

# The Distribution of Intervals between Zeros of a Stationary Random Function

M. S. Longuet-Higgins

*Phil. Trans. R. Soc. Lond. A* 1962 **254**, 557-599

doi: 10.1098/rsta.1962.0006

## Email alerting service

Receive free email alerts when new articles cite this article - sign up in the box at the top right-hand corner of the article or click [here](#)

To subscribe to *Phil. Trans. R. Soc. Lond. A* go to: <http://rsta.royalsocietypublishing.org/subscriptions>

# THE DISTRIBUTION OF INTERVALS BETWEEN ZEROS OF A STATIONARY RANDOM FUNCTION

BY M. S. LONGUET-HIGGINS

*National Institute of Oceanography, Wormley*

(Communicated by G. E. R. Deacon, F.R.S.—Received 21 August 1961)

## CONTENTS

	PAGE		PAGE
INTRODUCTION	558	4.2. The cases $n = 1, 2$ and $3$	573
1. GENERAL RELATIONS BETWEEN $P_m(\tau)$ , $p(n, \tau)$ AND $W(S)$	559	4.3. General values of $n$	574
1.1. Generalization of a result of McFadden	560	4.4. Asymptotic behaviour of $P_m(\tau)$	576
1.2. Series for $P_m(\tau)$	562	4.5. Approximations to $P_m(0)$	577
1.3. Relation to a series of Rice	563	4.6. Disproof of the 'exponential hypothesis'	578
1.4. Series for $p(n, \tau)$	564	4.7. Further estimates of $P_m(0)$	578
2. EVALUATION OF $W(S)$ FOR GAUSSIAN PROCESSES	565	4.8. Asymptotic behaviour of $p(n, \tau)$	579
2.1. A general expression for $W(S)$	565	5. A COMPARISON OF DIFFERENT APPROXI- MATIONS TO $P_0(\tau)$	580
2.2. The cases $n = 1, 2$ and $3$	567	5.1. Rice's approximation (1945)	580
2.3. General values of $n$	568	5.2. The approximation $P_0^{(2)}$	580
3. ASYMPTOTIC EXPANSIONS NEAR THE ORIGIN: $\psi_\tau$ REGULAR	569	5.3. The 'multiply conditioned' approximations	581
3.1. Expansion of $W(S)$	569	5.4. McFadden's first approximation	582
3.2. Evaluation of $P_m(\tau)$	570	5.5. $p_n(\tau)$ and $p_n^*(\tau)$	583
3.3. Evaluation of $p(n, \tau)$	571	5.6. McFadden's second approximation	586
4. ASYMPTOTIC EXPANSIONS NEAR THE ORIGIN: A SINGULAR CASE	572	5.7. The 'narrow-band' approximation	587
4.1. Expansion of $W(S)$	572	5.8. Numerical computations	588
		REFERENCES	599

The probability density  $P_m$  of the spacing between the  $i$ th zero and the  $(i+m+1)$ th zero of a stationary, random function  $f(t)$  (not necessarily Gaussian) is expressed as a series, of a type similar to that given by Rice (1945) but more rapidly convergent. The partial sums of the series provide upper and lower bounds successively for  $P_m$ . The series converges particularly rapidly for small spacings  $\tau$ . It is shown that for fixed values of  $\tau$ , the density  $P_m(\tau)$  diminishes more rapidly than any negative power of  $m$ .

The results are applied to Gaussian processes; then the first two terms of the series for  $P_m(\tau)$  may be expressed in terms of known functions. Special attention is paid to two cases:

(1) In the 'regular' case the covariance function  $\psi(t)$  is expressible as a power series in  $t$ ; then  $P_m(\tau)$  is of order  $\tau^{\frac{1}{2}(m+2)(m+3)-2}$  at the origin, and in particular  $P(\tau)$  is of order  $\tau$  (adjacent zeros have a strong mutual repulsion). The first two terms of the series give the value of  $P_0(\tau)$  correct to  $\tau^{18}$ .

(2) In a singular case, the covariance function  $\psi(t)$  has a discontinuity in the third derivative. This happens whenever the frequency spectrum of  $f(t)$  is  $O(\text{frequency})^{-4}$  at infinity. Then  $P_m(\tau)$  is

shown to tend to a positive value  $P_m(0)$  as  $\tau \rightarrow 0$  (neighbouring zeros are less strongly repelled). Upper and lower bounds for  $P_m(0)$  ( $m = 0, 1, 2, 3$ ) are given, and it is shown that  $P_0(0)$  is in the neighbourhood of  $1.155\psi'''/(-6\psi'')$ . The conjecture of Favreau, Low & Pfeffer (1956) according to which in one case  $P_0(\tau)$  is a negative exponential, is disproved.

In a final section, the accuracy of other approximations suggested by Rice (1945), McFadden (1958), Ehrenfeld *et al.* (1958) and the present author (1958) are compared and the results are illustrated by computations, the frequency spectrum of  $f(t)$  being assumed to have certain ideal forms: a low-pass spectrum, band-pass spectrum, Butterworth spectrum, etc.

### INTRODUCTION

The problem of finding the statistical distribution of the intervals  $\tau$  between zeros of a stationary, random function  $f(t)$  is one with many physical applications, for example, to noise in electrical circuits (Rice 1944, 1945; McFadden 1956, 1958), sea waves (Ehrenfeld *et al.* 1958), microseisms (Longuet-Higgins 1953) or irregularities in the ionosphere (Briggs & Page 1955; Longuet-Higgins 1957). Yet in one most useful case, when  $f(t)$  itself is Gaussian, only approximate solutions to the problem have been found. One such solution was given in a previous paper (1958). In the present paper the solution is expressed in the form of a series, in which each term is an integral of the joint probability

$$W(+, -, -, \dots, -) dt_1 \dots dt_n$$

that  $f(t)$  have an up-crossing in the infinitesimal interval  $(t_1, t_1 + dt_1)$  and a down-crossing in the remaining  $(n-1)$  intervals  $(t_i, t_i + dt_i)$  ( $i = 2, 3, \dots, n$ ). The series is somewhat similar to one given earlier by Rice (1945) but converges more rapidly. Moreover, successive partial sums of the series provide upper and lower bounds to the required distribution  $P_0(\tau)$ . The present series leads to a much more accurate estimate of the behaviour of  $P_0$  near the origin  $\tau = 0$  and to a systematic comparison of other approximations that have been previously proposed.

Following Rice (1945), the probability density of the interval between the  $i$ th and the  $(i+m+1)$ th zero of  $f(t)$  is denoted by  $P_m(\tau)$ ; and the probability of exactly  $n$  zeros occurring in the interval  $(t, t+\tau)$  is denoted by  $p(n, \tau)$ . In § 1 of this paper, some relations between the  $W$  and the  $P_m(\tau)$  are proved, and series are obtained for both  $P_m(\tau)$  and  $p(n, \tau)$  in terms of the  $W$ . One result is to show that  $P_m(\tau)$  decreases with  $m$  more rapidly than any negative power of  $m$ . The relations are quite general, and no assumption is made as to the Gaussian character of  $f(t)$ . It is assumed only that  $f(t)$  is statistically stationary; that  $-f(t)$  is statistically equivalent to  $f(t)$  (i.e. that  $f(t)$  is statistically symmetric with regard to the axis  $f = 0$ ); and that the various quantities defined actually exist.

In § 2 we obtain explicit expressions for the  $W$  in terms of the covariance function of  $f(t)$ , which is denoted by  $\psi(t)$ . By using some recent results of Kamat (1953) and Nabeya (1952) it is shown that  $W(+, -, -)$  can in fact be expressed in terms of known functionals of  $\psi(t)$ , a fact not apparently used previously. Moreover, in special cases the  $W$  can be evaluated for any integral  $n$ .

The results are applied in § 3 to the case when  $\psi(t)$  is itself a regular function at  $t = 0$  (implying the differentiability of  $f(t)$  up to all orders). Thus it is shown that for small intervals  $|t_i - t_j|$

$$W(+, -, +, \dots, (-)^{n-1}) \sim C_n \prod_{i < j} (t_j - t_i),$$

## INTERVALS BETWEEN ZEROS OF A RANDOM FUNCTION 559

where  $C_n$  is a constant independent of the  $t_i$ . Hence the asymptotic behaviour of  $P_m(\tau)$  and  $p(n, \tau)$  for small  $\tau$  is obtained. In particular it is shown that

$$P_m(\tau) \propto \tau^{\frac{1}{2}(m+2)(m+3)-2}$$

and

$$p(n, \tau) \propto \tau^{\frac{1}{2}n(n+1)}.$$

The power of  $\tau$  increases rapidly with  $m$  or  $n$ , indicating a strong 'mutual repulsion' of neighbouring zeros of  $f(t)$ .

A very interesting singular case is studied in §4, when the covariance function  $\psi(t)$ , instead of being regular at  $t = 0$ , has a finite discontinuity in the third derivative  $f'''(t)$ . This occurs, for example, whenever the spectral density of  $f(t)$  is proportional to (frequency)<sup>-4</sup> at high frequencies, and some examples have been studied experimentally by Favreau, Low & Pfeffer (1956). In contrast to the regular case, it is shown that, for small spacings  $|t_i - t_j|$ ,

$$W(+, -, +, \dots, (-)^{n-1}) \sim \frac{F(t_1, t_2, \dots, t_n)}{(t_3 - t_1)(t_4 - t_2) \dots (t_n - t_{n-2})} \quad (n \geq 2),$$

where  $F(t_1, t_2, \dots, t_n)$  is a function of the  $(t_i - t_j)$  lying between positive upper and lower bounds. It follows now that

$$P_m(\tau) \propto \tau^0 \quad (m \geq 0),$$

and

$$p(n, \tau) \propto \tau^2 \quad (n \geq 2),$$

so that, as  $\tau \rightarrow 0$ ,  $P_m(\tau)$  tends to a value  $P_m(0)$  independent of  $\tau$ .

Moreover, upper and lower bounds for  $P_m(0)$  can be found. Thus

$$1.147\alpha < P_0(0) < 1.218\alpha,$$

where  $\alpha = \psi'''(0+)/\{-6\psi''(0)\}$ . This result enables us to disprove rigorously the 'exponential hypothesis' of Favreau *et al.* (1956), whereby the distribution of intervals  $\tau$  for the function whose spectrum is  $(1 + \sigma^2)^{-2}$  is conjectured to be  $\pi^{-1} e^{-\tau/\pi}$ . For this would make

$$P_0(0) = 3\pi^{-1}\alpha = 0.955\alpha,$$

in contradiction to the above inequality.

Some rough work shows that a close approximation to  $P_0(\tau)$  is probably

$$P_0(0) \doteq 1.155\alpha.$$

Lastly in §5 we use the asymptotic expansions of  $P_m(\tau)$ ,  $p(n, \tau)$  and the  $W$  to compare the accuracy of the approximations proposed by Rice (1945), McFadden (1956, 1958), Ehrenfeld *et al.* (1958) and the present author (1958), especially in the neighbourhood of the origin. Numerical computation of the various approximations is also compared with experimental results of Blötekjaer (1958) and other authors.

### 1. GENERAL RELATIONS BETWEEN $P_m(\tau)$ , $p(n, \tau)$ AND $W(S)$

Let  $W(+, +, \dots, +) dt_1 \dots dt_n$  denote the probability that the function  $f(t)$  have an up-crossing (zero-crossing with positive gradient) in each of the intervals

$$(t_i, t_i + dt_i) \quad (i = 1, \dots, n);$$

by substituting a minus sign for any plus sign in  $W(+, +, \dots, +)$  we denote the corresponding probability for a down-crossing. Thus  $W(+, -, +, \dots, (-)^{n-1}) dt_1 \dots dt_n$  denotes the probability of alternate up-crossings and down-crossings in  $(t_i, t_i + dt_i)$ , the first being

an up-crossing. It will be seen later that the  $W$  may in many cases be evaluated explicitly in terms of the covariance function of the random process  $f(t)$ .

In this section we shall derive some quite general relations between  $P_m(\tau)$ ,  $p^{(n, \tau)}$  and the  $W$ , which are valid not only for Gaussian but for non-Gaussian processes.

### 1.1. Generalization of a result of McFadden

It was pointed out by McFadden (1958) that

$$\frac{W(+, -)}{W(+)} = P_0(\tau) + P_2(\tau) + P_4(\tau) + \dots, \quad (1.1.1)$$

$$\frac{W(+, +)}{W(+)} = P_1(\tau) + P_3(\tau) + P_5(\tau) + \dots, \quad (1.1.2)$$

where  $\tau = (t_2 - t_1)$ . The proof is very simple: the left-hand side of (1.1.1), when multiplied by  $dt_2$ , represents the probability that  $f$  has a down-crossing in  $(t_2, t_2 + dt_2)$  given that it has an up-crossing at  $t_1$ . This down-crossing must be either the next zero of  $f$ , or the next but two, and so on. These mutually exclusive events correspond to the individual terms on the right of (1.1.1); hence the identity. A similar argument proves (1.1.2).

Corresponding to (1.1.1) and (1.1.2) we may establish the following three relations:

$$\int_{t_1 < t_2 < t_3} \frac{W(+, -, +)}{W(+)} dt_2 = P_1(\tau) + 2P_3(\tau) + 3P_5(\tau) + \dots, \quad (1.1.3)$$

$$\int_{t_1 < t_2 < t_3} \frac{W(+, -, -)}{W(+)} dt_2 = P_2(\tau) + 2P_4(\tau) + 3P_6(\tau) + \dots, \quad (1.1.4)$$

$$\int_{t_1 < t_2 < t_3} \frac{W(+, +, +)}{W(+)} dt_2 = P_3(\tau) + 2P_5(\tau) + 3P_6(\tau) + \dots, \quad (1.1.5)$$

where  $\tau = (t_3 - t_1)$ . To prove (1.1.3), for example, consider

$$\frac{W(+, -, +)}{W(+)} dt_2 dt_3.$$

This represents the probability that  $f$  has a down-crossing in  $(t_2, t_2 + dt_2)$  and an up-crossing in  $(t_3, t_3 + dt_3)$ , given that it has an up-crossing at  $t_1$ . Now the up-crossing in  $(t_3, t_3 + dt_3)$  is either the second zero after  $t_1$  or the fourth or the sixth, etc. Suppose it is the  $(2r)$ th. Then the down-crossing in  $(t_2, t_2 + dt_2)$  is either the first zero after  $t_1$  or the third or the fifth, up to the  $(2r-1)$ th. But if the probabilities are each integrated with respect to  $t_2$  from  $t_1$  to  $t_3$  each gives precisely  $P_{2r-1} dt_3$ . The  $r$  possibilities together contribute  $rP_{2r-1} dt_3$ . Hence the series (1.1.3); and similarly for (1.1.4) and (1.1.5).

We now prove the following general theorem. Let  $S$  denote any sequence of  $n$  signs, plus or minus, the first sign being  $+$ , and let  $s$  denote the number of times that the sequence changes sign. (For example, if  $S = (+, -, +, -)$  then  $s = 3$ .) Then the expression  $X(S)$  defined by

$$X(S) \equiv \int \dots \int_{t_1 < t_2 < \dots < t_n} \frac{W(S)}{W(+)} dt_2 \dots dt_{n-1} \quad (1.1.6)$$

is the sum of the series

$$X(S) = \sum_{r=0}^{\infty} \binom{n-2+r}{r} P_{2n-3-s+2r}(\tau), \quad (1.1.7)$$

where  $\tau = (t_n - t_1)$  and  $\binom{p}{q}$  denotes the coefficient of  $x^q$  in  $(1+x)^p$ .

*Proof.* The integrand in (1.1.6) represents the probability that the  $(n-1)$  intervals  $(t_i, t_i + dt_i)$  ( $i = 2, 3, \dots, n$ ), contain zero-crossings of  $f(t)$  with gradients of the appropriate sign, given that  $f$  vanishes at  $t_1$ . Suppose that the last interval  $(t_n, t_n + dt_n)$  contains the  $(k+1)$ th zero of  $f$  after  $t_1$ . The remaining  $(n-2)$  intervals  $(t_2, t_2 + dt_2), \dots, (t_{n-1}, t_{n-1} + dt_{n-1})$  must contain  $(n-2)$  of the remaining  $k$  zeros between  $t_1$  and  $t_n$ . Each distinct way of choosing these  $(n-2)$  zeros contributes a term  $P_k(\tau)$  to the integral  $X(S)$ . Hence

$$X(S) = \sum_k c_k P_k(\tau),$$

where  $c_k$  denotes the number of distinct ways of choosing the sequence  $S$  from a sequence  $S'$  of  $(k+2)$  signs, alternately  $+$  and  $-$ , so that the first and last signs of  $S'$  correspond to those of  $S$ .

Now between each pair of successive signs of  $S$  that are both  $+$  (or both  $-$ ) there must be an odd number of signs of  $S$ . From the definition of  $s$  there are  $(n-1-s)$  such cases, and so

$$k+2 = n + (n-1-s) + 2r,$$

where  $r = 0, 1, 2, \dots$ . The remaining  $r$  pairs of signs of  $S'$  may occur anywhere in the  $(n-1)$  gaps between the signs of  $S$ . The number of distinct ways of disposing of these is

$$c_k = \binom{n-2+r}{r}.$$

On combining the last three equations we obtain (1.1.7).

It follows from the theorem that in the sequence  $S$  the only two relevant parameters are  $n$  and  $s$ . So we may write

$$X(S) = X_{n,s}. \quad (1.1.8)$$

The following special cases will be useful. When  $S = (+, +, \dots, +)$  then  $s = 0$  and

$$X_{n,0} = \sum_{r=0}^{\infty} \binom{n-2+r}{r} P_{2n-3+2r}. \quad (1.1.9)$$

When  $S = (+, -, -, \dots, -)$  then  $s = 1$  and

$$X_{n,1} = \sum_{r=0}^{\infty} \binom{n-2+r}{r} P_{2n-4+2r}. \quad (1.1.10)$$

When  $S = (+, -, +, \dots, (-)^{n-1})$  then  $s = (n-1)$  and

$$X_{n,n-1} = \sum_{r=0}^{\infty} \binom{n-2+r}{r} P_{n-2+2r}. \quad (1.1.11)$$

From these series there follows also a useful result on the order of magnitude of  $P_m(\tau)$  for large values of  $m$  (when  $\tau$  is fixed). For the binomial coefficient in each case is

$$\binom{n-2+r}{r} = \frac{(n-2+r)(n-3+r)\dots(1+r)}{(n-2)!} = O(r^{n-2})$$

as  $t$  tends to infinity. But if the series converges the individual terms must each tend to zero. So from (1.1.9), for example,

$$\lim_{r \rightarrow \infty} r^{n-2} P_{2n-3+2r} = 0.$$

In this expression the factor  $r^{n-2}$  may be replaced by  $(2n-3+2r)^{n-2}$  without altering the limit. So

$$\lim_{m \rightarrow \infty} m^{n-2} P_m = 0 \quad (1.1.12)$$

if  $m$  tends to infinity through the odd values; and similarly for the even values.

Thus  $P_m(\tau)$  tends to zero more rapidly than  $m^{-(n-2)}$ ,  $\tau$  being kept constant. Provided, then, that the  $X_{n,s}$  exist for all values of  $n$ , it follows that  $P_m(\tau)$  tends to zero more rapidly than any negative power of  $m$ .

