# Expected Number of Zeros of a Random Trigonometric Polynomial 

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#### Abstract

The asymptotic estimates of the expected number of real zeros of the polynomial $T(\theta) \equiv T_{0}(\theta, \sigma)=\sum_{j=1}^{n} g_{j}(\bar{\omega}) \cos j \theta$ i.e. $T(\theta)=g_{1} \cos \theta+g_{2} \cos 2 \theta+\ldots . . g_{n} \cos n \theta$ where $g_{j}(j=1,2, \ldots . . n)$ is a sequence of independent normally distributed random variables is such a number. The expected number of zeros of the above polynomial with coefficients $g_{j}(w)$, $j=1,2, \ldots . . n$ in be a sequence of independent random variables defined on probability space ( $\Omega, \lambda, \operatorname{Pr}$ ), each normally distributed with mean zero and variance one, then for all sufficiently large $n$ the variance of the number of real zeros of $T(\theta)$ is equal to $\operatorname{var}\{N(0, \pi)\}=O\left(n^{3 / 2}\right)$


Key words: Level crossings, Trigonometric functions Independent, identically distributed random variables, random algebraic polynomial, random algebraic equation, real roots, domain of attraction of the normal law, slowly varying function

## INTRODUCTION

Let $\quad T(\theta) \equiv T_{0}(\theta, \varpi)=\sum_{j=1}^{n} g_{j}(\varpi) \cos j \theta$
where $g_{1}(\bar{\varpi}), g_{2}(\bar{\sigma}), \ldots \ldots . . g_{n}(\bar{\sigma})$ is a sequence of independent random variables defined on a probability space $(\Omega, \lambda, \operatorname{Pr})$, each normally distributed with mean zero and variance one. Much has been written concerning $N_{K}(0,2 \pi)$, the number of crossings of a fixed level K by $\mathrm{T}(\theta)$, in the interval $(0,2 \pi)$. From the work of Dunnage [2] we know that, for all sufficiently large n , the mathematical expectation of is a $N_{0}(0,2 \pi) \equiv N(0,2 \pi)$ symptotic to $2 n / \sqrt{3}$. In Farahmand [3] and [5] we show that this asymptotic number of crossings remains invariant for any $K \equiv K_{n}$. such that $\mathrm{K}^{2} / \mathrm{n} \rightarrow 0$ as $\mathrm{n} \rightarrow \infty$ However, less information is known about the variance of $N(0,2 \pi)$. The only attempt so far is ....where an (fairly large) upper bound is obtained. Indeed this could be justified since the problem with finding the variance consists of different levels of difficulties with finding the mean. The degree of difficulty with this challenging problem is reflected in the delicate work of Maslova [8] and Sambandham et al, [7] with above obtained the variance of N for the case of random algebraic polynomial $\sum_{j}^{n} 0 g_{j} x^{j}$; a case involving analysis that is usually easier to handle. Qualls [9] also studied the variance of the number of real roots of a random trigonometric polynomial. However, he studied a different type of polynomial $\sum_{j}^{n} 0 a_{j} \cos j \theta+b_{j} \sin j \theta$ which has the property of being stationary and for which a special theorem has been developed by Cramer and Leadbetter [1].Here we look at the random trigonometric polynomial (1) as a non-stationary random process. First we are seeking to generalize Cramer and Leadbetter's [1] works concerning fractional moments which are mainly for the stationary case. To evaluate the variance specially, and some other applications generally it is important to consider the covariance of the number of real zeros of $\xi(t)$ in any two disjoint intervals. To this end, let $\xi(t)$ be a (non-stationary) real valued separable normal process possessing continuous sample paths, with probability one, such that for any $\theta_{1} \neq \theta_{2}$ the joint normal process $\xi\left(\theta_{1}\right), \xi\left(\theta_{2}\right), \xi^{\prime}\left(\theta_{1}\right)$ and $\xi^{\prime}\left(\theta_{2}\right)$ is non singular. Let ( $\mathrm{a}, \mathrm{b}$ ) and ( $\mathrm{c}, \mathrm{d}$ ) be any disjoint intervals on
which $\xi(t)$ is defined. The following theorem and the formula for the mean number of zero crossings Cramer and Leadbetter's [1] obtain the covariance of $\mathrm{N}(\mathrm{a}, \mathrm{b})$ and $\mathrm{N}(\mathrm{c}, \mathrm{d})$.

## Theorem -1

For any two disjoint intervals, (a,b) and (c,d) on which the process $\xi^{\prime}\left(\theta_{1}\right)$ is defined, we have

$$
E\{N(a, b) N(c, d)\}=\int_{c}^{d} \int_{a-\infty}^{\infty} \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} x y \mid p \theta_{1} \theta_{2}(0,0, x, y) d x d y d \theta_{1} d \theta_{2} \text { where for } a \leq \theta_{1} \leq b \text { and } \mathrm{c} \leq \theta_{2} \leq d, \mathrm{p} \theta_{1} \cdot \theta_{2}\left(x_{1}, x_{2}, x, y\right)
$$

denotes the four dimensional density function of $\xi\left(\theta_{1}\right), \xi\left(\theta_{2}\right), \xi^{\prime}\left(\theta_{1}\right), \xi^{\prime}\left(\theta_{2}\right)$.
A modification of the proof of Theorem 1 will yield the following theorem which, in reality, is only a corollary of Theorem 1.

