+2\lambda)}{(n!(\Gamma(\lambda))^2)} \int_0^\pi (t + i\sqrt{1-t^2} \cos \theta) n(\sin \theta)^{2\lambda-1} d\theta \quad (1.40)$$

Remembering that  $\in = n^{-1/(4+\delta)}$  we have

$$\begin{aligned}
|\Phi_n(1 + 2 \in e^{i\theta})| &< \frac{2^{1-2\lambda} \pi \Gamma(n+2\lambda)}{(n!(\Gamma(\lambda))^2)} (1 + 2 \in)^n \\
&< \alpha_3 n^{\alpha_2} (1 + 2 \in)^n \\
&< \alpha_3 n^{\alpha_2} \exp(2n^{\frac{3+\delta}{4+\delta}}), \quad (1.41)
\end{aligned}$$

where  $\alpha_2$  and  $\alpha_3$  are constants involving  $\lambda$  only.

Hence from (1.40), we get  $T_n < An^{\alpha_2+1/2} \exp(2n^{\frac{3+\delta}{4+\delta}})$ , (1.42)

where A is a constant

Also

$$\begin{aligned} |f(1+2 \in e^{i\theta})| &= \left| \sum_{k=0}^n y_k(w) \Psi_k(1+2 \in e^{i\theta}) \right| \\ &\leq \left| \sum_{k=0}^n y_k(w) \Psi_k(1+2 \in e^{i\theta}) \right| \\ &\leq \sum_{k=0}^n T_n = n! T_n \quad (1.43) \end{aligned}$$

Hence from (1.39), it follows that

$$P\left(|1+2 \in e^{i\theta}| \leq n^2 T_n\right) \geq 1 - e^{-n/2} \quad (1.44)$$

This together with (1.42), gives

$$P|f(1+2 \in e^{i\theta})| \leq An^\alpha \exp(2n^{\frac{3+\delta}{4+\delta}}) \geq 1 - e^{-n/2} \quad (1.45)$$

where  $\alpha = \alpha_2 + 5/2$ .

So from (1.37) and (1.45), we obtain

$$\begin{aligned} P \frac{|f(1+2 \in e^{i\theta})|}{f(1)} &\leq An^\alpha \exp(2n^{\frac{3+\delta}{4+\delta}} + 2n \in) \\ &\geq P|f(1+2 \in e^{i\theta})| \leq An^\alpha \exp(2n^{\frac{3+\delta}{4+\delta}}) \\ &- P|f(1)| \leq e^{-2ne} \\ &> 1 - e^{-n/2} - e^{-ne} \\ &> 1 - \frac{2}{n} \quad (1.46) \end{aligned}$$

Let  $n(\epsilon)$  denote the number of zeros of  $f(y(w), z) = 0$  inside the circle  $|z - 1| \leq \epsilon$ .

It is easy to see that the number of zeros of (3.1.1) inside the interval  $1 - \epsilon \leq t \leq 1$  does not exceed  $n(\epsilon)$ .

By Jensen's theorem, we have

$$n(\epsilon) \leq \frac{1}{2\pi \log 2} \int_{\theta}^{2\pi} \log \left| \frac{f(1 + 2 \epsilon e^{i\theta})}{f(1)} \right| d\theta \quad \text{for } f(1) \neq 0$$

$$\frac{1}{2\pi \log 2} \int_{\theta}^{2\pi} \log \left\{ An^{\alpha} \exp \left( 2n^{\frac{3+\delta}{4+\delta}} \right) + 2n \epsilon \right\} d\theta, \quad (1.47)$$

except for a set of measure at most  $2/n$ , as evident from (1.46).

Thus from (1.47) and remarks made earlier, we obtain that the number of zeros of (1.1) in  $(1 - \epsilon, 1)$  is at most  $O \left( n^{\frac{3+\delta}{4+\delta}} \right)$  with probability at least  $1 - 2/n$ .

An identical result is obtainable for the number of zeros of (1.1) in  $(-1, -1 + \epsilon)$ , so that  $EN_n(f; -1 + \epsilon, 1 - \epsilon) = O \left( n^{\frac{3+\delta}{4+\delta}} \right)$ .

The above derivation together with the estimate of  $EN_n(f; -1 + \epsilon, 1 - \epsilon)$  in section 1.3. proves that  $EN_n(f; -1, 1) = \frac{n}{\sqrt{3}} + O \left( n^{\frac{3+\delta}{4+\delta}} \right)$ .

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