### 1.2. Series for $P_m(\tau)$

Equations (1.1.1) to (1.1.5) may be written

$$\left. \begin{aligned} P_0 &= X(+, -) - (P_2 + P_4 + P_6 + \dots), \\ P_1 &= X(+, +) - (P_3 + P_5 + P_7 + \dots), \end{aligned} \right\} \quad (1.2.1)$$

and†

$$\left. \begin{aligned} P_1 &= X(+, -, +) - (2P_3 + 3P_5 + 4P_7 + \dots), \\ P_2 &= X(+, -, -) - (2P_4 + 3P_6 + 4P_8 + \dots), \\ P_3 &= X(+, +, +) - (2P_5 + 3P_7 + 4P_9 + \dots). \end{aligned} \right\} \quad (1.2.2)$$

Rice (1945) and McFadden (1958) neglected  $P_2, P_3, \dots$  and took as an approximation

$$\left. \begin{aligned} P_0 &\doteq X(+, -), \\ P_1 &\doteq X(+, +). \end{aligned} \right\} \quad (1.2.3)$$

However, by eliminating  $P_2$  and  $P_3$  from the right-hand side of (1.2.1) by means of equations (1.2.2) we have

$$\left. \begin{aligned} P_0 &= X(+, -) - X(+, -, -) + (P_4 + 2P_6 + 3P_8 + \dots), \\ P_1 &= X(+, +) - X(+, +, +) + (P_5 + 2P_7 + 3P_9 + \dots) \end{aligned} \right\} \quad (1.2.4)$$

so that a higher approximation, neglecting only  $P_4, P_5, \dots$  is given by

$$\left. \begin{aligned} P_0 &\doteq X(+, -) - X(+, -, -), \\ P_1 &\doteq X(+, +) - X(+, +, +). \end{aligned} \right\} \quad (1.2.5)$$

It is easy to show that these approximations are the first in an infinite sequence. Equations (1.1.10), which involve only the even  $P_m$ , may be solved for  $P_0$  by multiplying the first equation ( $n=2$ ) by 1, the next by  $-1$ , and so on up to  $n=N$ , and adding. On the right-hand side the coefficient of  $P_{2i}$ , when  $i \leq N$ , is

$$\binom{i}{0} - \binom{i}{1} + \binom{i}{2} - \dots + (-1)^i \binom{i}{i} = \begin{cases} 1 & (i=0), \\ 0 & (0 < i \leq N), \end{cases}$$

and when  $i > N$  it is

$$\binom{i}{0} - \binom{i}{1} + \binom{i}{2} - \dots + (-1)^N \binom{i}{N} = (-1)^{N+1} \binom{i-1}{N}.$$

Hence  $X_{2,1} - X_{3,1} + X_{4,1} - \dots + (-1)^N X_{N+2,1}$

$$= P_0 + (-1)^N \left[ P_{2N+2} + \binom{N+1}{1} P_{2N+4} + \binom{N+2}{2} P_{2N+6} + \dots \right]. \quad (1.2.6)$$

† The five equations (1.1.1) to (1.1.5), regarded as equations for the  $P_m$ , are not independent, for we have the identical relation

$$X(+, -, +) - X(+, +, +) = X(+, +).$$

Thus provided the expression in square brackets tends to zero we have

$$P_0 = X_{2,1} - X_{3,1} + X_{4,1} - \dots, \quad (1.2.7)$$

and similarly

$$P_1 = X_{2,0} - X_{3,0} + X_{4,0} - \dots \quad (1.2.8)$$

The approximations (1.2.3) and (1.2.5) correspond to the first two partial sums of these series.

Moreover, the remainder after  $(N+1)$  terms of the series is

$$(-1)^{N+1} \left[ P_{2N+2} + \binom{N+1}{1} P_{2N+4} + \binom{N+2}{2} P_{2N+6} + \dots \right] \quad (1.2.9)$$

which, since the  $P$  are all positive, has the same sign as  $(-1)^{N+1}$ . Hence the sums of the series (1.2.7) and (1.2.8) lie between any two successive partial sums.

The corresponding series for  $P_{2r}$  and  $P_{2r+1}$  ( $r \geq 0$ ) are found from (1.1.10) and (1.1.9) to be

$$\left. \begin{aligned} P_{2r} &= \sum_{i=0}^{\infty} (-1)^i \binom{r+i}{i} X_{r+2+i,1}, \\ P_{2r+1} &= \sum_{i=0}^{\infty} (-1)^i \binom{r+i}{i} X_{r+2+i,0} \end{aligned} \right\} \quad (1.2.10)$$

(the coefficients in the two series being identical). The solution of (1.1.11) for  $P_m$  is

$$P_m = (m+1) \sum_{i=0}^{\infty} (-1)^i \frac{(m+2i)!}{i!(m+i+1)!} X_{m+2+2i} \quad (1.2.11)$$

for all  $m \geq 0$ , where  $X_n$  is written more shortly for  $X_{n,n-1}$ . The solutions are valid provided each series is absolutely convergent.

### 1.3. Relation to a series of Rice

It is interesting to compare the series (1.2.7) for  $P_0(\tau)$  with one stated by Rice (1945, equation (3.4-11)). His series may be written

$$P_0 = Y_2 - Y_3 + Y_4 - \dots, \quad (1.3.1)$$

where

$$Y_n \equiv \frac{1}{(n-2)!} \int_{t_1}^{t_n} \dots \int_{t_1}^{t_n} \frac{W(\pm, \pm, \dots, \pm)}{W(\pm)} dt_2 \dots dt_{n-1} \quad (1.3.2)$$

and  $W(\pm, \pm, \dots, \pm) dt_1 \dots dt_n$  denotes the probability of a zero-crossing in each of the intervals  $(t_i, t_i + dt_i)$  irrespective of the sign of  $f'(t)$ .

Now from the point of view of calculation  $Y_n$  is of a similar complexity to  $X_{n,1}$ , since each involves an  $(n-2)$ -fold integration of a probability density such as  $W(S)$ . On the other hand in the series (1.3.1) the first term neglected after  $N$  terms is  $Y_{N+3}$ , which is of order  $P_{N+1}$  (see below), whereas the remainder after  $N$  terms in the series (1.2.7) is only of order  $P_{2N+2}$ . Clearly then (1.2.7) is more rapidly convergent than (1.3.1).

The reason for this difference is apparently that in deriving (1.2.7) we have made use of the continuity of  $f'(t)$  which implies that up-crossings and down-crossings follow one another alternately. In (1.3.1) no such property is used.

For completeness we now express  $Y_n$  in terms of the  $P_m(\tau)$ .



Since  $-f(t)$  is assumed statistically equivalent to  $f(t)$ , the integrand in (1.3.2) may be replaced by  $W(+, \pm, \pm, \dots, \pm)/W(+)$ ; and further by the symmetry of the integrand with respect to  $t_2, t_3, \dots, t_{n-1}$  we have

$$Y_n = \int \dots \int_{t_1 < t_2 < \dots < t_n} \frac{W(+, \pm, \pm, \dots, \pm)}{W(+)} dt_2 \dots dt_{n-1}.$$

Now  $W(+, \pm, \pm, \dots, \pm)$  can be considered as the sum of  $2^{n-1}$  expressions of the form  $W(S)$ , where  $S$  is a sequence of  $n$  signs such as was defined in § 1.1. Corresponding to any given value of  $s$  there are  $\binom{n-1}{s}$  such sequences  $S$ . Hence

$$Y_n = \sum_{s=0}^{n-1} \binom{n-1}{s} X_{n,s}.$$

But  $X_{n,s}$  or  $X(S)$  is given by (1.1.7). Thus

$$Y_n = \sum_{s=0}^{n-1} \sum_{r=0}^{\infty} \binom{n-1}{s} \binom{n-2+r}{r} P_{n-2+s+2r}.$$

The coefficient of  $P_{n-2+i}$  may be summed by comparing coefficients of  $x^i$  in the expansion of the identity

$$(1+x)^{n-1} (1-x^2)^{-(n-1)} \equiv (1-x)^{-(n-1)}$$

in powers of  $x$ . Hence

$$Y_n = \sum_{i=0}^{\infty} \binom{n-2+i}{i} P_{n-2+i}. \quad (1.3.3)$$

The first term in this series is  $P_{n-2}$ , which proves our statement concerning the order of magnitude of  $Y_{N+3}$  made above.

The solution of equations (1.3.3) for the  $P_m$  is

$$P_m = \sum_{i=0}^{\infty} (-1)^i \binom{m+1}{i} Y_{m+2+i} \quad (1.3.4)$$

of which (1.3.1) is the special case when  $m = 0$ .

#### 1.4. Series for $p(n, \tau)$

McFadden (1958) has shown that  $p(n, \tau)$  (the probability of exactly  $n$  zeros in the interval  $(t, t+\tau)$ ) is related to  $P_m(\tau)$  by the following set of equations:†

$$\left. \begin{aligned} p''(0, \tau) &= 2W(+)\, P_0, \\ p''(1, \tau) &= 2W(+)\, (P_1 - 2P_0), \\ p''(n, \tau) &= 2W(+)\, (P_n - 2P_{n-1} + P_{n-2}) \quad (n \geq 2), \end{aligned} \right\} \quad (1.4.1)$$

where a prime denotes differentiation with respect to  $\tau$ . On substitution for  $P_n, P_{n-1}, P_{n-2}$  from equations (1.2.11) we have in the general case

$$p''(n, \tau) = 2W(+)\, \sum_{i=0}^{\infty} C_{n,i} X_{n+i} \quad (1.4.2)$$

† The first of these relations is apparently due to Kohlenberg (1953).

where  $C_{n,0} = 1$

$$\left. \begin{aligned} C_{n,2r} &= (-1)^{r+1} (2r - n^2 + n) \frac{(n-2+2r)!}{r!(n+r)!} \quad (r \geq 1), \\ C_{n,2r+1} &= (-1)^{r+1} 2n \frac{(n-1+2r)!}{r!(n+r)!} \quad (r \geq 0). \end{aligned} \right\} \quad (1.4.3)$$

Now by definition

$$X_n = \int \dots \int_{t_1 < t_2 < \dots < t_n} \frac{W(+, -, +, \dots, (-)^{n-1})}{W(+)} dt_2 \dots dt_n \quad (1.4.4)$$

which is a function of  $\tau = (t_n - t_1)$ . Hence

$$X_n(\tau) = \frac{1}{W(+)} I_n''(\tau), \quad (1.4.5)$$

where 
$$I_n = \int \dots \int_{0 < t_1 < \dots < t_n < \tau} W(+, -, +, \dots, (-)^{n-1}) dt_1 \dots dt_n. \quad (1.4.6)$$

On substituting in (1.4.2) and integrating twice with respect to  $\tau$  from  $\tau = 0$  we have

$$p(n, \tau) = 2 \sum_{i=0}^{\infty} C_{n,i} I_{n+i}(\tau) \quad (1.4.7)$$

provided the constants of integration vanish. The first term in this expansion is  $2I_n$ .

Let  $R(t)$  denote the covariance function of the function  $\xi(t)$  which equals 1 when  $f(t) > 0$  and  $-1$  when  $f(t) < 0$ . Rice (1944) showed that

$$R(t) = p(0, \tau) - p(1, \tau) + p(2, \tau) - \dots$$

By differentiating twice and using equations (1.4.1) one obtains

$$R''(\tau) = 8W(+)(P_0 - P_1 + P_2 - \dots) \quad (1.4.8)$$

(McFadden 1958). From the first two equations of § 1.1 it follows that

$$R''(\tau) = 8[W(+, -) - W(+, +)]. \quad (1.4.9)$$

## 2. EVALUATION OF $W(S)$ FOR GAUSSIAN PROCESSES

We now specialize the discussion to the case when  $f(t)$  is Gaussian, and seek some explicit formulae for  $W(S)$  in terms of the covariance function of  $f(t)$ .

### 2.1. A general expression for $W(S)$

Consider first the probability  $W(+, +, \dots, +) dt_1 \dots dt_n$  that  $f(t)$  should have a zero up-crossing in each of the small intervals  $(t_i, t_i + dt_i)$  ( $i = 1, \dots, n$ ). For convenience write

$$f(t_i) = \xi_i, \quad f'(t_i) = \eta_i \quad (i = 1, \dots, n),$$

and let  $p(\xi_1, \dots, \xi_n; \eta_1, \dots, \eta_n)$  denote the joint probability density of the  $\xi_i$  and  $\eta_i$ . Thus

$$p(\xi_1, \dots, \xi_n; \eta_1, \dots, \eta_n) d\xi_1 \dots d\xi_n \cdot d\eta_1 \dots d\eta_n \quad (2.1.1)$$

is the probability that the  $\xi_i$  and  $\eta_i$  lie in given intervals  $(\xi_i, \xi_i + d\xi_i)$ ,  $(\eta_i, \eta_i + d\eta_i)$ . Now if  $f(t)$  has a zero-crossing in  $(t_i, t_i + dt_i)$ , with gradient  $\eta_i$ , then  $f(t_i)$  must lie in a small range of values of extent  $|\eta_i| dt_i$ . Thus to obtain the probability  $W(+, +, \dots, +) dt_1 \dots dt_n$  we replace

$d\xi_i$  in (2.1.1) by  $|\eta_i| dt_i$  and integrate over all positive values of the  $\eta_i$ . After dividing by  $dt_1 \dots dt_n$  we have

$$W(+, +, \dots, +) = \int_0^\infty \dots \int_0^\infty |\eta_1 \dots \eta_n| p(0, \dots, 0; \eta_1, \dots, \eta_n) d\eta_1 \dots d\eta_n. \quad (2.1.2)$$

For the covariance function of  $f(t)$  we write

$$\overline{f(t)f(t+\tau)} = \psi(\tau).$$

The function  $\psi(\tau)$  (or  $\psi_\tau$ ) is considered as given: it is the cosine transform of the spectral density of  $f(t)$ .

Then the covariance matrix of the  $2n$  variables  $\xi_1, \dots, \xi_n; \eta_1, \dots, \eta_n$  is

$$(\lambda_{ij}) = \begin{pmatrix} \psi_{11} & \dots & \psi_{1n} & \psi'_{11} & \dots & \psi'_{1n} \\ & & \vdots & \vdots & & \vdots \\ \psi_{n1} & \dots & \psi_{nn} & \psi'_{n1} & \dots & \psi'_{nn} \\ -\psi'_{11} & \dots & -\psi'_{1n} & -\psi''_{11} & \dots & -\psi''_{1n} \\ & & \vdots & \vdots & & \vdots \\ -\psi'_{n1} & \dots & -\psi'_{nn} & -\psi''_{n1} & \dots & -\psi''_{nn} \end{pmatrix}, \quad (2.1.3)$$

where  $\psi_{ij} = \psi(t_i - t_j)$  and a prime denotes differentiation.

By the Gaussian hypothesis we have

$$p(\xi_1, \dots, \xi_n; \eta_1, \dots, \eta_n) = \frac{1}{(2\pi)^n \Delta^{\frac{1}{2}}} \exp \left[ -\frac{1}{2} \sum_{i,j=1}^{2n} L_{ij} \xi_i \xi_j \right], \quad (2.1.4)$$

where  $\xi_{n+i} = \eta_i$  and  $\Delta = |(\lambda_{ij})|$ ,  $(L_{ij}) = (\lambda_{ij})^{-1}$ . (2.1.5)

Substitution in (2.1.2) gives

$$W(+, +, \dots, +) = \frac{1}{(2\pi)^n \Delta^{\frac{1}{2}}} \int_0^\infty \dots \int_0^\infty |\eta_1 \dots \eta_n| \exp \left[ -\frac{1}{2} \sum_{i,j=1}^n L_{n+i, n+j} \eta_i \eta_j \right] d\eta_1 \dots d\eta_n.$$

The summation in the last equation involves only the last  $n$  rows and columns of  $(L_{ij})$ . It is convenient to denote the inverse of this matrix by  $(\mu_{ij})$

$$(\mu_{ij}) = \begin{pmatrix} L_{n+1, n+1} & \dots & L_{n+1, 2n} \\ \vdots & & \vdots \\ L_{2n, n+1} & \dots & L_{2n, 2n} \end{pmatrix}^{-1}. \quad (2.1.7)$$

By Jacobi's theorem the  $(r, s)$ th element of this matrix is the bordered determinant

$$\mu_{r,s} = \begin{vmatrix} \psi_{11} & \dots & \psi_{1n} & \psi'_{1s} \\ \vdots & & \vdots & \vdots \\ \psi_{n1} & \dots & \psi_{nn} & \psi'_{ns} \\ -\psi'_{r1} & \dots & -\psi'_{rn} & -\psi''_{rs} \end{vmatrix} \div D, \quad (2.1.8)$$

where

$$D = \begin{vmatrix} \psi_{11} & \dots & \psi_{1n} \\ \vdots & & \vdots \\ \psi_{n1} & \dots & \psi_{nn} \end{vmatrix}. \quad (2.1.9)$$

The determinant of  $(\mu_{ij})$  is given by

$$|(\mu_{ij})| = \Delta/D. \quad (2.1.10)$$

$(\mu_{ij})$  will be recognized as the covariance matrix of  $(\eta_1, \dots, \eta_n)$  given that

$$\xi_1 = \xi_2 = \dots = \xi_n = 0.$$

For, if  $p(\eta_1, \dots, \eta_n | \xi_1, \dots, \xi_n)$  denotes the conditional distribution of  $(\eta_1, \dots, \eta_n)$  for given values of  $(\xi_1, \dots, \xi_n)$  we have

$$p(\eta_1, \dots, \eta_n | \xi_1, \dots, \xi_n) = \frac{p(\xi_1, \dots, \xi_n; \eta_1, \dots, \eta_n)}{p(\xi_1, \dots, \xi_n)},$$

where  $p(\xi_1, \dots, \xi_n)$  is the distribution of  $(\xi_1, \dots, \xi_n)$  only

$$p(\xi_1, \dots, \xi_n) = \frac{1}{(2\pi)^{\frac{1}{2}n} D^{\frac{1}{2}}} \exp \left[ -\frac{1}{2} \sum_{i,j=1}^n M_{ij} \xi_i \xi_j \right],$$

where  $(M_{ij})$  is the inverse of  $(\psi_{ij})$ . Hence when the  $\xi_i$  vanish we have, using (2.1.10),

$$p(\eta_1, \dots, \eta_n | 0, \dots, 0) = \frac{1}{(2\pi)^{\frac{1}{2}n} |(\mu_{ij})|^{\frac{1}{2}}} \exp \left[ -\frac{1}{2} \sum_{i,j=1}^n L_{n+i, n+j} \eta_i \eta_j \right] = Z(\boldsymbol{\eta}, \boldsymbol{\mu}),$$

say. With this notation (2.1.6) may be written

$$W(+, +, \dots, +) = \frac{1}{(2\pi)^{\frac{1}{2}n} D^{\frac{1}{2}}} \int_0^\infty \dots \int_0^\infty |\eta_1 \dots \eta_n| Z(\boldsymbol{\eta}, \boldsymbol{\mu}) d\eta_1 \dots d\eta_n. \quad (2.1.11)$$

It is convenient to introduce the normalized covariance matrix  $(v_{ij})$  whose  $(i, j)$ th element is

$$v_{ij} = \frac{\mu_{ij}}{(\mu_{ii} \mu_{jj})^{\frac{1}{2}}}. \quad (2.1.12)$$

Then on writing

$$\zeta_i = (\mu_{ii})^{-\frac{1}{2}} \eta_i$$

in equation (2.1.11) so that  $(v_{ij})$  is the covariance matrix of the new variables  $\zeta_i$ , we have

$$W(+, +, \dots, +) = \frac{(\mu_{11} \mu_{22} \dots \mu_{nn})^{\frac{1}{2}}}{(2\pi)^{\frac{1}{2}n} D^{\frac{1}{2}}} J_n, \quad (2.1.13)$$

where

$$J_n = \int_0^\infty \dots \int_0^\infty \zeta_1 \dots \zeta_n Z(\boldsymbol{\zeta}, \boldsymbol{v}) d\zeta_1 \dots d\zeta_n. \quad (2.1.14)$$

Now  $Z(\boldsymbol{\zeta}, \boldsymbol{v})$  is the ordinary normal probability density in the  $n$  variables  $\zeta_i$ , with covariance matrix  $(v_{ij})$ . Since the diagonal elements have been normalized (by equation (2.1.12)),  $J_n$  is a function only of the off-diagonal element  $v_{ij}$  ( $i \neq j$ ).