## Theorem -2

For $\mathrm{p} \theta_{1} \cdot \theta_{2}\left(x_{1}, x_{2}, x, y\right)$ defined as in Theorem 1 we have

$$
E N^{2}(a, b)=\int_{c}^{d} \int_{a}^{b} \int_{-\infty}^{\infty} \int_{-\infty}^{\infty}|x y| p \theta_{1} \theta_{2}(0,0, x, y) d x d y d \theta_{1} d \theta_{2}
$$

By applying Theorem 2 to the random trigonometric polynomial (1) we will be able to find an upper limit for the variance of its number of zeros. This becomes possible by using a surprising and nontrivial result due to Wilkins [12] which reduces the error term involved for $E N(0,2 \pi)$ to $0(1)$. We conclude by proving the following.

## Theorem -3

If the coefficients $\mathrm{g}_{\mathrm{j}}(\mathrm{w}), \mathrm{j}=1,2, \ldots . \mathrm{n}$ in (1.1) be a sequence of independent random variables defined on probability space $(\Omega, \lambda, \operatorname{Pr})$, each normally distributed with mean zero and variance one, then for all sufficiently large $n$ the variance of the number of real zeros of $\mathrm{T}(\theta)$ satisfies $\operatorname{var}\{N(0, \pi)\}=O\left(n^{3 / 2}\right)$

## THE COVARIANCE OF THE NUMBER OF CROSSINGS

To obtain the result for the covariance, we shall carry through the analysis for the number of upcrossings, $\mathrm{N}_{\mathrm{u}}$. Indeed, the analysis for the number of down crossings would be similar and therefore, the result for the total number of crossings will follow. In order to find $E\left\{N_{u}(a, b) N_{u}(c, d)\right\}$ we require refining and extending the proof presented by Cramer and Leadbetter [1]. However, our proof follows their method and in the following, we highlight the generalization required to obtain our result. Let $a_{k}=(b-a) k 2^{-m}+a$ and similarly $b_{j}=(d-c) l 2^{-m}+c$ for $\mathrm{k}, 1=0,1,2, \ldots .2^{\mathrm{m}}-1$ and we define the random variable $\mathrm{Xk}, \mathrm{m}$ and Xlm as

$$
X k, m=\left\{\begin{array}{cc}
1 & \text { if } \xi\left(a_{k}\right)<0<\xi\left(a_{k+1}\right)  \tag{2}\\
0 & \text { otherwise }
\end{array} \text { And } \quad \text { Xlm }=\left\{\begin{array}{cc}
1 & \text { if } \xi\left(b_{l}\right)<0<\xi\left(b_{l+1}\right) \\
0 & \text { otherwise }
\end{array}\right.\right.
$$

In the following we show that

$$
Y_{m}=\sum_{l=0}^{2^{m}} \sum_{k=0}^{-1} X k, m, X l, m
$$

tends to $N_{u}(a, b) N_{u}(c, d)$ as $\mathrm{m} \rightarrow \infty$ with probability one. See also Cramer and Leadbetter [1], we first note that i $E\left\{N_{u}(a, b) N_{u}(c, d)\right\}_{\mathrm{s}}$ finite and therefore $\left\{N_{u}(a, b) N_{u}(c, d)\right\}$ is finite with probability one. Let v and r be the number of up crossings of $\xi(t)$ in (a,b) and (c,d), respectively, and write $\mathrm{t}_{1}, \mathrm{t}_{2}, \ldots . . \mathrm{t}_{\mathrm{v}}$ and $\mathrm{t}^{\prime}{ }_{1}, \mathrm{t}^{\prime}{ }_{2} \ldots \mathrm{t}^{\prime}$, for the points of upcrossings of zero by $\xi(t)$, there can be found two sub intervals for each $\mathrm{I}_{\mathrm{s}, \mathrm{m}}$ and $\mathrm{J}_{s^{\prime} \mathrm{m}}$ such that $\xi(t)$ in one is strictly positive and in the other, it is strictly negative. Thus it is apparent that $Y_{m}$ will count each of $\mathrm{t}_{\mathrm{s}} \mathrm{t}_{\mathrm{s}}$ '. That is, $Y_{m} \geq v r$, for all sufficiently large $m$. On the other hand, if $\xi\left(a_{k}\right) \xi\left(b_{k+1}\right)<0$ and $\xi\left(b_{t}\right) \xi\left(b_{t+1}\right)<0$ then $\xi(t)$ must have a zero in $\left(a_{k}, a_{k+1}\right)$ and $\left(b_{l}, b_{l+1}\right)$ and hence $Y_{m} \leq v r$ and hence $Y_{m} \rightarrow N_{u}(a, b) N_{u}(c, d)$ as $m \rightarrow \infty$, with probability one. Now from (2.1) we can see at once that