Suppose that one of the zeros in the sequences (say the  $k$ th zero) is to be a down-crossing instead of an up-crossing. Then in calculating the corresponding probability density  $W$  we need only to take the range of integration of  $\eta_k$  in (2.1.2) from  $-\infty$  to 0 instead of 0 to  $\infty$ . Equivalently, we may simply reverse the sign of the  $(n+k)$ th row and column of  $L_{ij}$ , and hence the  $k$ th row and column of  $(\mu_{ij})$  and of  $(v_{ij})$ .

Hence to find  $W(+, -, +, \dots, (-)^{n-1})$ , in which each alternate zero-crossing is a down-crossing, we have to multiply  $L_{n+i, n+j}$  by  $(-1)^{i+j}$  and hence also multiply  $\mu_{ij}$  and  $v_{ij}$  by  $(-1)^{i+j}$ .

## 2.2. The cases $n = 1, 2$ and 3

The case  $n = 1$  is trivial, for then  $Z(\boldsymbol{\zeta}, \boldsymbol{v})$  is the normal distribution for a single variate and

$$J_1 = \int_0^\infty \zeta \frac{e^{-\frac{1}{2}\zeta^2}}{(2\pi)^{\frac{1}{2}}} d\zeta = \frac{1}{(2\pi)^{\frac{1}{2}}}.$$

Since in (2.1.8)  $\psi'_{11}$  vanishes we have  $\mu_{11} = -\psi''_{11} = -\psi''_0$  and so from (2.1.13)

$$W(+)=\left(\frac{\mu_{11}}{2\pi D}\right)^{\frac{1}{2}} J_1 = \frac{1}{2\pi} \left(\frac{-\psi''_0}{\psi_0}\right)^{\frac{1}{2}} \quad (2.2.1)$$

as is well known (Kac 1943).

Also well known is the case  $n = 2$ , when

$$J_2 = \frac{1}{2\pi} [(1-\nu_{12}^2)^{\frac{1}{2}} + \nu_{12} \cos^{-1}(-\nu_{12})]$$

(the angle being chosen so as to lie between 0 and  $\pi$ ). This gives

$$W(+,+) = \frac{(\mu_{11}\mu_{22})^{\frac{1}{2}}}{4\pi^2(\psi_0^2 - \psi_{12}^2)^{\frac{1}{2}}} [(1-\nu_{12}^2)^{\frac{1}{2}} + \nu_{12} \cos^{-1}(-\nu_{12})] \quad (2.2.2)$$

(Rice 1945, §3.10). By changing the sign of  $\nu_{12}$  we have

$$W(+,-) = \frac{(\mu_{11}\mu_{22})^{\frac{1}{2}}}{4\pi^2(\psi_0^2 - \psi_{12}^2)^{\frac{1}{2}}} [(1-\nu_{12}^2)^{\frac{1}{2}} - \nu_{12} \cos^{-1}(\nu_{12})]. \quad (2.2.3)$$

Not so well known is the case  $n = 3$ . However,  $J_3$  may be derived from some integrals calculated quite recently by Nabeya (1952) and Kamat (1953). One obtains

$$J_3 = \frac{1}{(2\pi)^{\frac{3}{2}}} [ |(\nu_{ij})|^{\frac{1}{2}} + (s_1\alpha_1 + s_2\alpha_2 + s_3\alpha_3) ],$$

where

$$s_1 = \cos^{-1} \frac{\nu_{31}\nu_{12} - \nu_{23}}{(1-\nu_{31}^2)^{\frac{1}{2}}(1-\nu_{23}^2)^{\frac{1}{2}}}, \quad \alpha_1 = \nu_{31}\nu_{12} + \nu_{23}.$$

( $s_2, s_3$ , etc., are obtained by cyclic permutation of the  $\nu_{ij}$ ). These angles are also to be taken in the range  $(0, \pi)$ . So from (2.1.13) we find

$$\left. \begin{aligned} W(+,+,+) &= \frac{(\mu_{11}\mu_{22}\mu_{33})^{\frac{1}{2}}}{8\pi^3 D^{\frac{3}{2}}} [ |(\nu_{ij})|^{\frac{1}{2}} + (s_1\alpha_1 + s_2\alpha_2 + s_3\alpha_3) ], \\ W(+,-,+) &= \frac{(\mu_{11}\mu_{22}\mu_{33})^{\frac{1}{2}}}{8\pi^3 D^{\frac{3}{2}}} [ |(\nu_{ij})|^{\frac{1}{2}} + (s_1 - \pi)\alpha_1 + s_2\alpha_2 + (s_3 - \pi)\alpha_3 ], \\ W(+,-,-) &= \frac{(\mu_{11}\mu_{22}\mu_{33})^{\frac{1}{2}}}{8\pi^3 D^{\frac{3}{2}}} [ |(\nu_{ij})|^{\frac{1}{2}} + s_1\alpha_1 + (s_2 - \pi)\alpha_2 + (s_3 - \pi)\alpha_3 ]. \end{aligned} \right\} \quad (2.2.4)$$

### 2.3. General values of $n$

When  $n > 3$ , the integral  $J_n$  cannot be expressed in terms of known functions, in general. However, two particular cases in which this is possible may be stated here for later use.

First, if the covariances  $\nu_{ij}$  all vanish when  $i \neq j$ , then  $\zeta_1, \dots, \zeta_n$  are statistically independent variates and

$$J_n = \left[ \int_0^\infty \zeta \frac{e^{-\frac{1}{2}\zeta^2}}{(2\pi)^{\frac{1}{2}}} d\zeta \right]^n = \frac{1}{(2\pi)^{\frac{1}{2}n}}$$

giving

$$W = \frac{1}{(2\pi)^n} \frac{(\mu_{11} \dots \mu_{nn})^{\frac{1}{2}}}{D^{\frac{1}{2}}}. \quad (2.3.1)$$

Secondly, if all the covariances  $\nu_{ij}$  are unity, then  $\zeta_1, \dots, \zeta_n$  all reduce to the same Gaussian variate, giving

$$J_n = \int_0^\infty \zeta^n \frac{e^{-\frac{1}{2}\zeta^2}}{(2\pi)^{\frac{1}{2}}} d\zeta = \frac{1}{(2\pi)^{\frac{1}{2}}} 2^{\frac{1}{2}(n-1)} \left(\frac{n-1}{2}\right)!$$

Hence

$$W = \frac{1}{2\pi^{\frac{1}{2}(n+1)}} \left(\frac{n-1}{2}\right)! \frac{(\mu_{11} \dots \mu_{nn})^{\frac{1}{2}}}{D^{\frac{1}{2}}}. \quad (2.3.2)$$

3. ASYMPTOTIC EXPANSIONS NEAR THE ORIGIN:  $\psi$  REGULAR

In this section we evaluate the probabilities defined earlier, for small time intervals  $\tau$ . It will be assumed that the covariance function  $\psi(t)$  is regular at the origin

$$\psi(t) = \psi_0 + \frac{\psi_0''}{2!} t^2 + \frac{\psi_0^{iv}}{4!} t^4 + \dots$$

(coefficients of the odd powers vanish, since  $\psi(t)$  is an even function of  $(t)$ ).

3.1. Expansion of  $W(S)$ 

Our first object is to obtain a multiple power series in the  $t_i$  for the probability density  $W(S)$ . Since  $W$  depends only on the covariance matrix  $(\lambda_{ij})$  of equation (2.1.3), it is evidently a function of the time differences  $(t_i - t_j)$ . We shall see that the leading terms in  $W$  are homogeneous and of degree  $\frac{1}{2}n(n-1)$  in the  $(t_i - t_j)$ .

We use the following lemma: if  $F(x)$  is any function of  $x$  regular at  $x = 0$ , then the leading term in the expansion of

$$\begin{vmatrix} F(x_1 + y_1) & \dots & F(x_1 + y_n) \\ \vdots & & \vdots \\ F(x_n + y_1) & \dots & F(x_n + y_n) \end{vmatrix}$$

in powers of the  $x_i$  and  $y_i$  is

$$\begin{vmatrix} F(0) & F'(0) & \dots & F^{(n-1)}(0) \\ F'(0) & F''(0) & \dots & F^{(n)}(0) \\ \vdots & \vdots & & \vdots \\ F^{(n-1)}(0) & F^{(n)}(0) & \dots & F^{(2n-2)}(0) \end{vmatrix} \frac{\prod_{i < j} (x_j - x_i) (y_j - y_i)}{[1! 2! \dots (n-1)!]^2}.$$

To prove this, first express each term as a Taylor series in the  $x_i$

$$F(x_i + y_i) = F(y_i) + x_i F'(y_i) + \frac{x_i^2}{2!} F''(y_i) + \dots$$

Subtract the first row of the determinant from the remaining rows, taking out the factors  $(x_2 - x_1), \dots, (x_n - x_1)$ ; then subtract the second row from the rows beneath it, taking out the factors  $(x_3 - x_2), \dots, (x_n - x_2)$ ; and so on. In the result write

$$F^{(i)}(y_j) = F^{(i)}(0) + y_j F^{(i+1)}(0) + \frac{y_j^2}{2!} F^{(i+2)}(0) + \dots$$

and proceed similarly with the columns. This gives the result.

Setting  $F = \psi$ ,  $x_i = t_i$  and  $y_i = -t_i$  in the lemma we find, from (2.1.9),

$$D \sim D_n \frac{\prod_{i < j} (t_j - t_i)^2}{[1! 2! \dots (n-1)!]^2}, \quad (3.1.1)$$

where

$$D_m = (-1)^{\frac{1}{2}m(m-1)} \begin{vmatrix} \psi_0 & \psi_0' & \dots & \psi_0^{(m-1)} \\ \psi_0' & \psi_0'' & \dots & \psi_0^{(m)} \\ \vdots & \vdots & & \vdots \\ \psi_0^{(m-1)} & \psi_0^{(m)} & \dots & \psi_0^{(2m-2)} \end{vmatrix}. \quad (3.1.2)$$

It will be noticed that since the odd derivatives of  $\psi$  all vanish, every alternate element of  $D_m$  is zero, so that  $D_m$  may be factorized into two determinants

$$D_m = (-1)^{\frac{1}{2}m(m-1)} \begin{vmatrix} \psi_0 & \psi_0'' & \cdots \\ \psi_0'' & \psi_0^{(4)} & \cdots \\ \vdots & \vdots & \ddots \end{vmatrix} \times \begin{vmatrix} \psi_0'' & \psi_0^{iv} & \cdots \\ \psi_0^{iv} & \psi_0^{(6)} & \cdots \\ \vdots & \vdots & \ddots \end{vmatrix}.$$

By a similar method we may evaluate the leading term in  $\mu_{rs}$  (equation (2.1.8)). One finds in fact

$$\mu_{rs} \sim \frac{D_{n+1}}{D_n} \frac{\prod_{i \neq r} (t_r - t_i) \prod_{j \neq s} (t_s - t_j)}{(n!)^2}. \quad (3.1.3)$$

Hence

$$\nu_{rs} = \frac{\mu_{rs}}{(\mu_{rr} \mu_{ss})^{\frac{1}{2}}} \sim \pm 1$$

according as  $\mu_{rs}$  is positive or negative. Now since  $(\psi_{ij})$  is a positive-definite matrix its determinant  $D$  must be positive, so that by (3.1.1)  $D_m$  is positive also. On the other hand when  $t_1 < t_2 < \dots < t_n$  the product  $\prod_{i \neq r} (t_r - t_i)$  has the same sign as  $(-1)^{r+s}$ , and so  $\mu_{rs}$  has the same sign as  $(-1)^{r+s}$ . It follows that

$$\nu_{rs} \sim (-1)^{r+s}.$$

Now to calculate  $W(+, -, +, \dots, (-)^{n-1})$  we recall that  $\nu_{rs}$  was to be multiplied by  $(-1)^{r+s}$ . The elements of the corresponding covariance matrix thus all become equal to unity, in the limit, and so (2.3.2) applies. On substituting for  $\mu_{ii}$  and  $D$  we find

$$W(+, -, +, \dots, (-)^{n-1}) \sim C_n \prod_{i < j} (t_j - t_i), \quad (3.1.4)$$

where

$$C_n = \frac{1! 2! \dots (n-1)!}{2\pi^{\frac{1}{2}(n+1)} (n!)^n} \left(\frac{n-1}{2}\right)! \left(\frac{D_{n+1}^n}{D_n^{n+1}}\right)^{\frac{1}{2}}. \quad (3.1.5)$$

In particular when  $n = 1$  and  $2$  we have the known results

$$W(+) = \frac{1}{2\pi} \frac{D_2^{\frac{1}{2}}}{D_1} \quad (3.1.6)$$

and

$$W(+, -) \sim \frac{1}{16\pi} \frac{D_3}{D_2^{\frac{3}{2}}} (t_2 - t_1) \quad (3.1.7)$$

(cf. Rice 1945, § 3.4).

### 3.2. Evaluation of $P_m(\tau)$

The integral  $I_n$  defined by (1.4.6) can now be evaluated. We use the identity

$$\int \dots \int_{0 < t_1 < t_2 < \dots < t_n < \tau} \prod_{i < j} (t_j - t_i) dt_1 \dots dt_n = \frac{[1! 2! 3! \dots (n-1)!]^2}{1! 3! 5! \dots (2n-1)!} \tau^{\frac{1}{2}n(n+1)}, \quad (3.2.1)$$

a proof of which is given, for example, by Mehta (1960). From (1.4.6) and (3.1.4) we have then

$$I_n \sim \frac{[1! 2! 3! \dots (n-1)!]^2}{1! 3! 5! \dots (2n-1)!} C_n \tau^{\frac{1}{2}n(n+1)}, \quad (3.2.2)$$

which is of order  $\tau^{\frac{1}{2}n(n+1)}$ . From this it follows that

$$X_n = \frac{1}{W(+)} I_n'' = O(\tau^{\frac{1}{2}n(n+1)-2}). \quad (3.2.3)$$

Since the power of  $\tau$  increases with  $n$ , one sees that  $P_m(\tau)$  is given asymptotically by the first term in the series (1.2.11), i.e.

$$P_m(\tau) \sim X_{m+2}. \quad (3.2.4)$$

From the last three equations and (3.1.5) we have

$$P_m(\tau) \sim \frac{[1!2!3! \dots (m+1)!]^2 C_{m+2}}{1!3!5! \dots (2m+3)! C_1} \frac{d^2}{d\tau^2} \tau^{\frac{1}{2}(m+2)(m+3)}. \quad (3.2.5)$$

In particular

$$\left. \begin{aligned} P_0(\tau) &\sim \frac{C_2}{C_1} \tau = \frac{1}{8} \frac{D_1^{\frac{1}{2}} D_3}{D_2^2} \tau, \\ P_1(\tau) &\sim \frac{C_3}{C_1} \frac{\tau^4}{6} = \frac{1}{648\pi} \frac{D_1^{\frac{1}{2}} D_4^{\frac{3}{2}}}{D_2^{\frac{1}{2}} D_3^2} \tau^4, \end{aligned} \right\} \quad (3.2.6)$$

in agreement with Rice (1945) and Palmer (1956), respectively. In general we have

$$P_m(\tau) = O(\tau^{\frac{1}{2}(m+2)(m+3)-2}), \quad (3.2.7)$$

a power of  $\tau$  that increases very rapidly with  $m$ .

### 3.3. Evaluation of $p(n, \tau)$

When  $n = 0$  and 1 we have trivially

$$p(0, \tau) \sim 1 \quad (3.3.1)$$

and

$$p(1, \tau) \sim 2W(+)\tau = \frac{1}{\pi} \frac{D_2^{\frac{1}{2}}}{D_1} \tau \quad (3.3.2)$$

from (3.1.6).

When  $n \geq 2$ , both  $I_n$  and  $I'_n$  vanish at  $\tau = 0$ , and therefore (1.4.5) is valid provided both  $p(n, 0)$  and  $p'(n, 0)$  are zero. Both conditions are satisfied if we assume  $p(n, \tau) = O(\tau^{1+\epsilon})$ , where  $\epsilon > 0$ . In the series (1.4.7) the terms  $I_{n+i}(\tau)$  are proportional to increasing powers of  $\tau$  and hence  $p(n, \tau)$  is given asymptotically by the first term

$$p(n, \tau) \sim 2I_n, \quad (3.3.3)$$

or on substitution from (3.2.2)

$$p(n, \tau) \sim 2 \frac{[1!2!3! \dots (n-1)!]^2}{1!3!5! \dots (2n-1)!} C_n \tau^{\frac{1}{2}n(n+1)} \quad (n \geq 2). \quad (3.3.4)$$

For example

$$\left. \begin{aligned} p(2, \tau) &\sim \frac{1}{3} C_2 \tau^3 = \frac{1}{48\pi} \frac{D_3}{D_2^{\frac{3}{2}}} \tau^3, \\ p(3, \tau) &\sim \frac{1}{96} C_3 \tau^6 = \frac{1}{19440\pi^2} \frac{D_4^{\frac{3}{2}}}{D_3^2} \tau^6. \end{aligned} \right\} \quad (3.3.5)$$

Since  $p(2, \tau)$  is of order  $\tau^3$  and not  $\tau^2$  we see that neighbouring zeros of  $f(t)$  are not independent of one another. The effect may be called a 'mutual repulsion' of the zeros. It is connected with the property, seen in the previous section, that  $P_0(\tau) \rightarrow 0$  as  $\tau \rightarrow 0$ , that is to say, small intervals  $\tau$  are unlikely.

Moreover, as  $n$  increases, so the power of  $\tau$  in  $p(n, \tau)$  increases very rapidly.

A heuristic argument for the rapidly increasing power of  $\tau$  may be given as follows. If  $f(t)$  is to have  $n$  zeros in  $(t, t+\tau)$  then by Rolle's theorem  $f'(t)$  must have at least  $(n-1)$  zeros in the same interval, and further  $f''(t)$  must have at least  $(n-2)$  zeros, and so on, till finally  $f^{(n-1)}(t)$  must have at least one zero in the interval. Therefore (assuming the existence



of  $f^{(n)}$   $f^{(n-1)}$  must be of order  $\tau$  throughout the interval, and by integration  $f^{(n-2)}, f^{(n-3)}, \dots, f$  must be of order  $\tau^2, \tau^3, \dots, \tau^n$ , respectively. That is to say,  $f^{(n-1)}, f^{(n-2)}, \dots, f$  at some fixed point in the interval, must lie within ranges  $\delta f^{(n-1)}, \delta f^{(n-2)}, \dots, \delta f$  of order  $\tau, \tau^2, \dots, \tau^n$ , respectively. The probability of such an event is of order

$$\delta f^{(n-1)} \delta f^{(n-2)} \dots \delta f \approx O(\tau^{1+2+\dots+n}) \approx O(\tau^{\frac{1}{2}n(n+1)}).$$

#### 4. ASYMPTOTIC EXPANSIONS NEAR THE ORIGIN: A SINGULAR CASE

We shall now seek expansions at the origin in a very interesting singular case. Instead of the Taylor series for  $\psi(t)$  (equation (3.1)), suppose now that  $\psi(t)$  has an expansion of the form

$$\psi(t) = \psi_0 + \frac{\psi_0''}{2!} t^2 + \frac{\psi_0'''}{3!} |t|^3 + \dots \quad (4.1)$$

In other words, the third derivative of  $\psi(t)$  possesses a finite discontinuity at the origin.† Some examples of such functions were studied experimentally by Favreau *et al.* (1956), and McFadden (1958) has considered them theoretically. They occur whenever the spectral density of  $f(t)$  is of order (frequency)<sup>-4</sup> at high frequencies.

##### 4.1. Expansion of $W(S)$

If the procedure of §3.1 is attempted it is found that altogether fewer factors can be extracted from the determinants. For example, to evaluate  $D$ , defined by (2.1.9), we begin by subtracting row  $(n-1)$  from row  $n$  of the determinant, then row  $(n-2)$  from row  $(n-1)$ , and so on, in turn extracting the factors  $(t_n - t_{n-1}), (t_{n-1} - t_{n-2}), \dots, (t_2 - t_1)$ ; and similarly for the columns. The process is then repeated as far as row 2 only, and without extracting any factors. The leading term in the determinant is then seen to be

$$D \sim -\psi_0 \psi_0'' c^{n-2} (\tau_1 \tau_2 \dots \tau_{n-1})^2 |\mathbf{A}| \quad (n > 2), \quad (4.1.1)$$

where we have written  $\frac{1}{6} \psi_0''' \approx c, \quad t_{i+1} - t_i = \tau_i,$  (4.1.2)

and where  $\mathbf{A}$  is the  $(n-2) \times (n-2)$  square matrix

$$\mathbf{A} \approx \begin{pmatrix} 4(\tau_1 + \tau_2) & 2\tau_2 & 0 & \dots & 0 \\ 2\tau_2 & 4(\tau_2 + \tau_3) & 2\tau_3 & \dots & 0 \\ 0 & 2\tau_3 & 4(\tau_3 + \tau_4) & \dots & 0 \\ \vdots & \vdots & \dots & \vdots & \vdots \\ 0 & 0 & 0 & \dots & 4(\tau_{n-2} + \tau_{n-1}) \end{pmatrix}. \quad (4.1.3)$$

When  $n = 2$ ,  $|\mathbf{A}|$  is replaced by unity in equation (4.1.1).

Instead of calculating  $(\mu_{rs})$  directly, it is rather more convenient to determine first  $(L_{n+i, n+j})$ . Now  $(L_{ij})$  is the reciprocal of the covariance matrix  $(\lambda_{ij})$ , given by (2.1.3). The determinant of  $(\lambda_{ij})$  is found, by a process similar to the above, to be

$$\Delta \sim -\psi_0 \psi_0'' c^{2n-2} (\tau_1 \tau_2 \dots \tau_{n-1})^2 |\mathbf{E}|, \quad (4.1.4)$$

where  $\mathbf{E}$  denotes the  $(2n-2) \times (2n-2)$  square matrix made up as follows:

$$\mathbf{E} = \begin{pmatrix} \mathbf{A} & \mathbf{C} \\ \mathbf{C}^* & \mathbf{B} \end{pmatrix}$$

† The existence of  $\psi''(t)$  is sufficient to ensure the joint distribution of  $f(t)$  and  $f'(t)$  as in §2. If the expansion of  $\psi(t)$  contains a term in  $|t|$  then the mean density of zeros no longer exists in the usual sense. Such a case was considered by Siegert (1951). See also Rice (1958, §9).

in which  $\mathbf{A}$  is given by (4.1.3),  $\mathbf{B}$  is the  $n \times n$  matrix

$$\mathbf{B} = \begin{pmatrix} 4\tau_1 & -2\tau_1 & 0 & 0 & \dots & 0 \\ -2\tau_1 & 4\tau_1 & 0 & 0 & \dots & 0 \\ 0 & 0 & 4\tau_2 & 0 & \dots & 0 \\ 0 & 0 & 0 & 4\tau_3 & \dots & 0 \\ \vdots & \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & 0 & \dots & 4\tau_{n-1} \end{pmatrix}$$

and  $\mathbf{C}$  is the  $(n-2) \times n$  matrix

$$\mathbf{C} = \begin{pmatrix} -2\tau_1 & 4\tau_1 & 2\tau_2 & 0 & \dots & 0 & 0 & 0 \\ 0 & 0 & 4\tau_2 & 2\tau_3 & \dots & 0 & 0 & 0 \\ 0 & 0 & 0 & 4\tau_3 & \dots & 0 & 0 & 0 \\ \vdots & \vdots & \vdots & \vdots & \ddots & \vdots & \vdots & \vdots \\ 0 & 0 & 0 & 0 & \dots & 4\tau_{n-3} & 2\tau_{n-2} & 0 \\ 0 & 0 & 0 & 0 & \dots & 0 & 4\tau_{n-2} & 2\tau_{n-1} \end{pmatrix}$$

( $\mathbf{C}^*$  denotes the transpose of  $\mathbf{C}$ ). By subtracting the  $(n-1+i)$ th row of  $\mathbf{E}$  from the  $i$ th row ( $i = 1, 2, \dots, (n-2)$ ), and similarly for the columns we find

$$|\mathbf{E}| = 12^{n-1}(\tau_1\tau_2 \dots \tau_{n-1})^2$$

and so 
$$\Delta \sim 12^{n-1}(-\psi_0\psi_0'')c^{2n-2}(\tau_1\tau_2 \dots \tau_{n-1})^4. \quad (4.1.5)$$

The quantities  $\Delta L_{n+i, n+j}$  involve the cofactors of the last  $n$  rows and columns of  $(\lambda_{ij})$  and hence of  $\mathbf{E}$ . Hence we find

$$(L_{n+i, n+j}) \sim \frac{1}{3c} \begin{pmatrix} u_1 & \frac{1}{2}u_1 & 0 & \dots & 0 & 0 \\ \frac{1}{2}u_1 & (u_1+u_2) & \frac{1}{2}u_2 & \dots & 0 & 0 \\ 0 & \frac{1}{2}u_2 & (u_2+u_3) & \dots & 0 & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & 0 & 0 & \dots & (u_{n-2}+u_{n-1}) & \frac{1}{2}u_{n-1} \\ 0 & 0 & 0 & \dots & \frac{1}{2}u_{n-1} & u_{n-1} \end{pmatrix}, \quad (4.1.6)$$

where 
$$u_i = \frac{1}{\tau_i} = \frac{1}{t_{i+1} - t_i}. \quad (4.1.7)$$

#### 4.2. The cases $n = 1, 2$ and $3$

In the special case  $n = 1$  the above expansions do not apply, but the well-known result

$$W(+)=\frac{1}{2\pi}\left(\frac{-\psi_0''}{\psi_0}\right)^{\frac{1}{2}} \quad (4.2.1)$$

is easily derived as in § 2.

When  $n = 2$  we have

$$(L_{n+i, n+j}) \sim \frac{1}{3c} \begin{pmatrix} u_1 & \frac{1}{2}u_1 \\ \frac{1}{2}u_1 & u_1 \end{pmatrix}$$

and so by inversion

$$(\mu_{ij}) \sim c \begin{pmatrix} 4\tau_1 & -2\tau_1 \\ -2\tau_1 & 4\tau_1 \end{pmatrix}$$

giving  $\nu_{12} \sim -\frac{1}{2} = \cos^{-1}(\frac{2}{3}\pi)$ . Thus from (2.2.2) and (2.2.3)

$$\left. \begin{aligned} W(+, +) &\sim \frac{1}{\pi^2} \left( \frac{\sqrt{3}}{2} - \frac{\pi}{6} \right) \frac{c}{(-\psi_0 \psi_0'')^{\frac{1}{2}}}, \\ W(+, -) &\sim \frac{1}{\pi^2} \left( \frac{\sqrt{3}}{2} + \frac{\pi}{3} \right) \frac{c}{(-\psi_0 \psi_0'')^{\frac{1}{2}}}. \end{aligned} \right\} \quad (4.2.2)$$

When  $n = 3$  we have  $(L_{n+i, n+j}) \sim \frac{1}{3c} \begin{pmatrix} u_1 & \frac{1}{2}u_1 & 0 \\ \frac{1}{2}u_1 & (u_1 + u_2) & \frac{1}{2}u_2 \\ 0 & \frac{1}{2}u_2 & u_2 \end{pmatrix}$

and so  $(\mu_{ij}) \sim \frac{c}{\tau_1 + \tau_2} \begin{pmatrix} \tau_1(3\tau_1 + 4\tau_2) & -2\tau_1\tau_2 & \tau_1\tau_2 \\ -2\tau_1\tau_2 & 4\tau_1\tau_2 & -2\tau_1\tau_2 \\ \tau_1\tau_2 & -2\tau_1\tau_2 & (4\tau_1 + 3\tau_2)\tau_2 \end{pmatrix}$ .

Thus  $\nu_{23}, \nu_{31}, \nu_{12} = -\left(\frac{\tau_2}{3\tau_1 + 4\tau_2}\right)^{\frac{1}{2}}, \left(\frac{\tau_1\tau_2}{(3\tau_1 + 4\tau_2)(4\tau_1 + 3\tau_2)}\right)^{\frac{1}{2}}, -\left(\frac{\tau_1}{4\tau_1 + 3\tau_2}\right)^{\frac{1}{2}}$ .

Writing for short  $\left. \begin{aligned} \frac{\tau_1}{\tau_1 + \tau_2} = \frac{t_2 - t_1}{t_3 - t_1} = x, \\ \frac{\tau_2}{\tau_1 + \tau_2} = \frac{t_3 - t_2}{t_3 - t_1} = y, \end{aligned} \right\} \quad (4.2.3)$

so that  $x + y = 1$ , we find from (2.2.4)

$$W(S) = \frac{1}{8\pi^3} \frac{c}{(-\psi_0 \psi_0'')^{\frac{1}{2}}} \frac{1}{t_3 - t_1} Q^{(S)}, \quad (4.2.4)$$

where

$$\left. \begin{aligned} Q^{(+, +, +)} &= 3 - \frac{(5-2x)x^{\frac{1}{2}}}{(4-x)^{\frac{1}{2}}} \cos^{-1}\left(\frac{x^{\frac{1}{2}}}{2}\right) + \pi(xy)^{\frac{1}{2}} - \frac{(5-2y)y^{\frac{1}{2}}}{(4-y)^{\frac{1}{2}}} \cos^{-1}\left(\frac{y^{\frac{1}{2}}}{2}\right), \\ Q^{(+, -, +)} &= 3 + \frac{(5-2x)x^{\frac{1}{2}}}{(4-x)^{\frac{1}{2}}} \cos^{-1}\left(\frac{-x^{\frac{1}{2}}}{2}\right) + \pi(xy)^{\frac{1}{2}} + \frac{(5-2y)y^{\frac{1}{2}}}{(4-y)^{\frac{1}{2}}} \cos^{-1}\left(\frac{-y^{\frac{1}{2}}}{2}\right), \\ Q^{(+, -, -)} &= 3 - \frac{(5-2x)x^{\frac{1}{2}}}{(4-x)^{\frac{1}{2}}} \cos^{-1}\left(\frac{x^{\frac{1}{2}}}{2}\right) - \pi(xy)^{\frac{1}{2}} + \frac{(5-2y)y^{\frac{1}{2}}}{(4-y)^{\frac{1}{2}}} \cos^{-1}\left(\frac{-y^{\frac{1}{2}}}{2}\right). \end{aligned} \right\} \quad (4.2.5)$$

These three functions are plotted in figure 1.  $Q^{(+, +, +)}$  and  $Q^{(+, -, +)}$  are symmetrical about the mid-point  $x = \frac{1}{2}$ , as would be expected, whereas  $Q^{(+, -, -)}$  is asymmetrical.

The probability density of a down-crossing at  $t_2$  given up-crossings at  $t_1$  and  $t_3$  is proportional to  $Q^{(+, -, +)}$ . Figure 1 then shows that the probability density is a maximum when  $t_2$  is mid-way between the two ends of the interval. On the other hand the curve for  $Q^{(+, +, +)}$  shows that given up-crossings at both  $t_1$  and  $t_3$  the probability density of an up-crossing at an intermediate point  $t_2$  is fairly insensitive to the position of  $t_2$ . At the mid-point, the density is actually a minimum.

#### 4.3. General values of $n$

Exact expressions do not appear to exist in general, but upper and lower bounds for  $W(S)$  may be obtained in the following way.

In equation (2.1.6) write

$$\eta_i = \frac{x_i}{(L_{ii})^{\frac{1}{2}}}, \quad l_{ij} = \frac{L_{ij}}{(L_{ii}L_{jj})^{\frac{1}{2}}}. \quad (4.3.1)$$

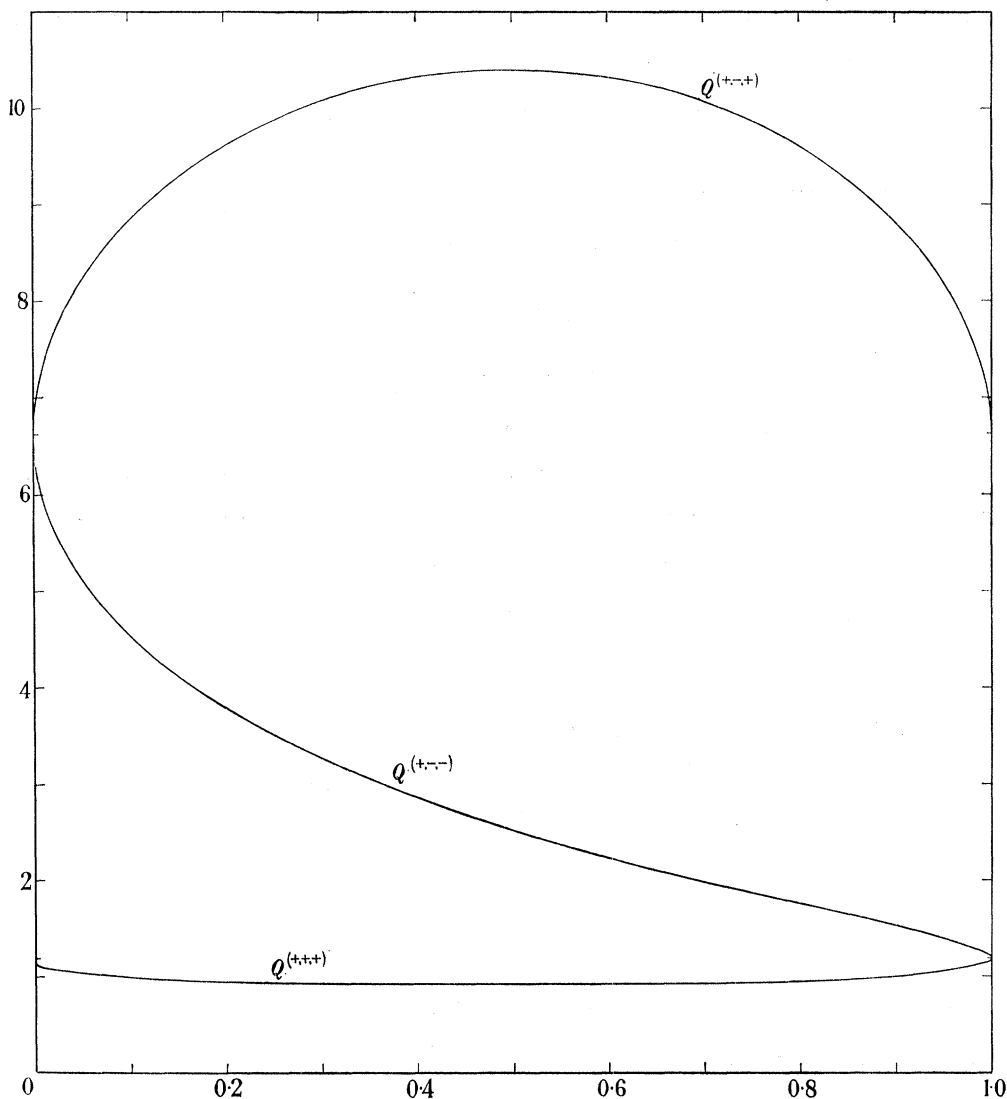


FIGURE 1. Graphs of  $Q^{(+,+,+)}$ ,  $Q^{(+,-,+)}$  and  $Q^{(+,-,-)}$ .

This gives 
$$W(+, +, \dots, +) = \frac{1}{(2\pi)^n \Delta^{\frac{1}{2}}} \frac{\Phi(\mathbf{1})}{L_{11} L_{22} \dots L_{nn}}, \tag{4.3.2}$$

where 
$$\Phi(\mathbf{1}) = \int_0^\infty \dots \int_0^\infty x_1 x_2 \dots x_n \exp\left[-\frac{1}{2} \sum_{i,j=1}^n l_{ij} x_i x_j\right]. \tag{4.3.3}$$

Using the asymptotic formulae for  $\Delta$  and  $L_{ij}$  (equations (4.1.4) and (4.1.6)) we have

$$W(+, +, \dots, +) \sim \frac{1}{(2\pi)^n} \frac{3^n}{12^{\frac{1}{2}(n-1)}} \frac{c}{(-\psi_0 \psi_0'')^{\frac{1}{2}}} \frac{\Phi(\mathbf{1})}{(\tau_1 + \tau_2)(\tau_2 + \tau_3) \dots (\tau_{n-2} + \tau_{n-1})} \tag{4.3.4}$$

and

$$(l_{ij}) = \begin{pmatrix} 1 & \frac{1}{2} \left(\frac{u_1}{u_1 + u_2}\right)^{\frac{1}{2}} & 0 & \dots & 0 \\ \frac{1}{2} \left(\frac{u_1}{u_1 + u_2}\right)^{\frac{1}{2}} & 1 & \frac{1}{2} \left(\frac{u_2}{u_1 + u_2}\right)^{\frac{1}{2}} \left(\frac{u_2}{u_2 + u_3}\right)^{\frac{1}{2}} & \dots & 0 \\ 0 & \frac{1}{2} \left(\frac{u_2}{u_1 + u_2}\right)^{\frac{1}{2}} \left(\frac{u_2}{u_2 + u_3}\right)^{\frac{1}{2}} & 1 & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \dots & 1 \end{pmatrix}. \tag{4.3.5}$$

It is now easy to find bounds for  $\Phi$ . For since  $u_i > 0$ , all the elements of  $(l_{ij})$  adjacent to the diagonal lie between 0 and  $\frac{1}{2}$ . Therefore, the  $x_i$  being non-negative,

$$\sum_i x_i^2 \leq \sum_{i,j} l_{ij} x_i x_j \leq \sum_i x_i^2 + \sum x_i x_{i+1}. \quad (4.3.6)$$

In the right-hand inequality substitute

$$x_i x_j \leq \frac{1}{2}(x_i^2 + x_j^2)$$

giving

$$\sum x_i^2 \leq \sum_{i,j} l_{ij} x_i x_j \leq 2 \sum_i x_i^2.$$

Thus from (4.3.3)

$$1 \geq \Phi \geq 1/2^n. \quad (4.3.7)$$

These bounds may now be substituted in (4.3.4).