$$
\begin{equation*}
E\left(Y_{m}\right)=\sum_{l=0}^{2^{m}-12^{m}-1} \sum_{k=0} \operatorname{Pr}\left(X_{k}, m X_{l}, m=1\right)=\sum_{l=0}^{2^{m}-12^{m}-1} \sum_{k=0} \operatorname{Pr}\left(X_{k}, m=X_{l}, m=1\right) \tag{3}
\end{equation*}
$$

We write $\eta_{k}$ for the random variable $2^{m}\left\{\xi\left(a_{k+1}\right)-\xi\left(a_{k}\right)\right\}$ and similarly $\eta_{1}^{\prime}$ for $2^{m}\left\{\xi\left(b_{k+1}\right)-\xi\left(b_{k}\right)\right\}$, then we have $\operatorname{Pr}\left(X_{k}, m=X_{l}, m=1\right)$

$$
\begin{align*}
& =\operatorname{Pr}\left(0>\xi\left(a_{k}\right)>2^{-m} \eta_{k}, \text { and } 0>\xi\left(b_{l}\right)>2^{-m} \eta_{l}\right) \\
& =\int_{0}^{\infty} \int_{0}^{2^{-m_{x}}} \int_{0}^{2^{-m_{y}}} \int_{0} \operatorname{Pm}, k, l\left(z_{1}, z_{2}, x, y\right) d z_{1} d z_{2} d x d y \tag{4}
\end{align*}
$$

For $\mathrm{k}, \mathrm{l},\left(\mathrm{z}_{1}, \mathrm{z}_{2}, \mathrm{x}, \mathrm{y}\right)$ denotes the four dimensional normal density function for $\eta_{k}$ and $\eta_{k}^{\prime}$. A simple calculation shows see Farahmand [6], that if $\theta_{1}$ and $\theta_{2}$ are the fixed interval $(\mathrm{a}, \mathrm{b})$ and ( $\left.\mathrm{c}, \mathrm{d}\right)$, respectively and $\mathrm{k}_{\mathrm{m}}$ and $\mathrm{l}_{\mathrm{m}}$ are such that $a_{k_{m}}<\theta_{1}<a_{k_{m+1}}$ and $b_{l_{m}}<\theta_{2}<b_{l_{m+1}}$ for each m, then all members of the covariance matrix of $\mathrm{p}_{\mathrm{m}, \mathrm{k}, 1}\left(\mathrm{z}_{1}, \mathrm{z}_{2}, \mathrm{x}, \mathrm{y}\right)$ will tend to the corresponding members of the covariance matrix of $p_{\theta_{1}}, \theta_{2}\left(\mathrm{z}_{1}, \mathrm{z}_{2}, \mathrm{x}, \mathrm{y}\right)$. This co-variance matrix is, indeed, nonsingular. Now let $t=2^{m}{ }_{z_{1}}$ and $\mathrm{r}=2^{m}{ }_{z_{2}}$ then from (3) and (4) we have

$$
\begin{align*}
& E\left(Y_{m}\right)=\sum_{l=0}^{2^{m}} \sum_{k=0}^{-1} 2^{-2 m} \int_{0}^{\infty} \int_{0}^{\infty} \int_{0}^{x} \int_{0}^{y} P m, k, l\left(2^{-m} t 2^{-m} r, x, y\right) d t d r d x d y  \tag{5}\\
& =\int_{a}^{b} \int_{c}^{d} \int_{0}^{\infty} \int_{0}^{\infty} \int_{0}^{x} \int_{0}^{y} \Psi_{m}, \theta_{1}, \theta_{2}\left(2^{-m} t 2^{-m} r, x, y\right) d t d r d x d y d \quad \boldsymbol{\theta}_{1} d \theta_{2}
\end{align*}
$$

in which $\Psi_{m}, \theta_{1}, \theta_{2} \quad(t, r, x, y)=P m, k, l(t, r, x, y)$ for $a_{k}<\theta_{1}<a_{k+1}$ and $b_{l}<\theta_{2}<b_{l+1}$. It follows, similar to Cramer and Leadbetter [1], that $m \rightarrow \infty$
$\Psi_{m}, \theta_{1}, \theta_{2}\left(2^{-m} t 2^{-m} r, x, y\right) \rightarrow p \theta_{1} \theta_{2}(0,0, x, y)$ which together with dominated convergence proves Theorem 1.