For  $W(S)$ , when the signs of  $S$  are not all +, one or more of the  $l_{ij}$  may be reversed in sign. Hence the left-hand inequality in (4.3.6) is not valid but may be replaced by

$$\sum_i x_i^2 - \sum_i x_i x_{i+1} \leq \sum_{i,j} x_i x_{i+1}.$$

Now it can be shown that

$$\lambda \sum_i x_i^2 \leq \sum_i x_i^2 - \sum_i x_i x_{i+1},$$

where  $\lambda$  is the smallest root of the equation

$$\begin{vmatrix} (1-\lambda) & -\frac{1}{2} & 0 & \dots & 0 \\ -\frac{1}{2} & (1-\lambda) & -\frac{1}{2} & \dots & 0 \\ 0 & -\frac{1}{2} & (1-\lambda) & \dots & 0 \\ \vdots & \vdots & \vdots & \dots & \vdots \\ 0 & 0 & 0 & \dots & (1-\lambda) \end{vmatrix} = 0,$$

that is to say

$$\lambda = 2 \sin^2 \{ \pi/2(n+1) \}.$$

Since, then, in all possible cases

$$\lambda \sum_i x_i^2 \leq \sum_{i,j} l_{ij} x_i x_j \leq 2 \sum_i x_i^2$$

we have

$$\frac{1}{2^n \sin^{2n} \{ \pi/2(n+1) \}} \geq \Phi \geq \frac{1}{2^n}. \quad (4.3.6)$$

Our general result then is that

$$W(S) \sim \frac{1}{(2\pi)^n} \left( \frac{3}{4} \right)^{\frac{1}{2}n} \frac{c}{(-\psi_0 \psi_0'')^{\frac{1}{2}}} \frac{2\sqrt{3} \Phi}{(t_3 - t_1)(t_4 - t_2) \dots (t_n - t_{n-2})}, \quad (4.3.7)$$

where  $\Phi$  is a function of the  $t_i$  lying between the bounds (4.3.6).

#### 4.4. Asymptotic behaviour of $P_m(\tau)$

By the mean-value theorem for integrals, the integral  $X_n$  of (1.4.4) can be expressed as

$$X_n \sim \frac{1}{(2\pi)^{n-1}} \left( \frac{3}{4} \right)^{\frac{1}{2}n} \frac{c}{-\psi_0''} 2\sqrt{3} \Phi' K_n, \quad (4.4.1)$$

where

$$K_n = \int \dots \int_{t_1 < t_2 < \dots < t_n} \frac{dt_2 \dots dt_{n-1}}{(t_3 - t_1)(t_4 - t_2) \dots (t_n - t_{n-2})} \quad (4.4.2)$$

## INTERVALS BETWEEN ZEROS OF A RANDOM FUNCTION 577

and  $\Phi'$  is some value of  $\Phi$  within the bounds (4.3.6). It can be shown that  $K_n$  is finite, and since the denominator of the integrand is homogeneous and of degree  $(n-2)$ ,  $K_n$  is independent of  $(t_n - t_1)$ , or  $\tau$ . In fact when  $n = 2, 3, 4, 5, \dots$

$$K_n = 1, 1, \frac{1}{6}\pi^2, \frac{1}{3}\pi^2, \dots \quad (4.4.3)$$

Thus, as  $\tau \rightarrow 0$ ,  $X_n$  tends asymptotically to a positive value independent of  $\tau$ . From the expansion (1.2.11) it now appears that, for each value of  $m$ ,  $P_m(\tau)$  tends asymptotically to a limiting value  $P_m(0)$ .

This behaviour of  $P_m(\tau)$  is in marked contrast to the corresponding behaviour when  $\psi(t)$  is a regular function. Then, as was seen in §3.2,  $P_m(\tau)$  is proportional to an increasing power of  $\tau$  as  $m$  increases. A further discussion will be given in connexion with  $p(n, \tau)$  (see §4.8). Meanwhile, however, we shall establish some close inequalities for  $P_m(0)$  when  $m = 1, 2$ .

4.5. Approximations to  $P_m(0)$ 

From the results of §4.2 we may evaluate  $X(S)$  explicitly when  $n = 2$  and 3. Thus from (4.2.1) and (4.2.2) we have

$$\left. \begin{aligned} X(+, -) &\sim \left(\frac{\sqrt{3}}{\pi} + \frac{2}{3}\right)\alpha \\ X(+, +) &\sim \left(\frac{\sqrt{3}}{\pi} - \frac{1}{3}\right)\alpha \end{aligned} \right\} \quad (4.5.1)$$

where we have written for short  $\alpha = \frac{c}{-\psi''_0} = \frac{\psi'''_{0+}}{-6\psi''_0}$ . (4.5.2)

Further, on integrating the expressions (4.2.4) with respect to  $t_2$  over  $t_1 < t_2 < t_3$  we find

$$\left. \begin{aligned} X(+, -, +) &\sim \left(\frac{3}{8\pi^2} + \frac{2}{\sqrt{3}\pi} - \frac{47}{288}\right)\alpha, \\ X(+, -, -) &\sim \left(\frac{3}{8\pi^2} + \frac{1}{2\sqrt{3}\pi} - \frac{17}{288}\right)\alpha, \\ X(+, +, +) &\sim \left(\frac{3}{8\pi^2} - \frac{1}{\sqrt{3}\pi} + \frac{49}{288}\right)\alpha. \end{aligned} \right\} \quad (4.5.3)$$

The identity  $X(+, -, +) - X(+, +, +) = X(+, +)$  can be readily verified.

Equations (1.2.1) and (1.2.2) give, in the limit when  $\tau \rightarrow 0$

$$\left. \begin{aligned} P_0 &= 1.217\,996\alpha - (P_2 + P_4 + P_6 + \dots), \\ P_1 &= 0.217\,996\alpha - (P_3 + P_5 + P_7 + \dots) \end{aligned} \right\} \quad (4.5.4)$$

and 
$$\left. \begin{aligned} P_2 &= 0.070\,856\alpha - (2P_4 + 3P_6 + 4P_8 + \dots), \\ P_3 &= 0.024\,358\alpha - (2P_5 + 3P_7 + 4P_9 + \dots) \end{aligned} \right\} \quad (4.5.5)$$

whence also 
$$\left. \begin{aligned} P_0 &= 1.147\,139\alpha + (P_4 + 2P_6 + 3P_8 + \dots), \\ P_1 &= 0.193\,638\alpha + (P_5 + 2P_7 + 3P_9 + \dots). \end{aligned} \right\} \quad (4.5.6)$$

Hence the inequalities 
$$\left. \begin{aligned} 1.147\alpha &< P_0(0) < 1.218\alpha, \\ 0.193\alpha &< P_1(0) < 0.218\alpha, \end{aligned} \right\} \quad (4.5.7)$$

and 
$$\left. \begin{aligned} 0 &< P_2(0) < 0.071\alpha, \\ 0 &< P_3(0) < 0.025\alpha. \end{aligned} \right\} \quad (4.5.8)$$

4.6. *Disproof of the 'exponential hypothesis'*

We now apply the inequalities of the previous section to a particular case which was studied experimentally by Favreau *et al.* (1956). This is the Gaussian process  $f(t)$  whose spectral density is given by

$$E \propto 1/(1 + \sigma^2)^2, \quad (4.6.1)$$

where  $\sigma =$  frequency. The covariance function  $\psi_\tau$ , being the cosine transform of  $E$ , has the form

$$\psi(t) \propto (1 + |t|) e^{-|t|} = 1 - \frac{1}{2}t^2 + \frac{1}{3}|t|^3 - \dots$$

and so is of the form (4.1).

The experimental results showed that the distribution of zero-crossing intervals  $P_0(\tau)$  was quite close to a negative exponential. Since the mean of the distribution must be

$$\frac{1}{2W(+)} = \frac{1}{\pi} \left( \frac{\psi_0}{-\psi_0''} \right)^{\frac{1}{2}} = \frac{1}{\pi},$$

the only possible exponential law is

$$P_0(\tau) = (1/\pi) e^{-\tau/\pi}$$

which makes  $P_0(0) = 1/\pi$ , or, since

$$\alpha = \psi_0''' / (-6\psi_0'') = \frac{1}{3}$$

in this case,

$$P_0(0) = 3\alpha/\pi = 0.955 \dots \alpha. \quad (4.6.2)$$

McFadden (1956) doubted the conjecture but was unable to disprove it (McFadden's assumption that  $p''(n, \tau) = 0$  when  $n \geq 4$  is actually incorrect), since the only inequalities then available to him were the right-hand inequalities of (4.5.7). However, the left-hand inequality

$$1.147\alpha < P_0(0)$$

is definitely contradictory to (4.6.2). Thus the exponential hypothesis is disproved.

It may be pointed out that because of certain limitations in the experiments (indicated by Favreau *et al.*), the value of  $P_0(\tau)$  is liable to be underestimated at the small values of  $\tau$ ; so that it is not surprising that the experiments suggested a too low value of  $P_0(0)$ .

4.7. *Further estimates of  $P_m(0)$* 

If  $P_4, P_5, \dots$  are neglected in equations (4.5.5) and (4.5.6), the resulting estimates of  $P_0, P_1, P_2$  and  $P_3$  show that  $P_2/P_1$  and  $P_3/P_2$  are about equal to  $1/3$ . Now in §1.1 it was shown that  $P_m(\tau)$  tended to zero with  $m$  more rapidly than any negative power of  $m$ . It is consistent with this result to conjecture that the ratio  $P_{m+1}/P_m$  tends to a constant value. If we take roughly

$$\left. \begin{aligned} P_4(0) &\doteq \frac{1}{3}P_3(0) \doteq 0.006\alpha, \\ P_5(0) &\doteq \frac{1}{3}P_4(0) \doteq 0.002\alpha, \\ P_6(0) &\doteq \frac{1}{3}P_5(0) \doteq 0.001\alpha, \end{aligned} \right\} \quad (4.7.1)$$

then on substituting in equations (4.5.5) and (4.5.6) we find as possibly closer approximations

$$\left. \begin{aligned} P_0(0) &\doteq 1.155\alpha, \\ P_1(0) &\doteq 0.196\alpha, \\ P_2(0) &\doteq 0.055\alpha, \\ P_3(0) &\doteq 0.017\alpha. \end{aligned} \right\} \quad (4.7.2)$$

4.8. Asymptotic behaviour of  $p(n, \tau)$ 

As in § 3.3 we have 
$$p(0, \tau) \sim 1, \quad p(1, \tau) \sim \frac{1}{\pi} \left( \frac{-\psi_0''}{\psi_0} \right) \tau. \quad (4.8.1)$$

When  $n \geq 2$  we have from (1.4.1) by integration

$$p(n, \tau) = \int_0^\tau d\tau' \int_0^{\tau'} d\tau'' 2W(+)[P_n(\tau'') - 2P_{n-1}(\tau'') + P_{n-2}(\tau'')]$$

provided that  $p(n, 0)$  and  $p'(n, 0)$  are both zero. This will be satisfied provided

$$p(n, \tau) = O(\tau^{1+\epsilon}), \quad \text{where } \epsilon > 0.$$

Now we have seen earlier that in the singular case  $P_m(\tau)$  tends to a positive value  $P_m(0)$  as  $\tau \rightarrow 0$ . Hence by integration

$$p(n, \tau) \sim W(+)[P_n(0) - 2P_{n-1}(0) + P_{n-2}(0)]\tau^2 \quad (4.8.2)$$

as  $\tau \rightarrow 0$ .

From (4.5.8) and (4.5.9) we have the strict inequalities

$$\left. \begin{aligned} 0.711\beta\tau^2 < p(2, \tau) < 0.903\beta\tau^2, \\ 0.051\beta\tau^2 < p(3, \tau) < 0.243\beta\tau^2, \end{aligned} \right\} \quad (4.8.3)$$

where

$$\beta = W(+)\alpha = \frac{1}{12\pi} \frac{\psi_{0+}'''}{(-\psi_0\psi_0'')^{\frac{1}{2}}}. \quad (4.8.4)$$

The rough estimates (4.7.2) would yield

$$\left. \begin{aligned} p(2, \tau) &\doteq 0.818\beta\tau^2, \\ p(3, \tau) &\doteq 0.103\beta\tau^2. \end{aligned} \right\} \quad (4.8.5)$$

Equation (4.8.2) shows that, in contrast to the regular case,  $p(n, \tau)$  is of order  $\tau^2$  for all values of  $n$  greater than or equal to 2; there is no longer a strong mutual repulsion of the zeros.

Again, a heuristic argument suggests that this result is not unreasonable. Since  $\psi_\tau''$  has no continuous derivative at the origin, the second derivative of  $f$  is, in this case, non-existent almost everywhere (cf. Bartlett 1955, chapter 5) and the first derivative  $f'(t)$  may be expected to behave like a random-walk process in which the standard deviation of

$$[f'(t_1) - f'(t_2)]$$

increases like  $|t_1 - t_2|^{\frac{1}{2}}$  for small time-differences. Now in the fixed interval  $(0, \tau)$ , if  $f$  has two or more zeros,  $f'$  has at least one. So  $f'$  is of order  $\tau^{\frac{1}{2}}$  in the interval while  $f$ , by integration, is of order  $\tau^{\frac{3}{2}}$ . That is to say  $f$  and  $f'$  lie within intervals  $\delta f$  and  $\delta f'$  of order  $\tau^{\frac{3}{2}}$  and  $\tau^{\frac{1}{2}}$ , respectively. Since the joint probability density of  $f$  and  $f'$  at some fixed point  $t$  in the interval exists by hypothesis it follows that  $p(n, \tau)$  is of order

$$\delta f \cdot \delta f' = O(\tau^{\frac{3}{2}}\tau^{\frac{1}{2}}) = O(\tau^2).$$

One consequence of (4.8.2) is that, given the existence of two zeros in the interval  $(0, \tau)$ , the probability of  $(n-2)$  further zeros in the same interval is of order  $\tau^2/\tau^2 = 1$ . Roughly speaking, we may say that the first two zeros serve to 'pin down' the function  $f$  and its derivative so that the probability of any further number of zeros in the interval is



finite, no matter how short the interval is. However, the probability density of, say, a third zero lying somewhere between the first two depends upon the situation of the third zero relative to the first two, as was seen from the curves of figure 1.

### 5. A COMPARISON OF DIFFERENT APPROXIMATIONS TO $P_0(\tau)$

In the following we shall compare the accuracy of the approximations suggested by Rice (1945), McFadden (1956, 1958), Ehrenfeld *et al.* (1958), and Longuet-Higgins (1958; this paper is referred to as (I)), with the approximations suggested in the present paper. Discussion is purposely restricted to those methods of approximation on the basis of which numerical computation has been, or readily could be, carried out.

Two different aspects of the approximations are first considered: (a) their accuracy for small values of  $\tau$ , both in the regular and singular case of §§ 3 and 4, and (b) their accuracy for large values of  $\tau$ . The results are tabulated in table 1.

Then the 'narrow spectrum' approximation is considered in § 5.8, and lastly the approximations are compared numerically with experimental results obtained by analogue methods when the spectrum of  $f(t)$  has certain ideal forms.

#### 5.1. Rice's approximation (1945)

This has been used as a starting point for several of the later approximations. It is

$$P_0(\tau) \doteq \frac{W(+, -)}{W(+)} = X(+, -), \quad (5.1.1)$$

in our notation. The right-hand side, being the first term in the series (1.2.7) may also be written as  $P_0^{(1)}$ , where  $P_0^{(N)}$  is the sum of  $N$  terms. As we have seen, the calculation of  $W(+, -)$  involves the evaluation of a bivariate normal integral.

From equation (1.1.1) the error in  $P_0^{(1)}$  is equal to

$$P_2 + P_4 + P_6 + \dots, \quad (5.1.2)$$

which is always positive. Thus  $P_0^{(1)}$  always exceeds  $P_0$ .

(a) *Small values of  $\tau$ .* In the *regular* case the highest term in the remainder is

$$P_2 \sim \frac{1}{280}(C_4/C_1)\tau^8. \quad (5.1.3)$$

Thus  $P_0^{(1)}$  is correct to order  $\tau^7$  near the origin. In the *singular* case equations (4.7.1) and (4.7.2) give

$$P_2 + P_4 + P_6 + \dots = 0.062\alpha, \quad (5.1.4)$$

an error of about 5%.

(b) *Large values of  $\tau$ .* When  $\tau$  is so large that  $f(t)$  and  $f(t+\tau)$  are uncorrelated then we have

$$P_0^{(1)} = \frac{W(+, -)}{W(+)} \sim W(-) = W(+). \quad (5.1.5)$$

$P_0$ , on the other hand, must tend to zero, in order that  $\int P_0(\tau) d\tau$  shall converge. Thus (5.1.5) represents also the error in  $P_0^{(1)}$ .

#### 5.2. The approximation $P_0^{(2)}$

The approximation discussed in the present paper, namely

$$P_0^{(2)} = \frac{W(+, -)}{W(+)} - \int_{t_1 < t_2 < t_3} \frac{W(+, -, -)}{W(+)} dt_2 = X(+, -) - X(+, -, -), \quad (5.2.1)$$

## INTERVALS BETWEEN ZEROS OF A RANDOM FUNCTION 581

appears as a natural second approximation to  $P_0$ . Its evaluation involves the single integration of  $W(+, -, -)/W(+)$ , which, as we have seen in § 2.2, is expressible in terms of known functions. Higher approximations

$$P_0^{(N)} = X_{2,1} - X_{3,1} + X_{4,1} - \dots (-)^{N+1} X_{N+1,1}$$

will each involve additional integrations, in general.

Equation (1.2.4) shows that the error in  $P_0^{(2)}$  is equal to

$$-(P_4 + 2P_6 + 3P_8 + \dots) \quad (5.2.2)$$

which is always negative. Thus  $P_0^{(2)}$  is always a lower bound for  $P_0$ .

(a) *Small values of  $\tau$ .* In the regular case the highest term in the remainder is

$$-P_4 \sim -\frac{8}{21 \cdot (11)!} \frac{C_6}{C_1} \tau^{19}. \quad (5.2.3)$$

Thus  $P_0^{(2)}$  is correct to order  $\tau^{18}$  near the origin. In the singular case equations (4.7.1) give

$$-(P_4 + 2P_6 + \dots) \doteq -0.008\alpha, \quad (5.2.4)$$

an error of 0.7%.

It is clear that near the origin this approximation leaves little to be desired.

(b) *Large values of  $\tau$ .* Asymptotically we have

$$\frac{W(+, -, -)}{W(+)} \sim W(+)^2$$

and hence

$$P_0^{(2)} \sim -W(+)^2 \tau \quad (5.2.5)$$

which is  $O(\tau)$  at infinity. It is clear that the approximation fails radically for large values of  $\tau$ .

Indeed it will be seen generally that the approximation of  $P_0$  by  $P_0^{(N)}$  is analogous, for large  $\tau$ , to the approximation of  $e^{-x}$  by a finite number of terms of its power series; the convergence of the approximation is non-uniform over  $(0, \infty)$ .

### 5.3. The 'multiply conditioned' approximations

In § 3.10 of his original paper (1945) Rice suggested that the approximation (5.1.1) might be improved by including in the probability density  $W(+, -)$  the condition that  $f(t)$  be positive at one or more given points of the range  $(t_1, t_2)$ . The inclusion of just one extra point leads to a threefold normal integral that can be expressed in closed form. The inclusion of more than one point leads to fourfold and higher integrals.

The suggestion was taken up by Ehrenfeld *et al.* (1958), who refer to such approximations as 'multiply conditioned' approximations. Thus (5.1.1) is denoted by MC-0; with one condition at the mid-point of the range the approximation is MC-1, and so on. Clearly all such approximations are, like MC-0, upper bounds for the true value  $P_0$ .

(a) *Small values of  $\tau$ .* The error in MC-1 is just equal to the probability density of a down-crossing at  $t = \tau$  plus a zero in  $(0, \tau)$ , given that  $f(0) = 0$  and  $f(\frac{1}{2}\tau) > 0$ . Now if  $f(\frac{1}{2}\tau) > 0$ , there must be at least two zeros in the interval  $(0, \frac{1}{2}\tau)$  and/or at least two zeros in  $(\frac{1}{2}\tau, \tau)$ . If we ignore  $p(5, \tau)$ ,  $p(6, \tau)$ , etc., relative to  $p(4, \tau)$ , the error is clearly

$$2 \iint_{0 < t_2 < t_3 < \frac{1}{2}\tau} \frac{W(+, -, +, -)}{W(+)} dt_2 dt_3. \quad (5.3.1)$$

In the regular case, the neglect of  $p(5, \tau)$ , etc., is justified by § 3.3, and we have from (3.1.4)

$$\frac{W(+, -, +, -)}{W(+)} \sim \frac{C_4}{C_1} t_2 t_3 \tau (t_3 - t_2) (\tau - t_2) (\tau - t_3).$$

Substituting in (5.3.1) we find for the error

$$\frac{39}{8!} \frac{C_4}{C_1} \tau^8. \quad (5.3.2)$$

This may be compared with (5.1.3). Clearly the order of the error is the same as in MC-0, but is less by the ratio

$$\frac{39 \times 280}{8!} = \frac{13}{48}.$$

In the *singular* case the limiting value of MC-1 as  $\tau \rightarrow 0$  was calculated by Rice (1958) in the case of the spectrum  $(1 + \sigma^2)^{-2}$  (cf. § 4.6 above), with the result

$$\text{MC-1} \sim \frac{1.251}{\pi} = 1.195\alpha,$$

since  $\alpha = 1/3$  in this case. From (4.7.2) the error appears to be

$$1.195\alpha - 1.155\alpha = 0.040\alpha \quad (5.3.2)$$

or about 3% of  $P_0(0)$ . Compared with MC-0, the error is reduced by about one-third.

In the case of the higher multiply-conditioned solutions, if the subintervals of  $(0, \tau)$  are denoted by  $(t^{(i)}, t^{(i+1)})$  (with  $t^{(1)} = 0$ ) then the expression corresponding to (5.3.1) is

$$\sum_i \iint_{t^{(i)} < t_2 < t_3 < t^{(i+1)}} \frac{W(+, -, +, -)}{W(+)} dt_2 dt_3.$$

Since  $W$  is of order  $\tau^6$  it follows that the error is always of order  $\tau^8$ .

(b) *Large values of  $\tau$ .* When the interval  $\tau$  is sufficiently large, the sign of  $f(\frac{1}{2}\tau)$  becomes independent of the other conditions, and the probability of  $f(\frac{1}{2}\tau)$  being positive is one-half. Hence

$$\text{MC-1} \sim \frac{1}{2} \text{MC-0} \sim \frac{1}{2} W(+). \quad (5.3.3)$$

Thus the error is reduced relative to MC-0 by one-half.

Generally, if the  $N$  'conditioned' points in the multiply conditioned approximation MC- $N$  are spaced so that their separation tends to infinity with  $\tau$ , then

$$\text{MC-}N \sim \frac{1}{2^N} \text{MC-0} \sim \frac{1}{2^N} W(+). \quad (5.3.4)$$

#### 5.4. McFadden's first approximation

McFadden (1956) gave the following approximation to  $P_0(\tau)$ , valid for small intervals  $\tau$

$$P_0(\tau) \doteq \frac{R''(\tau)}{8W(\tau)} = F(\tau),$$

say. Here  $R''(\tau)$  denotes the correlation function of the 'clipped' form of  $f(t)$ , defined in § 1.4. From (1.4.8),

$$F(\tau) = P_0 - P_1 + P_2 - P_3 + \dots = X(+, -) - X(+, +). \quad (5.4.1)$$

## INTERVALS BETWEEN ZEROS OF A RANDOM FUNCTION 583

The approximation is equivalent to neglecting  $P_1, P_2, \dots$  in the above series; thus it is actually of a lower order of accuracy than  $P_0^{(1)}$ .

In the Gaussian case we have the well-known formula

$$R(\tau) = \frac{2}{\pi} \sin^{-1} \left( \frac{\psi(t)}{\psi_0} \right)$$

and so

$$F(\tau) = \frac{1}{2} \left( \frac{\psi_0}{-\psi_0''} \right)^{\frac{1}{2}} \frac{d^2}{d\tau^2} \cos^{-1} \left( \frac{-\psi(t)}{\psi_0} \right). \quad (5.4.2)$$

(a) *Small values of  $\tau$ .* In the regular case the error is of order

$$-P_1(\tau) \sim -\frac{1}{6} \frac{C_3}{C_1} \tau \quad (5.4.3)$$

and in the singular case we find by expansion in powers of  $\tau$

$$F(\tau) \sim -\psi_0''' / 6\psi_0'' = \alpha.$$

By comparison with (4.7.2) the error is

$$-0.155\alpha. \quad (5.4.4)$$

(b) *Large values of  $\tau$ .* In (5.4.1) each of the terms  $X(+, -), X(+, +)$  tends to 0, and so  $F(\tau)$  tends to 0. The error thus vanishes.

5.5.  $p_r(\tau)$  and  $p_r^*(\tau)$ 

The sequence of approximations proposed in (I) depends on writing the first of equations (1.4.1) in the form

$$P_0(\tau) = -\frac{1}{W(+)} \frac{\partial^2}{\partial t_1 \partial t_n} U(t_n - t_1),$$

or

$$P_0(\tau) = \frac{1}{W(+)} \frac{d^2}{d\tau^2} U(\tau),$$

where  $U(\tau) = \frac{1}{2} p(0, \tau)$ , is the probability that  $f(t)$  be positive throughout the interval  $(0, \tau)$ . Let  $U(\tau)$  be replaced by the probability  $U_r(t^{(1)}, \dots, t^{(r)})$  that  $f(t)$  be positive at  $r$  suitably spaced points in  $(0, \tau)$ . (For convenience it is supposed that the points are equally spaced and that  $t^{(1)}, t^{(r)}$  are at the end-points.) As the number  $r$  of points is increased,  $U_r$  becomes an increasingly good approximation to  $U$ . The corresponding approximations to  $P_0(\tau)$  are defined by

$$p_r(\tau) = -\frac{1}{W(+)} \frac{\partial^2}{\partial t^{(1)} \partial t^{(r)}} U_r(t^{(1)}, \dots, t^{(r)}) \quad (5.5.1)$$

and

$$p_r^*(\tau) = \frac{1}{W(+)} \frac{d^2}{d\tau^2} U_r(\tau). \quad (5.5.2)$$

It turns out that in the Gaussian case  $p_3(\tau)$  is identical with  $F(\tau)$  given by equation (5.4.2). Generally, although  $U_r$  involves an  $r$ -fold normal integral, the approximations  $p_r$  and  $p_r^*$ , which depend on the derivatives of  $U_r$ , involve only  $(r-2)$ -fold normal integrals. Thus  $p_3, p_4, p_5$  and  $p_3^*, p_4^*, p_5^*$  can all be evaluated in terms of elementary functions.

(a) *Small values of  $\tau$ .* The difference  $(U_r - U)$  is equal to the probability that  $f(t)$  be positive at each of the points  $t^{(i)}$ , and have a zero-crossing at some point in the interval  $(0, \tau)$ . Hence

$f(t)$  must have two, four or more zeros in at least one of the subintervals  $(t^{(i)}, t^{(i+1)})$  (the first such zero a down-crossing) and certainly not one, three or five zeros in any of the remaining subintervals. If we neglect  $p(4, \tau)$  relative to  $p(2, \tau)$  and  $p(3, \tau)$  the probability of such an event for the subinterval  $(t^{(i)}, t^{(i+1)})$  is

$$\begin{aligned} & \iint_{t^{(i)} < t_1 < t_2 < t^{(i+1)}} W(-, +) dt_1 dt_2 - \iiint_{t^{(1)} < t_1 < t_2; t^{(i)} < t_2 < t_3 < t^{(i+1)}} W(+, -, +) dt_1 dt_2 dt_3 \\ & \quad - \iiint_{t^{(i)} < t_1 < t_2 < t^{(i+1)}; t_2 < t_3 < t^{(r)}} W(-, +, -) dt_1 dt_2 dt_3. \end{aligned} \quad (5.5.3)$$

Again, if  $p(4, \tau)$  is neglected the probability of a pair of zeros in more than one of the subintervals is negligible, so that the events are mutually independent. So  $(U_r - U)$  is equal to the sum of  $(r-1)$  expressions like (5.5.3).

On differentiating (5.5.3) partially with respect to both  $t^{(1)}$  and  $t^{(r)}$  the first integral vanishes identically whenever  $r > 2$ ; the other integrals also vanish except when  $i = 1$  or  $(r-1)$ . Hence we have

$$p_r(\tau) - P_0(\tau) \sim \frac{2}{W(+)} \frac{\partial^2}{\partial t^{(1)} \partial t^{(r)}} \iiint_{t^{(1)} < t_1 < t_2; t^{(r-1)} < t_2 < t_3 < t^{(r)}} W(+, -, +) dt_1 dt_2 dt_3.$$

Substituting for  $W(+, -, +)$  from (3.1.4) we find

$$p_r(\tau) - P_0(\tau) = -\frac{1}{3} (C_3/C_1) (t^{(r)} - t^{(1)}) (t^{(r)} - t^{(r-1)}) (t^{(r)} + 2t^{(r-1)} - 3t^{(1)}).$$

Now putting  $(t^{(r)} - t^{(1)}) = \tau$  and  $(t^{(r)} - t^{(r-1)}) = \tau/(r-1)$  we obtain for the error in  $p_r$

$$-\frac{(3r-5)}{3(r-1)^3} \frac{C_3}{C_1} \tau^4. \quad (5.5.4)$$

Thus  $p_r(\tau)$  is correct to order  $\tau^3$  (not  $\tau^4$ , as was stated in (I)). In particular when  $r = 3, 4, 5$  the errors are, respectively,

$$-\frac{1}{6} \frac{C_3}{C_1} \tau^4, \quad -\frac{7}{81} \frac{C_3}{C_1} \tau^4, \quad -\frac{5}{96} \frac{C_3}{C_1} \tau^4. \quad (5.5.5)$$

The case  $r = 3$  is in agreement with (5.4.3).

On the other hand  $p_r^*$  involves the first term in (5.5.3) which is of a lower order. Thus

$$U_r - U = (r-1) \iint_{0 < t_2 < t_3 < \tau/(r-1)} W(+, -) dt_2 dt_3 + O(\tau^4).$$

On substituting in (5.5.2) and using (3.2.4) we find

$$p_r^*(\tau) - P_0(\tau) = \frac{1}{r-1} P_0\left(\frac{\tau}{r-1}\right) + O(\tau^4).$$

Therefore at  $\tau = 0$

$$\left. \begin{aligned} p_r^* &= \left(1 + \frac{1}{r-1}\right) P_0 = 0, \\ \frac{dp_r^*}{d\tau} &= \left(1 + \frac{1}{(r-1)^2}\right) \frac{dP_0}{d\tau}, \\ \frac{d^2 p_r^*}{d\tau^2} &= \left(1 + \frac{1}{(r-1)^3}\right) \frac{d^2 P_0}{d\tau^2}, \\ \frac{d^3 p_r^*}{d\tau^3} &= \left(1 + \frac{1}{(r-1)^4}\right) \frac{d^3 P_0}{d\tau^3}. \end{aligned} \right\}$$

## INTERVALS BETWEEN ZEROS OF A RANDOM FUNCTION 585

For example, when  $r = 3$  we have

$$\frac{d\hat{p}_3^*}{d\tau} = \frac{5}{4} \frac{dP_0}{d\tau},$$

a relation proved independently in (I). For  $r \geq 3$ , since  $P_0(\tau) \sim (C_2/C_1)\tau$ , the error in  $\hat{p}_r^*(\tau)$  is

$$\frac{1}{(r-1)^2} \frac{C_2}{C_1} \tau. \quad (5.5.6)$$

In the singular case we make a straightforward expansion of  $\hat{p}_r(\tau)$  in powers of  $\tau$ . The calculations lead eventually to the following, when  $r = 3, 4, 5$

$$\begin{aligned} \hat{p}_3(0) &= \alpha, \\ \hat{p}_4(0) &= \frac{1}{\pi} \left[ \cos^{-1} \left( \frac{-7}{8} \right) + \frac{8}{3\sqrt{15}} \right] \alpha = 1.0583\alpha, \\ \hat{p}_5(0) &= \frac{1}{2\pi} \left[ \left\{ 2 \cos^{-1} \left( \frac{-11}{12} \right) - \cos^{-1} \left( \frac{7}{9} \right) \right\} + \frac{1}{16} \left\{ \frac{83}{\sqrt{23}} + 11\sqrt{2} \right\} \right] \alpha \\ &= 1.0879\alpha. \end{aligned}$$

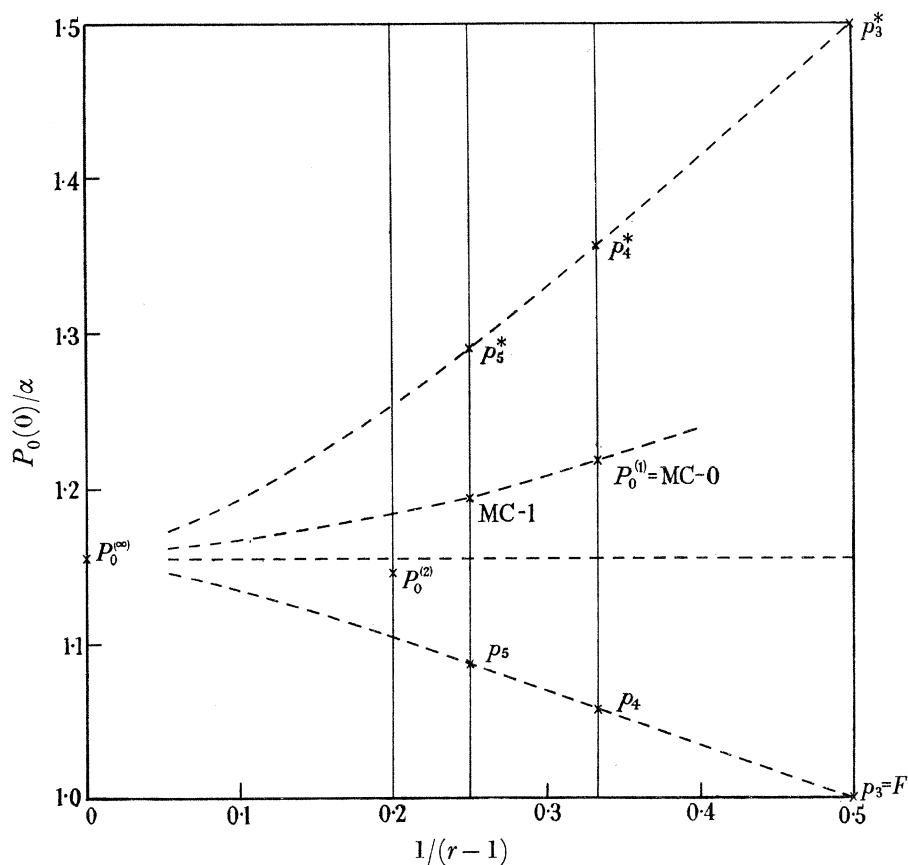
The corresponding expansions of  $\hat{p}_r^*(\tau)$  lead to

$$\begin{aligned} \hat{p}_3^*(0) &= 1.5\alpha, \\ \hat{p}_4^*(0) &= \frac{\alpha}{9\pi} \left[ \cos^{-1} \left( \frac{-1}{4} \right) + 2 \cos^{-1} \left( \frac{-3\sqrt{3}}{4\sqrt{2}} \right) + 8 \cos^{-1} \left( \frac{\sqrt{3}}{2\sqrt{2}} \right) + 9 \cos^{-1} \left( \frac{-7}{8} \right) \right] \\ &= 1.3551\alpha, \\ \hat{p}_5^*(0) &= \frac{\alpha}{16\pi} \left[ \left( \cos^{-1} \frac{-7}{6\sqrt{2}} + \cos^{-1} \frac{-5}{3\sqrt{3}} + \cos^{-1} \frac{-1}{4} + \cos^{-1} \frac{1}{2\sqrt{6}} \right) + 16 \cos^{-1} \left( \frac{-11}{12} \right) \right. \\ &\quad \left. - 2 \left( \cos^{-1} \frac{3\sqrt{3}}{4\sqrt{2}} + \cos^{-1} \frac{1}{3} \right) + 4 \left( \cos^{-1} \frac{3}{4\sqrt{2}} - \cos^{-1} \frac{13}{8\sqrt{3}} \right) \right. \\ &\quad \left. + 8 \left( \cos^{-1} \frac{3}{2\sqrt{2}} - \cos^{-1} \frac{7}{9} \right) + 9 \left( \cos^{-1} \frac{5}{4\sqrt{3}} + \cos^{-1} \frac{1}{3} + \cos^{-1} \frac{7}{8} \right) \right] \\ &= 1.2899\alpha. \end{aligned}$$

These results are plotted in figure 2 against the abscissa  $1/(r-1)$ , and taking  $\alpha = 1$ . The point labelled  $P_0^{(\infty)}$  and plotted at  $r = \infty$  corresponds to the estimate  $P_0 \doteq 1.155\alpha$  of § 4.7. It will be seen that  $\hat{p}_r$  and  $\hat{p}_r^*$  both approach  $P_0^{(\infty)}$ , the one from below and the other from above.

For comparison, the other approximations to  $P_0(0)$  discussed above have been plotted in the same diagram. Because they are of comparable complexity, MC-0 and MC-1 have been plotted on the same ordinates as  $\hat{p}_4$  and  $\hat{p}_5$ , respectively.  $P_0^{(2)}$ , which involves the single integration of a known function (corresponding to a fourfold integration) is plotted level with  $r = 6$ . It is obviously the closest approximation.

(b) *Large values of  $\tau$ .* As  $\tau \rightarrow \infty$  so  $U_r(\tau)$  tends to zero; for the probability that  $f(t)$  remain of constant sign throughout the infinite interval becomes vanishingly small. Hence also  $\hat{p}_r$  and  $\hat{p}_r^*$  tend to zero at infinity, and the error in both  $\hat{p}_r$  and  $\hat{p}_r^*$  is vanishingly small.

FIGURE 2. Approximations to  $P_0(0)$  in the singular case.

### 5.6. McFadden's second approximation

A quite different method was suggested in a later paper by McFadden (1958). On the assumption that a given interval  $\tau$  is independent of the sums of the previous  $(2m+2)$  intervals ( $m = 0, 1, 2, \dots$ ) McFadden derived the integral equation

$$P_0(\tau) \doteq X(+, -) - X(+, +) * P_0(\tau). \quad (5.6.1)$$

$X(+, -)$  and  $X(+, +)$  are as in § 1.1 and a star  $*$  denotes convolution:

$$F_1(\tau) * F_2(\tau) = \int_0^\tau F_1(\tau') F_2(\tau - \tau') d\tau'.$$

The solution of this (approximate) integral equation may be denoted by  $\text{McF}(\tau)$ .