## THE VARIANCE OF THE NUMBER OF REAL ZEROS

It will be convenient to evaluate the $\mathrm{EN}(\mathrm{N}-1)$ rather than the variance itself since $\mathrm{N}(\mathrm{N}-1)$ can be expressed much more simply. The proof is similar to that established above for covariance, therefore we only point out the generalization required to obtain the result. To avoid degeneration of the joint normal density $p \theta_{1}, \theta_{2}\left(z_{1}, z_{2}, x, y\right)$, we should omit those zeros in the squares of side $2^{-\mathrm{m}}$ obtained from equal points in the axes (and therefore to evaluate $\mathrm{EN}(\mathrm{N}-1))$. To this and for any $\mathrm{g}=\left(\mathrm{g}_{1}, \mathrm{~g}_{2}\right)$ lying in the unit square and $\mathrm{c}>0$, let $\mathrm{A}_{\text {me }}$ denote the set of all points g in the unit square that for all $s$ belonging to the squares of side $2^{-\mathrm{m}}$ set containing $g$ we have $\left|s_{1}-s_{2}\right|>\varepsilon$. Let $\lambda_{m \varepsilon}$ denote the characteristic function of the set $\lambda_{m \varepsilon}$. Finally, similar to the covariance case, let

$$
X_{k, m}=\left\{\begin{array}{cc}
1 & \text { if } \xi\left(a_{k}\right)<0<\xi\left(a_{k+1}\right) \\
0 & \text { otherwise }
\end{array}\right.
$$

for $\mathrm{k}=0,1,2, \ldots .2^{\mathrm{m}}-1$, where $a_{k}=(b-a) k 2^{-m}+a$. Now let

$$
\begin{equation*}
M_{m \varepsilon}=\sum_{k=0}^{2^{m}-1} \sum_{(l=0, l \neq k)}^{2^{m}} X_{k, m}^{-1} X_{l, m} \lambda_{m \varepsilon}\left(2^{-m} k, 2^{-m} l\right) \tag{6}
\end{equation*}
$$

Similar to Cramer and Leadbetter [1] we show that $M_{m e}$ is a non decreasing function of $m$ for any fixed e. It is obvious that $\mathrm{M}_{\mathrm{me}}$ is a non decreasing function of e for fixed to m , and then by two applications of monotone convergence it would be justified to change the order of limits in $\lim _{\varepsilon \rightarrow 0} \lim _{m \rightarrow \infty} \lim m \varepsilon$. To this end, we note that each term of the sums of $M_{m \varepsilon}$ corresponds to a square of side $2^{-\mathrm{m}}$. For fixed $\varepsilon>0$, the typical term is one if both of the followings statements are satisfied; (i) every point $\mathrm{s}=\left(\mathrm{s}_{1}, \mathrm{~s}_{2}\right)$ in the square is such that $\left|s_{1}-s_{2}\right|>\varepsilon$ and (ii) $X_{k, \mathrm{~m}}=\mathrm{X}_{\mathrm{l}, \mathrm{m}}=1$. When m is increase by one unit, the square is divided into four subsquares, in each of which property (i) still holds. Correspondingly, the typical term of sum is divided into four terms, formed by replacing $m$ by $m+1$ and each $k$ or 1 by $2 k$ and 21 , for $a_{k+1}$ and $a_{1+1}$. Since $X_{k, m}=X_{1, m}=1$ we must, with probability one, have at least one of these four terms equal one. Hence $\mathrm{M}_{\mathrm{me}}$ is a non decreasing function of m .
In the following, we show that $\lim _{m \rightarrow \infty} \lim _{\varepsilon \rightarrow 0} s=N_{u}\left(N_{u}-1\right)$.
We first note that if the typical term in the sum of $M_{m \varepsilon}$ is nonzero it follows that $\left|s_{1}-s_{2}\right|>\varepsilon$, since it is impossible to have $\xi\left(a_{k}\right)<0<\xi\left(a_{k+1}\right)$ and $\xi\left(a_{k+1}\right)<0<\xi\left(a_{k+2}\right)$. Therefore, the characteristic function appearing in the formula for $M_{m \varepsilon}$ in in (6) is one and hence

$$
\begin{equation*}
\lim _{\varepsilon \rightarrow 0} M_{m \varepsilon}=\sum_{k=0}^{2^{m}} \sum_{(l=0, l \neq k)}^{2^{m}} X_{k, m}^{-1} X_{l, m} \tag{7}
\end{equation*}
$$

is clearly in the form of $\mathrm{Y}_{\mathrm{m}}$ defined in Section 2 except that the summations in (7) cover all the k and 1 such that $k \neq l$. Hence from (7), we can write $\lim _{m \rightarrow \infty} \lim _{\varepsilon \rightarrow 0} M_{m \varepsilon}=N_{u}\left(N_{u}-1\right)$
Therefore the same pattern as for the covariance case yields