(a) *Small values of  $\tau$ .* In the regular case,  $X(+, +) \sim P_1$  (by (1.2.1)) and so the second term in (5.6.1) is, by (3.2.6),

$$-X(+, +) * P_1(\tau) \sim -\frac{C_2 C_4}{6C_1^2} \int_0^\tau \tau'^4 (\tau - \tau') d\tau' = \frac{-1}{180} \frac{C_2 C_4}{C_1^2} \tau^6.$$

But we saw that  $X(+, -)$ , or  $P_0^{(1)}$  is correct to order  $\tau^7$ , so that the major part of the error is in fact due to the second term. This corresponds to the fact that for small intervals at least the assumption of independence is invalid (Palmer 1956).

In the singular case the second term is zero at the origin and so

$$\text{McF}(\tau) = X(+, -) = 1.218\alpha.$$

The error is therefore the same as in  $P_0^{(1)}$  (see equation (5.1.4)).

## INTERVALS BETWEEN ZEROS OF A RANDOM FUNCTION 587

(b) *Large values of  $\tau$ .* Since  $X(+, -)$  and  $X(+, +)$  both become equal to  $W(+)$  at infinity. The solution to the limiting equation

$$F(\tau) \doteq W(+)-W(+)\int_0^{\tau} F(\tau-\tau')d\tau'$$

is

$$F(\tau) \doteq W(+)\text{e}^{-W(+)\tau},$$

a negative exponential. Hence we expect  $\text{McF}(\tau)$  to tend to zero exponentially at infinity.

The results of §§ 5.1 and 5.6 are summarized in table 1.

TABLE 1. COMPARISON OF DIFFERENT APPROXIMATIONS TO  $P_0(\tau)$

approximation	error for small $\tau$		error for large $\tau$
	regular case	singular case	
$P_0^{(1)} = \text{MC-0}$	$\frac{1}{280} \frac{C_4}{C_1} \tau^8$	$0.062\alpha$	$W(+)$
$P_0^{(2)}$	$-\frac{8}{21 \cdot (11)!} \frac{C_6}{C_1} \tau^{19}$	$-0.008\alpha$	$-W(+)^2 \tau$
MC-1	$\frac{39}{8!} \frac{C_4}{C_1} \tau^8$	$0.040\alpha$	$\frac{1}{2}W(+)$
$p_3 = F$	$-\frac{1}{6} \frac{C_3}{C_1} \tau^4$	$-0.155\alpha$	$o(1)$
$p_4$	$-\frac{7}{81} \frac{C_3}{C_1} \tau^4$	$-0.097\alpha$	$o(1)$
$p_5$	$-\frac{5}{96} \frac{C_3}{C_1} \tau^4$	$-0.067\alpha$	$o(1)$
$p_3^*$	$\frac{1}{4} \frac{C_2}{C_1} \tau$	$0.345\alpha$	$o(1)$
$p_4^*$	$\frac{1}{9} \frac{C_2}{C_1} \tau$	$0.200\alpha$	$o(1)$
$p_5^*$	$\frac{1}{16} \frac{C_2}{C_1} \tau$	$0.135\alpha$	$o(1)$
McF	$-\frac{1}{180} \frac{C_2 C_4}{C_1^2} \tau^6$	$0.062\alpha$	$o(1)$

### 5.7. The narrow-band approximation

Let  $E(\sigma)$  denote the spectral density of  $f(t)$ , related to  $\psi(t)$  by the equations

$$\left. \begin{aligned} E(\sigma) &= \frac{2}{\pi} \int_0^{\infty} \psi(t) \cos \sigma t dt, \\ \psi(t) &= \int_0^{\infty} E(\sigma) \cos \sigma t d\sigma. \end{aligned} \right\}$$

The mean frequency of the spectrum is defined as  $\bar{\sigma} = m_1/m_0$ , where

$$m_r = \int_0^{\infty} E(\sigma) \sigma^r d\sigma$$

is the  $r$ th moment of the spectrum. It is convenient to write also

$$\mu_r = \int_0^{\infty} E(\sigma) (\sigma - \bar{\sigma})^r d\sigma.$$



The spectrum is said to be *narrow* if

$$\frac{\mu_2}{\bar{\sigma}^2 \mu_0} = \delta^2 \ll 1$$

and it can be shown (I) that in that case  $\psi(t)$  has the form

$$\psi(t) = A(t) \cos \bar{\sigma}t + O(\delta \bar{\sigma}t)^3, \quad (5.7.1)$$

where  $A(t)$  is a slowly varying function of  $t$

$$A(t) = \mu_0 - \frac{1}{2} \mu_2 t^2 = \psi_0 (1 - \frac{1}{2} \delta^2 \bar{\sigma}^2 t^2).$$

Under these conditions one expects  $f(t)$  to have the form of a sine wave of almost constant frequency  $\bar{\sigma}$  and slowly varying amplitude, so that the greater part of the distribution  $P_0(\tau)$  lies within the neighbourhood of  $\tau_0 = \pi/\bar{\sigma}$ . In fact the approximation  $F(\tau)$  of equation (5.4.2) reduces to

$$F(\tau) \doteq \frac{1}{2\delta\tau_0 [1 + (\tau - \tau_0)^2 / (\delta\tau_0)^2]^{\frac{3}{2}}} \quad (5.7.2)$$

provided  $(\tau - \tau_0)$  is comparable with  $\delta\tau_0$ . It can be shown that all the other approximations discussed in this paper have the same limiting form as  $\delta \rightarrow 0$ . This approximation will be called the narrow-band approximation and will be denoted by NB( $\tau$ ). It has the following properties:

- (1) It is symmetrical about the mean point  $\tau = \tau_0$ .
- (2) For large values of  $(\tau - \tau_0)$  it is of order  $|\tau - \tau_0|^{-3}$ , and so has no second moment.
- (3) The maximum probability density is

$$\text{NB}(\tau_0) = \frac{1}{2\delta\tau_0}.$$

- (4) The width of the curve where it reaches half its maximum height is given by

$$2 \times (2^{\frac{2}{3}} - 1)^{\frac{1}{2}} \delta\tau_0 = 1.533\delta\tau_0.$$

- (5) The cumulative distribution function is

$$\int_{-\infty}^{\tau} \text{NB}(\tau) d\tau = \frac{(\tau - \tau_0) / (\delta\tau_0)}{2[1 + (\tau - \tau_0)^2 / (\delta\tau_0)^2]^{\frac{3}{2}}} + \frac{1}{2}.$$

- (6) Hence the quartiles are given by

$$\frac{\tau - \tau_0}{\delta\tau_0} = \frac{1}{\sqrt{3}} = 0.5774$$

and the interquartile range is  $1.155\delta\tau_0$ .

### 5.8. Numerical computations

In this last section, the various approximations to  $P_0(\tau)$  described in §§ 5.1 to 5.7 are compared by numerical computation over values of  $\tau$  not necessarily very small or very large. Where possible, the results are compared with those obtained experimentally by analogue methods (Favreau *et al.* 1956; Blötekjaer 1958).

Only these approximations are shown which are the highest of their type at present available. For example, MC-0 is not shown if MC-1 is available, and  $p_3, p_4$  are not shown if  $p_5$  is available.

## INTERVALS BETWEEN ZEROS OF A RANDOM FUNCTION 589

Figures 3 to 7 show the approximations  $p_5, p_5^*, P_0^{(2)}$ , McF and MC-0 (or MC-1) for spectra of the form

$$E(\sigma) = \sigma^{2n}/(1 + \sigma^2)^m$$

(see table 2). In two cases (figures 3 and 6) the spectra are  $O(\sigma^{-4})$  at infinity so that  $\psi$  has the singularity discussed in § 4. Figure 3 corresponds to the case discussed in § 4.6, where it was shown that  $P_0(\tau)$  is not a negative exponential, as the observations might suggest.

The experimental results of Favreau *et al.* are indicated in figure 3 by the plotted points (the vertical lines indicate the estimated uncertainty of the observations); in figures 4 to 7 the experimental values (Favreau *et al.*) are shown by broken curves.

The curves which form lower and upper bounds of  $P_0$ —namely  $P_0^{(2)}$  and MC-0 or MC-1—are drawn rather more heavily than the others. From figures 3, 6 and 7 it will be seen that at small values of  $\tau$  the experimental points lie considerably below the theoretical lower bound. This implies that in some other parts of the curve the experimental points must be too high, since the total area under each curve must be unity.

TABLE 2

figure	$E(\sigma)$	$\psi(t)$
3	$(1 + \sigma^2)^{-2}$	$e^{- t }(1 +  t )$
4	$(1 + \sigma^2)^{-3}$	$e^{- t }(1 +  t  + \frac{1}{3}t^2)$
5	$(1 + \sigma^2)^{-5}$	$e^{- t }(1 +  t  + \frac{5}{2}t^2 + \frac{2}{21} t ^3 + \frac{1}{105}t^4)$
6	$\sigma^4(1 + \sigma^2)^{-4}$	$e^{- t }(1 +  t  - 2t^2 + \frac{1}{3} t ^3)$
7	$\sigma^4(1 + \sigma^2)^{-5}$	$e^{- t }(1 +  t  - \frac{1}{3}t^2 - \frac{2}{3} t ^3 + \frac{1}{15}t^4)$

Nevertheless, there is substantial agreement between the experimental points and the three approximations  $p_5^*, P_0^{(2)}$  and McF. For all except small values of  $\tau$  the agreement with  $p_5$  is less good than with  $p_5^*$ .

Figure 8 shows a similar study in the case of the Butterworth spectrum

$$E(\sigma) = 1/(1 + \sigma^{14}),$$

which has a fairly sharp cut-off at about  $\sigma = 1$ . The plotted points are those of Favreau *et al.* (1956). Evidently the agreement between the observations and the theoretical curves  $p_5^*, P_0^{(2)}$  and McF is quite close. In the range  $6 < \tau < 8$ , where  $P_0^{(2)}$  is the uppermost of the three curves  $P_0^{(2)}$  must also be the closest approximation, since it is a lower bound.

For the low-pass spectrum

$$E(\sigma) = \begin{cases} 1 & (0 < \sigma < 1), \\ 0 & (1 < \sigma < \infty), \end{cases}$$

the best experimental results available appear to be those of Blötekjaer (1958), which are represented by the broken curve (B) in figure 9. It will be seen that the agreement with McF is fairly good, with  $p_5^*$  somewhat better, and with  $P_0^{(2)}$  very close indeed, as far as  $\tau = 14$ .

For the band-pass spectrum

$$E(\sigma) = \begin{cases} 0 & (0 < \sigma < \sigma_1), \\ 1 & (\sigma_1 < \sigma < 1), \\ 0 & (1 < \sigma < \infty), \end{cases}$$

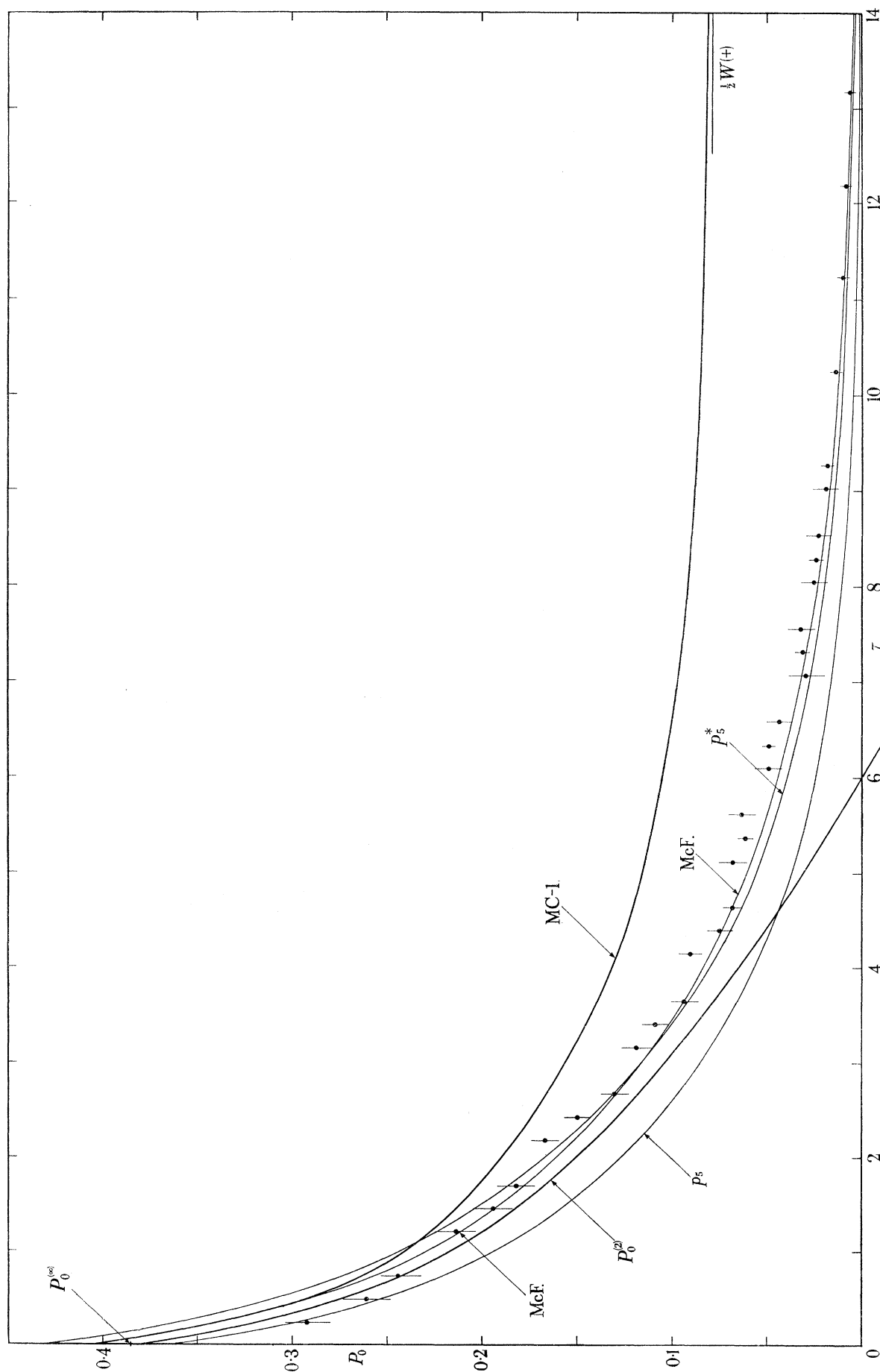


FIGURE 3. Approximations to  $P_0(\tau)$  for the spectrum  $E = (1 + \sigma^2)^{-2}$ . The plotted points are the experimental results of Favreau, Low and Pfeffer.

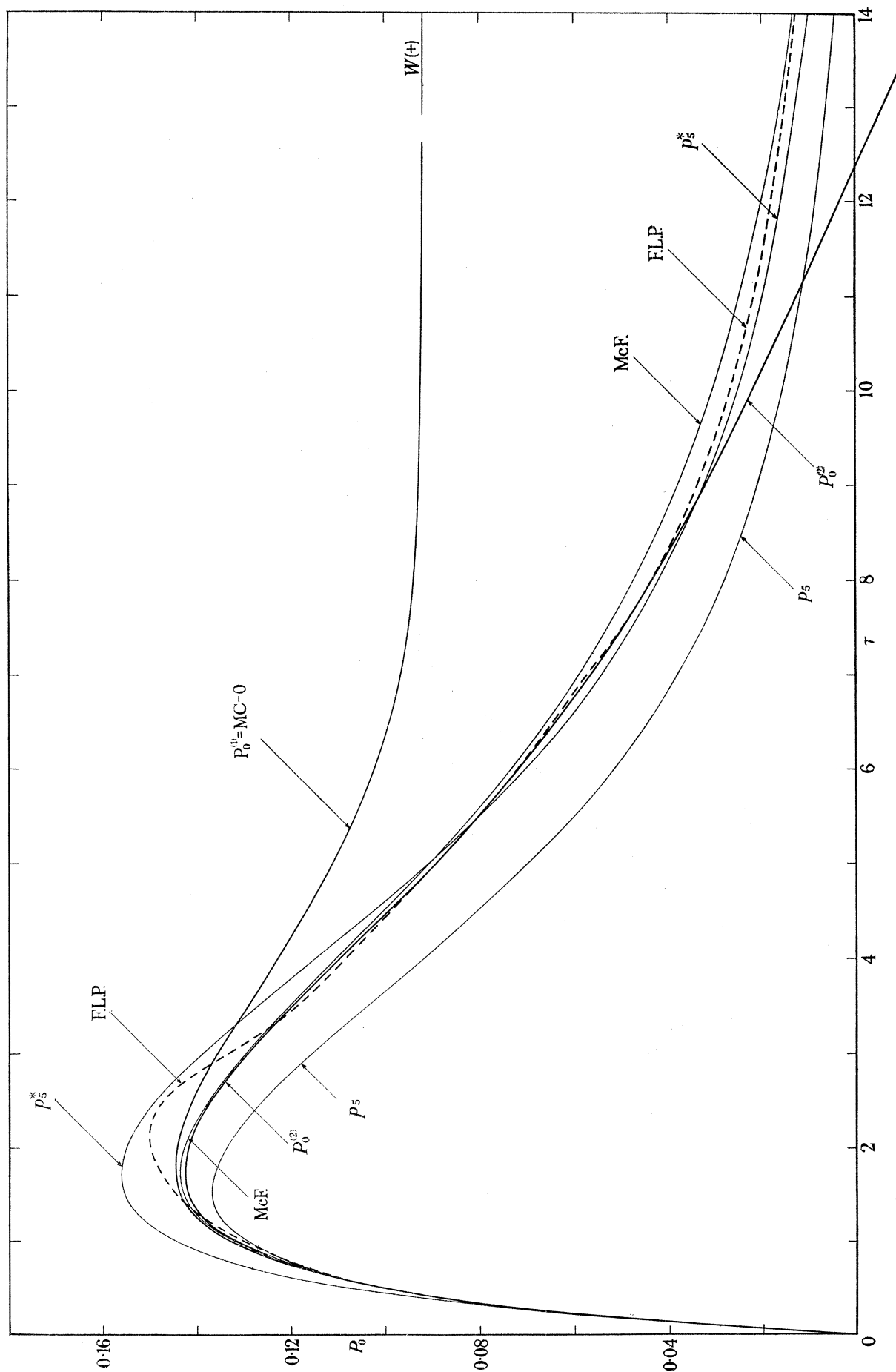


Figure 4. Approximations to  $P_0(\tau)$  for the spectrum  $E = (1 + \sigma^2)^{-3}$ .