$$
\begin{equation*}
E\left[N_{u}(a, b)\left\{N_{u}(a, b)-1\right\}\right]=\lim { }_{\varepsilon \rightarrow 0} \iint_{D(\varepsilon)} \int_{0}^{\infty} \int_{0}^{\infty}|x y|, p_{\theta_{1}}, \theta_{2}(0,0, x, y) d x d y d \theta_{1} d \theta_{2} \tag{8}
\end{equation*}
$$

where $D(\varepsilon)$ denotes the domain in the two dimensional space with coordinates $\theta_{1}, \theta_{2}$ such that $a<\theta_{1}, \theta_{2}<b$ and $\left|\theta_{1}-\theta_{2}\right|>\varepsilon$. Now notice that for $\theta_{1}=\theta_{2}=0$ the $p \theta_{1} \cdot \theta_{2}(0,0, x, y)$ degenerates to just $p_{\theta}(0, x)$, the two dimensional joint density function of $\xi(\theta)$.and $\xi^{\prime}(\theta)$. Hence from (8), we have

$$
\begin{equation*}
=E\left[N_{u}(a, b)\left\{N_{u}(a, b)-1\right\}\right] \int_{a}^{b} \int_{a}^{b} \int_{0}^{\infty} \int_{0}^{\infty}|x y|, p_{\theta_{1}, \theta_{2}}(0,0, x, y) d x d y d \theta_{1} d \theta_{2}-\int_{a}^{b} \int_{0}^{\infty}|x|, p_{\theta}(0, x) d x d \theta \tag{9}
\end{equation*}
$$

Now since $\int_{a}^{b} \int_{0}^{\infty}|x|, p_{\theta}(0, x) d x d \theta$ is $E N_{u}(a, b)$ the result of Theorem 2 follows.

## RANDOM TRIGONOMETRIC POLYNOMIAL

To evaluate the variance of the number of real roots of(!) (1) in the interval $(0, \pi)$ we use Theorem 2 to consider the interval $\left(\varepsilon^{\prime}, \pi-\varepsilon^{\prime}\right)$. The variance for the intervals and $\left(\varepsilon^{\prime}, \pi-\varepsilon^{\prime}\right)$ are obtained using an application of Jenson's theorem Rudin [10] and Titchmarsh, [11]. We chose $\varepsilon^{\prime}=n^{-1 / 2}$ which as we will see later, yields the smallest possible error term. First, for any $\theta_{1}$ and $\theta_{2}$ in $\left(\varepsilon^{\prime}, \pi-\varepsilon^{\prime}\right)$ such that $\left|\theta_{1}-\theta_{2}\right|>\varepsilon$ where $\mathcal{E}^{\prime}=n^{-1 / 2}$, we evaluate the joint density function of the random variable $T\left(\theta_{1}\right), T\left(\theta_{2}\right), T^{\prime}\left(\theta_{1}\right)$ and $T^{\prime}\left(\theta_{2}\right)$. Since for any $\theta$ we have
$\sum_{j=1}^{n} \cos j \theta=[\sin \{(n+1 / 2) \theta\} / \sin (\theta / 2)-1] / 2$ and also since for the above choice of $\theta_{1}$ and $\theta_{2}, \theta_{1}+\theta_{2}<2\left(\pi-\mathcal{E}^{\prime}\right)$ we can show

$$
\begin{align*}
& A\left(\theta_{1}, \theta_{2}\right)=\operatorname{cov}\left\{T\left(\theta_{1}\right), \mathrm{T}\left(\theta_{2}\right)\right\}=\sum_{j=1}^{n} \cos j \theta_{1} \operatorname{cosj} \theta_{2} \\
& =\left[\begin{array}{l}
\sin \left\{(n+1 / 2)\left(\theta_{1}-\theta_{2}\right)\right\} / \sin \left\{\left(\theta_{1}-\theta_{2}\right) / 2\right\} \\
\left.+\sin \left\{(n+1 / 2)\left(\theta_{1}+\theta_{2}\right)\right\} / \sin \left\{\left(\theta_{1}+\theta_{2}\right) / 2\right\}-2\right] / 4 \\
=O(1 / \varepsilon)+O\left(1 / \varepsilon^{\prime}\right)
\end{array}\right.
\end{align*}
$$

Similarly, we can obtain the following two estimates

$$
\begin{equation*}
C\left(\theta_{1}, \theta_{2}\right)=\operatorname{cov}\left\{T\left(\theta_{1}\right), \mathrm{T}\left(\theta_{2}\right)\right\}=-\sum_{j=1}^{n} j \sin j \theta_{1} \operatorname{cosj} \theta_{2}=\left(\vartheta / \vartheta \theta_{1}\right)\left\{A \theta_{1}, \theta_{2}\right\}=O\left(n / \varepsilon+\varepsilon^{-2}+n / \varepsilon^{\prime}+\varepsilon^{\prime-2}\right) \tag{11}
\end{equation*}
$$

and

$$
\begin{equation*}
B\left(\theta_{1}, \theta_{2}\right)=\operatorname{cov}\left\{T\left(\theta_{1}\right), \mathrm{T}\left(\theta_{2}\right)\right\}=\sum_{j=1}^{n} j \sin j \theta_{1} \sin j \theta_{2}=\left(\vartheta / \vartheta \theta_{2}\right)\left\{C \theta_{1}, \theta_{2}\right\}=O\left(n^{2} / \varepsilon+n / \varepsilon^{2}+n^{2} / \varepsilon+n / \varepsilon^{2}+\varepsilon^{{ }^{3}}\right) \tag{12}
\end{equation*}
$$

Also in the lemma in Farahmand [3], we obtain

$$
\operatorname{var}\left(T\left(\theta_{1} 0\right)\right)=n / 2+O\left(\varepsilon^{\prime-1}\right), \operatorname{var}\left(T^{\prime}\left(\theta_{1} 0\right)\right)=n^{3} / 6+O\left(n^{2} / \varepsilon^{\prime}+n / \varepsilon^{\prime 2}+\varepsilon^{\prime-3}\right)
$$

and

$$
\operatorname{cov}\left\{T\left(\theta_{1}\right) T\left(\theta_{2}\right)\right\}=O\left(n / \varepsilon^{\prime}+\varepsilon^{\prime-2}\right)
$$

These together with (10-12) give the covariance matrix for the joint density function

$$
T\left(\theta_{1}\right), T\left(\theta_{2}\right), T^{\prime}\left(\theta_{1}\right) \text { and } T^{\prime}\left(\theta_{2}\right) \text { as } \Sigma\left[\begin{array}{cccc}
n / 2+O\left(\varepsilon^{-1}\right) & A\left(\theta_{1}, \theta_{2}\right) & C\left(\theta_{1}, \theta_{1}\right) & C\left(\theta_{2}, \theta_{1}\right)  \tag{13}\\
A\left(\theta_{1}, \theta_{2}\right) & n / 2+O\left(\varepsilon^{-1}\right) & C\left(\theta_{1}, \theta_{2}\right) & C\left(\theta_{2}, \theta_{2}\right) \\
C\left(\theta_{1}, \theta_{1}\right) & C\left(\theta_{1}, \theta_{2}\right) & n^{3} / 6+O\left(n^{2} / \varepsilon^{\prime}\right) & B\left(\theta_{1}, \theta_{2}\right) \\
C\left(\theta_{2}, \theta_{1}\right) & C\left(\theta_{2}, \theta_{2}\right) & B\left(\theta_{1}, \theta_{2}\right) & n^{3} / 6+O\left(n^{2} / \varepsilon^{\prime}\right)
\end{array}\right]
$$

This covariance matrix for all $n \geq 4,0<\theta_{1}, \theta_{2}<\pi$ such that $\theta_{1} \neq \theta_{2}$ is positive definite. Hence $|\Sigma|>0$ and, if $\Sigma_{i j}$ is cofactor of the (ij)th element of $\Sigma$, then $\Sigma_{33}>0, \Sigma_{44}>0$ and $\Sigma_{34}=\Sigma_{43}$. From [1] we have

$$
\begin{equation*}
p \theta_{1} \cdot \theta_{2}(0,0, x, y)=\left(4 \pi^{2}\right)^{-1}\left|\sum\right|^{-1 / 2} \exp \left[-\left\{\sum_{33} x^{2}+\sum_{44} y^{2}+\left(\sum_{34}+\sum_{43}\right) x y\right\} / 2|\Sigma|\right] \tag{14}
\end{equation*}
$$

Now let $\mathrm{q}=\left(\sum_{33} /|\Sigma|+\right)^{1 / 2} x$ and $\mathrm{s}=\left(\sum_{44} /|\Sigma|+\right)^{1 / 2} y$.
Then from (14) we can write

$$
\begin{equation*}
\int_{-\infty}^{\infty} \int_{-\infty}^{\infty}|x, y| p \theta_{1} \cdot \theta_{2}(0,0, x, y) d x d y=\left(4 \pi^{2}\right)^{-1}\left|\sum_{33}\right|^{-1} \sum_{44}{ }^{-1} \int_{-\infty}^{\infty} \int_{-\infty}^{\infty}|q, s| \exp \left\{-\left(q^{2}+s^{2}+2 p q s\right) / 2\right\} d q d s \tag{15}
\end{equation*}
$$

where $\rho=\left(\sum_{34}+\sum_{43}\right) / 2\left(\sum_{33} \sum_{44}\right)^{1 / 2}$ and $0 \leq p^{2}<1$. The value of the integral in (15) can be obtained by a similar method to Cramer and Leadbetter [1]. Let $u=\left(1-\rho^{2}\right)^{1 / 2} q$ and $\mathrm{v}=\left(1-\rho^{2}\right)^{1 / 2} s$ then we have

$$
\begin{align*}
& I=\int_{0}^{\infty} \int_{0}^{\infty} \exp \left\{-\left(q^{2}+s^{2}+2 p q s\right) / 2\right\} d q d s=\left(1-\rho^{2}\right)^{-1} \int_{00}^{\infty} \int_{0} \exp \left\{-\left(u^{2}+v^{2}+2 p u v\right) / 2\left(1-\rho^{2}\right)\right\} d u d v \\
& =\pi\left(1-\rho^{2}\right)^{-1 / 2}\left\{1 / 2-(\pi)^{-1}\right\} \int_{0}^{\rho}\left(1-x^{2}\right)^{-1 / 2} d x=\left(1-\rho^{2}\right)^{-1 / 2} \arccos \rho=\phi \csc \phi \tag{16}
\end{align*}
$$

where $\rho=\cos \phi$. Use has been made of the fact that (see for example Cramer and Leadbetter [1])

$$
\int_{0}^{\infty} \int_{0}^{\infty} \exp \left\{-\left(u^{2}+v^{2}+2 p u v\right) / 2\left(1-\rho^{2}\right)\right\} d u d v=\pi\left(1-\rho^{2}\right)^{-1 / 2}\left\{1 / 2-(\pi)^{-1}\right\} \int_{0}^{\rho}\left(1-x^{2}\right)^{-1 / 2} d x
$$

Therefore from (16) by differentiation we can obtain

$$
\begin{equation*}
\int_{0}^{\infty} \int_{-\infty}^{\infty} q s \exp \left\{-\left(q^{2}+s^{2}+2 p q s\right) / 2\right\} d q d s=-(\mathrm{dI}) /(\mathrm{d} \rho)=\csc ^{2} \phi(1-\phi \cot \phi) \tag{17}
\end{equation*}
$$

We can easily show that

$$
\int_{0}^{\infty} \int_{-\infty}^{\infty} q s \exp \left\{-\left(q^{2}+s^{2}+2 p q s\right) / 2\right\} d q d s=\csc ^{2} \phi\{1+(\pi-\phi \cot \phi)\}
$$

Which together with (17) evaluates the integral in (15) as

$$
\begin{equation*}
\int_{-\infty}^{\infty} \int_{-\infty}^{\infty}|q s| \exp \left\{-\left(q^{2}+s^{2}+2 p q s\right) / 2\right\} d q d s=4 \csc ^{2} \phi\{1+(\pi / 2-\phi) \cot \phi\} \tag{18}
\end{equation*}
$$

Now from (13) we can show

$$
\begin{equation*}
\sum_{44}=n^{5} / 24+O\left(n^{4} / \varepsilon^{\prime}\right)=\sum_{33} \quad \text { and } \quad \sum_{34}=O\left(n^{4} / \varepsilon^{\prime}\right)=\sum_{43} \tag{19-20}
\end{equation*}
$$

Also from (19) and (20) and with the above choice from (18) we can obtain

$$
\rho=\left(\sum_{34}+\sum_{33}\right) / 2\left(\sum_{34} \sum_{33}\right)^{1 / 2}=O\left(1 / n \varepsilon^{\prime}\right) \rightarrow 0 \text { as } \mathrm{n} \rightarrow \infty
$$

Therefore $\phi \rightarrow \pi / 2$ for all sufficiently large n and hence from (18), we can see

$$
\begin{equation*}
\int_{-\infty}^{\infty} \int_{-\infty}^{\infty}|q s| \exp \left\{-\left(q^{2}+s^{2}+2 p q s\right) / 2\right\} d q d s=4+O\left(1 / n \varepsilon^{\prime}\right) \tag{21}
\end{equation*}
$$

Also from (10)-(11) we can write

$$
\left|\sum\right|=\left\{n / 2+O\left(\varepsilon^{\prime-1}\right)\right\}^{2}\left\{n^{3} / 6+O\left(\varepsilon+\varepsilon^{\prime-1}\right)\right\}^{2}
$$

Therefore from this (15) and (21) the integrand that appears in (8) is asymptotically independent of $\theta_{1}$ and $\theta_{2}$ and since by the definition of $\mathrm{D}(\mathrm{e})$, the area of the integration is $\left(\pi-2 \varepsilon^{\prime}\right)^{2} \varepsilon\left(\pi-2 \varepsilon^{\prime}\right)+\varepsilon^{2}=\pi^{2}+O\left(\varepsilon+\varepsilon^{\prime}\right)$ we have

$$
\begin{equation*}
E\left[N\left(\varepsilon^{\prime}, \pi-\varepsilon^{\prime}\right)\left\{N\left(\varepsilon^{\prime}, \pi-\varepsilon^{\prime}\right)-1\right\}\right]=n^{2} / 3+O\left(n / \varepsilon^{\prime}+n \varepsilon+n \varepsilon^{\prime}\right) \tag{22}
\end{equation*}
$$

We now denote the mathematical expectation of $\mathrm{N}^{2}$ in the interval ( $0, \mathrm{e}$ ). Similar to Farahmand [2-3] we apply Jensen's theorem on a random integral function of the complex variable z ,

$$
T(z, w)=\sum_{j=1}^{n} g_{j}(w) \cos j z
$$

Let $\mathrm{N}(\mathrm{r})$ denote the number of real zeros of $\mathrm{T}(\mathrm{z}, \mathrm{w})$ in $\mathrm{z}<\mathrm{r}$. For any integer j from Farahmand [3], we have

$$
\begin{equation*}
\operatorname{Pr}\left[N\left(\varepsilon^{\prime}\right)>3 n \varepsilon^{\prime}+j\right]<(2 / \sqrt{n}) \varepsilon^{-j / 2}+\exp \left(-j / 2-n^{2} \varepsilon^{\prime} j / 2\right)<3 \varepsilon^{-j / 2} \tag{23}
\end{equation*}
$$

Let $n^{\prime}=\left[3 n \varepsilon^{\prime}\right]$ be the smallest integer greater than or equal to $3 n \varepsilon^{\prime}$ then since $N\left(\varepsilon^{\prime}\right) \leq 2 n$ is a non negative integer, from (23) and by dominated convergence, for efficiently large $n$ we have

$$
\begin{align*}
& E N^{2}\left(\varepsilon^{\prime}\right)=\sum_{j=0}(2 j-1) \operatorname{Pr}\left(N\left(\varepsilon^{\prime}\right) \geq j\right) \\
& =\sum_{0<j \leq n^{\prime}}(2 j-1) \operatorname{Pr}\left(N\left(\varepsilon^{\prime}\right) \geq j\right)+\sum_{j=1}^{n}\left(2 n^{\prime}-1+2 j\right) \operatorname{Pr}\left(N\left(\varepsilon^{\prime}\right) \geq n^{\prime}+j\right) \\
& \leq \sum_{j=1}^{n^{\prime}}\left(2 n^{\prime}-1\right)+3+\sum_{j=1}^{n}\left(2 n^{\prime}-1+1+2 j\right) \varepsilon^{-j / 2}=n^{\prime}+O\left(n^{\prime}\right) \\
& =O\left(n^{2} \varepsilon^{\prime 2}\right) \tag{24}
\end{align*}
$$

The interval $\left(\pi-\varepsilon^{\prime}, \pi\right)$ can also be treated in exactly the same way to give the same result. Now we can use delicate result due to Wilkins [12] which states that $E N(0, \pi)=n / \sqrt{2}+O(1)$. From this and (22 \& 24) and since $\varepsilon=\varepsilon^{\prime}$, we obtain

$$
\begin{align*}
& \operatorname{var}\{N(0, \pi)\}=E\left\{N\left(0, \varepsilon^{\prime}\right)+N\left(\varepsilon^{\prime}, \pi-\varepsilon^{\prime}\right)+N\left(\pi-\varepsilon^{\prime}, \pi\right)\right\}^{2}-\{E N(0, \pi)\}^{2} \\
& =n^{2} / 3+O\left(n^{2} \varepsilon^{\prime 2}+n / \varepsilon^{\prime}+n^{2} \varepsilon^{\prime}\right)-\{n / \sqrt{3}+O(1)\}^{2} \\
& O\left(n^{2} \varepsilon^{\prime 2}+n / \varepsilon^{\prime}+n^{2} \varepsilon^{\prime}\right) \tag{25}
\end{align*}
$$

Use has been made of the fact that $E N\left(\varepsilon^{\prime}, \pi-\varepsilon^{\prime}\right) \sim E N\left(0, \varepsilon^{\prime}\right)=O\left(n \varepsilon^{\prime}\right)$, see Farahmand [5] and therefore $E\left[N\left(0, \varepsilon^{\prime}\right) N\left(\varepsilon^{\prime}, \pi-\varepsilon^{\prime}\right)\right]=n O\left(N\left(0, \varepsilon^{\prime}\right)\right)=O\left(n^{2}, \varepsilon^{\prime}\right)$ and also from (24) $E N^{2}\left(0, \varepsilon^{\prime}\right) \sim E N^{2}\left(\pi-\varepsilon^{\prime}, \pi\right)=O\left(n^{2}, \varepsilon^{\prime 2}\right)$.
Finally from (25) and since $\varepsilon^{\prime}=n^{-1 / 2}$

## CONCLUSION

After proving all theorem and lemmas we conclude that considering a polynomial (1) where the coefficients are a sequence of independent random variables defined on probability space $(\Omega, \lambda, \operatorname{Pr})$, each normally distributed with mean zero and variance one, then for all sufficiently large $n$ the variance of the number of real zeros of $\mathrm{T}(\theta)$ satisfies

$$
\operatorname{var}\{N(0, \pi)\}=O\left(n^{3 / 2}\right)
$$

Hence the theorem proved.

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