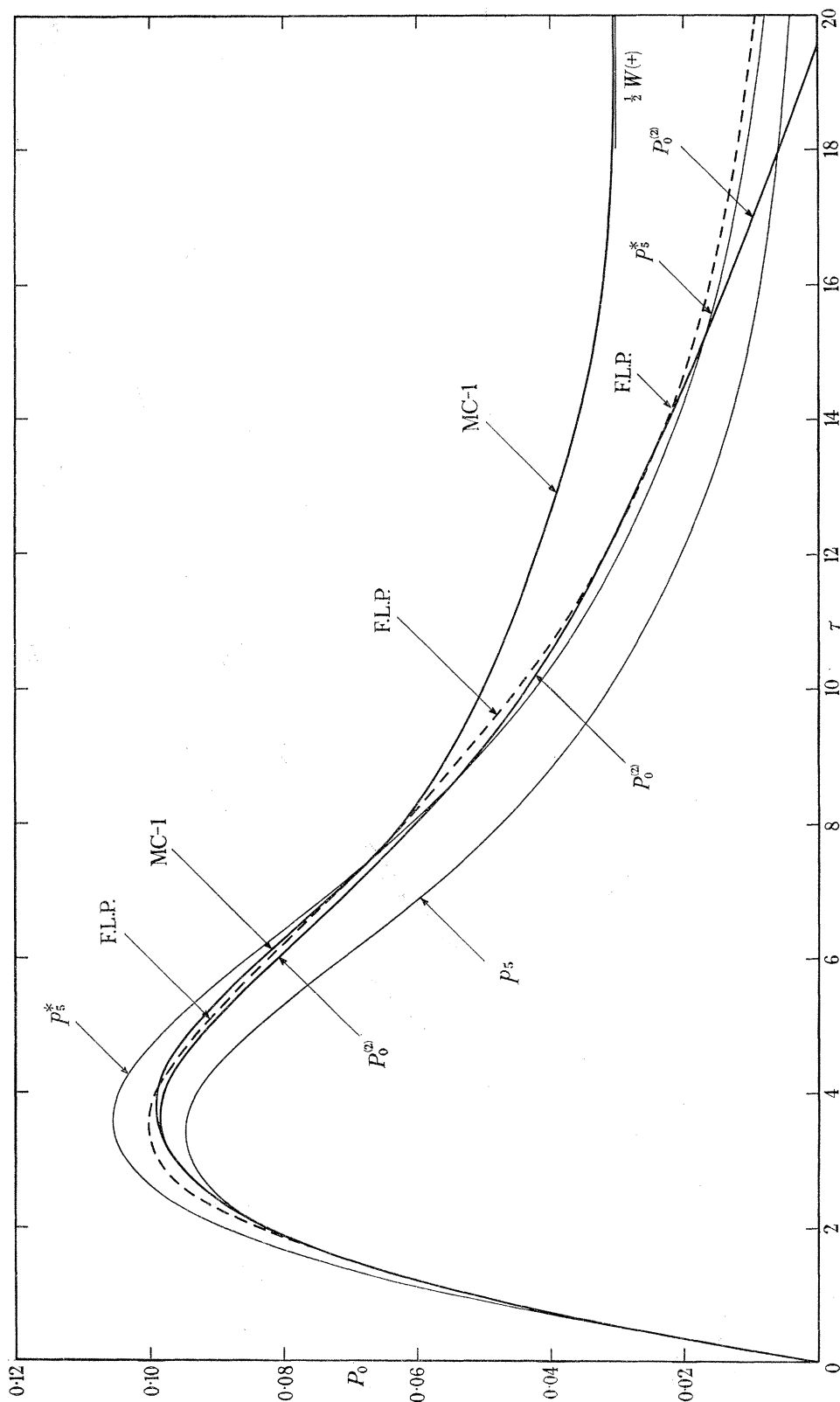


FIGURE 5. Approximations to  $P_0(\tau)$  for the spectrum  $E = (1 + \sigma^2)^{-5}$ .

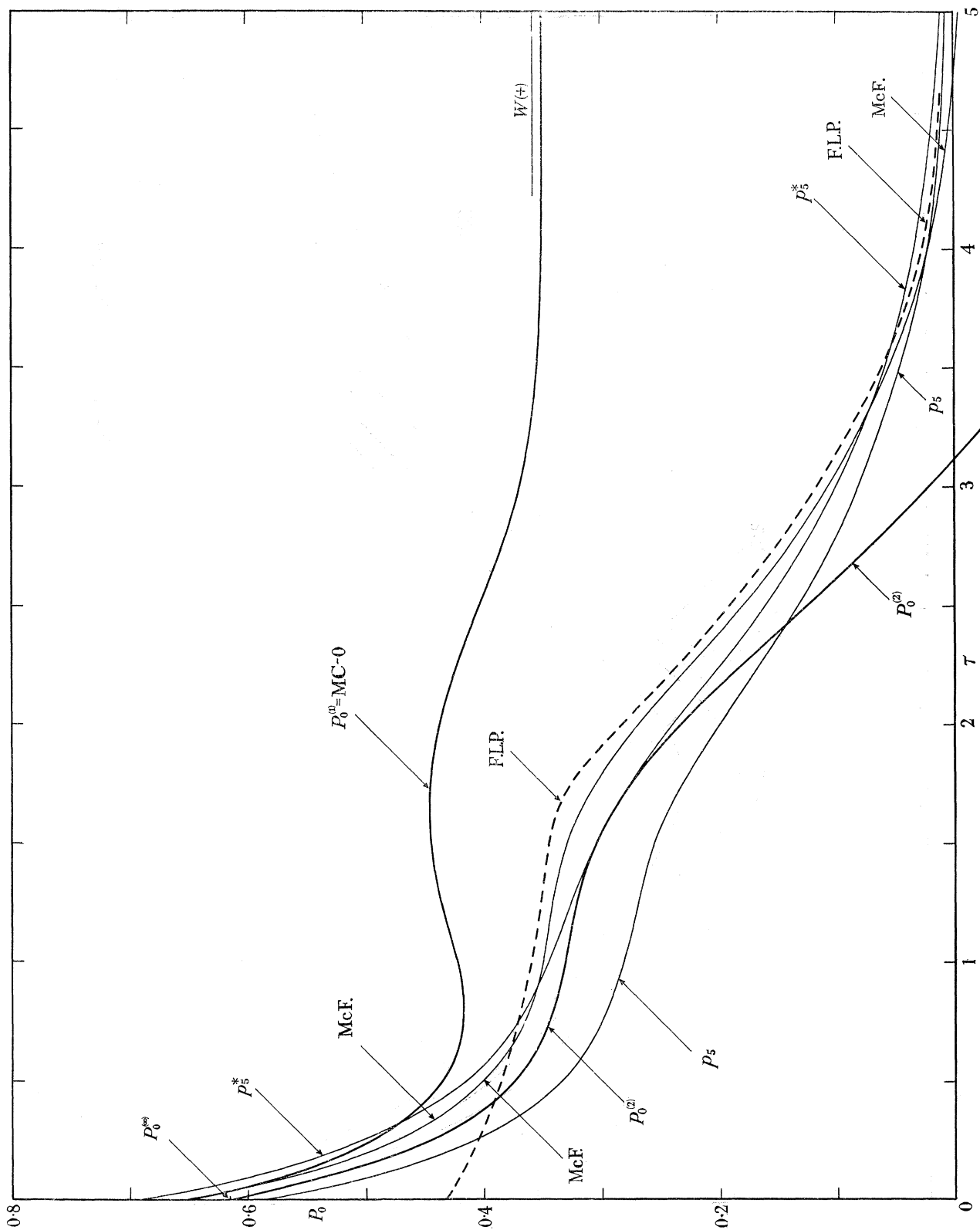


FIGURE 6. Approximations to  $P_0(\tau)$  for the spectrum  $E = \sigma^4(1 + \sigma^2)^{-4}$ .

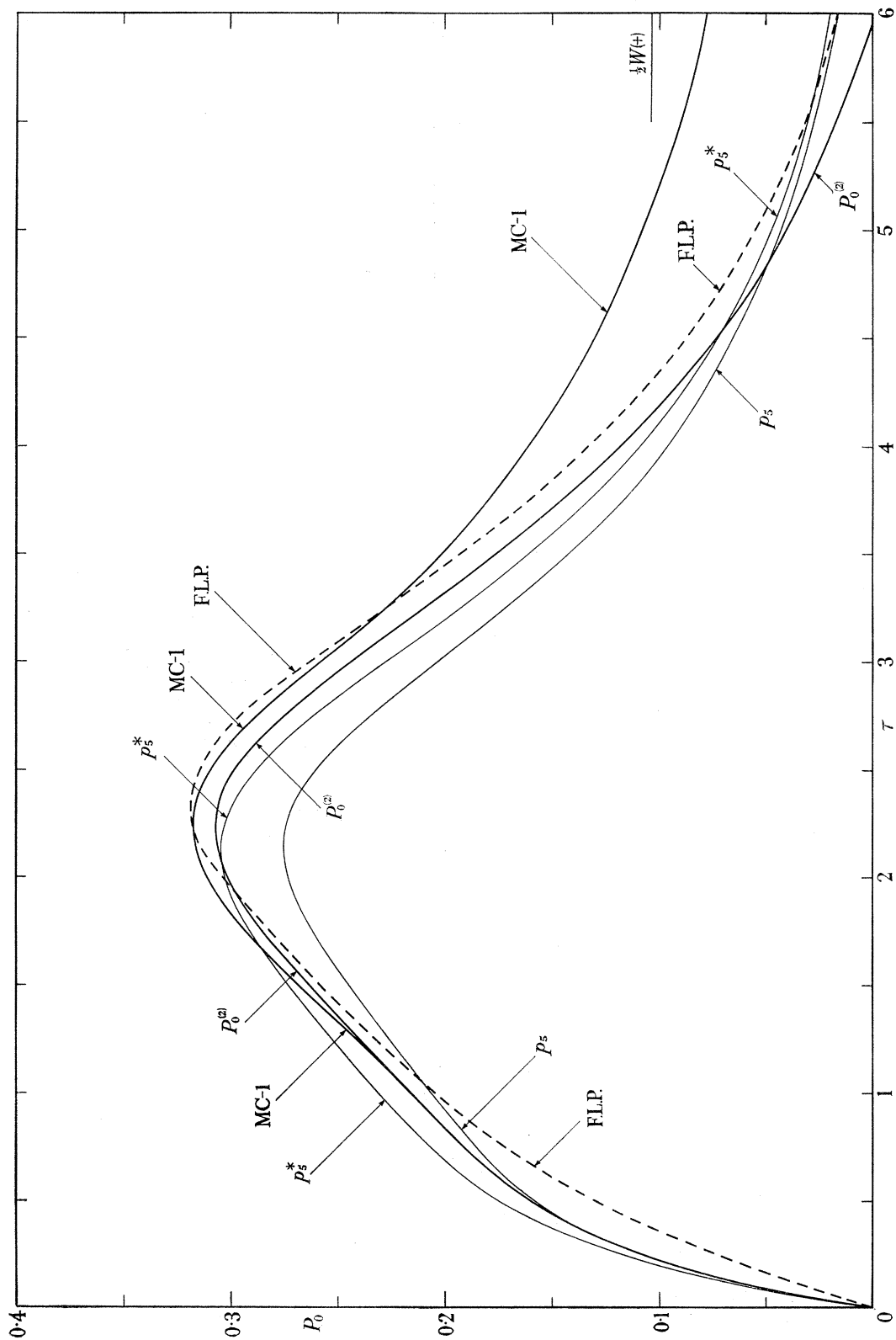


FIGURE 7. Approximations to  $P_0(\tau)$  for the spectrum  $E = \sigma^4(1 + \sigma^2)^{-5}$ .

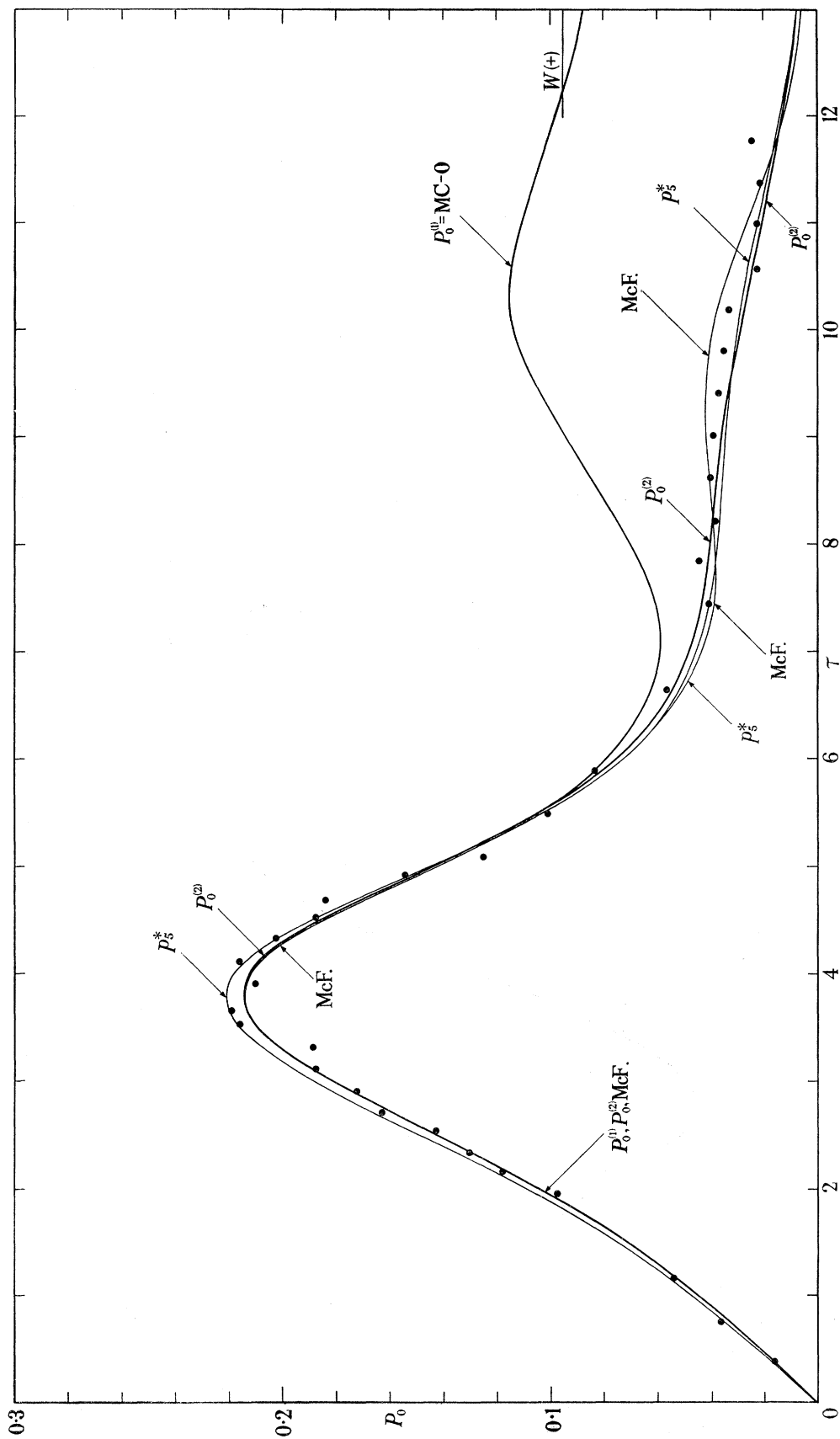


FIGURE 8. Approximations to  $P_0(\tau)$  for the spectrum  $E = (1 + \sigma^{14})^{-1}$ . The plotted points are the experimental results of Favreau, Low & Pfeffer.



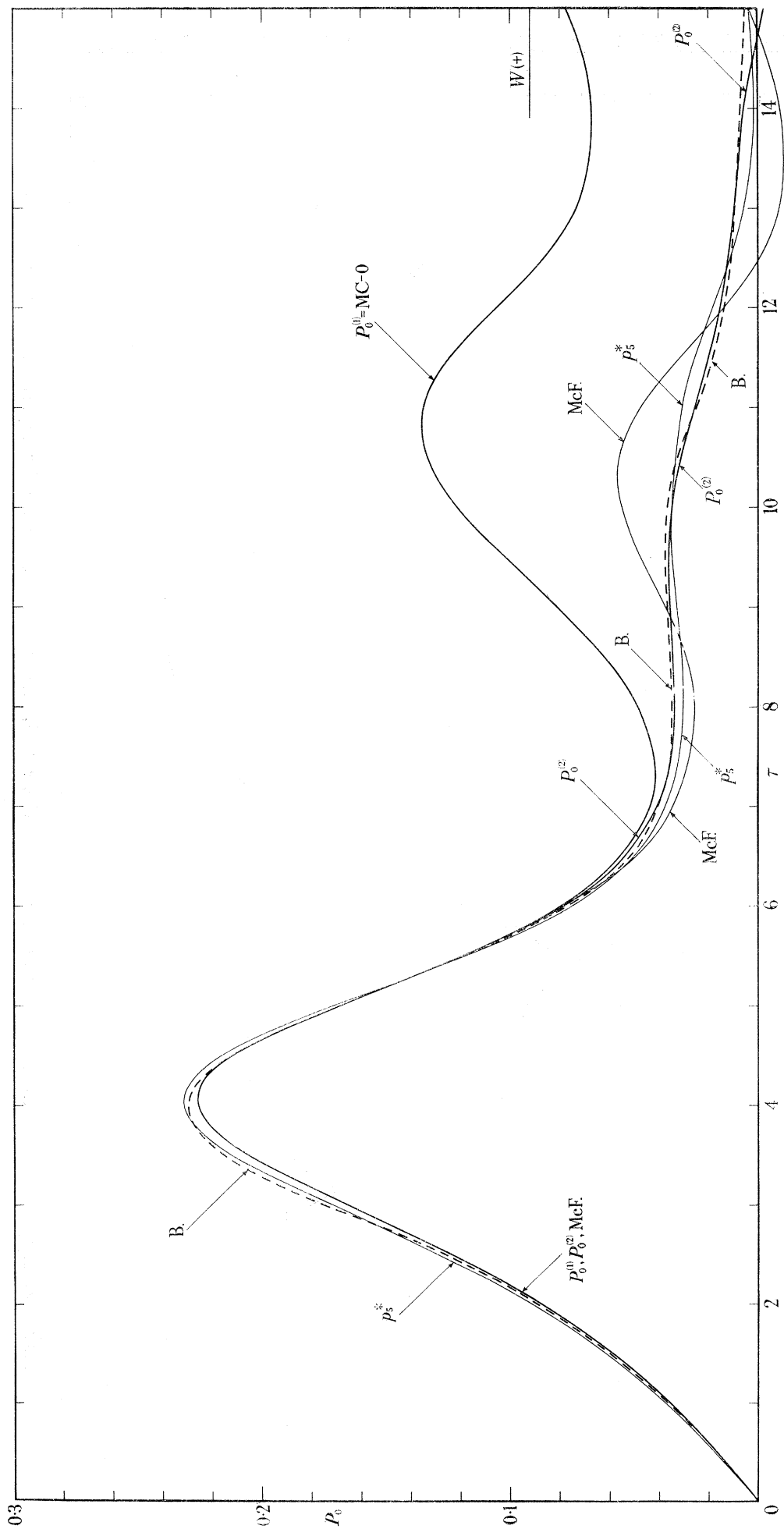


FIGURE 9. Approximations to  $P_0(\tau)$  for a low-pass spectrum (cut-off at  $\sigma = 1$ ).

## INTERVALS BETWEEN ZEROS OF A RANDOM FUNCTION 597

$P_0^{(2)}$  can be expected to be an even closer approximation than that for the low-pass spectrum, since some of the low frequencies will have been eliminated. Accordingly  $P_0^{(2)}$  has been plotted in figure 10 for  $\sigma_1 = 0, \frac{1}{2}, \frac{3}{4}$  and  $\frac{7}{8}$ . It will be seen that as  $\sigma_1$  approaches 1 and the spectrum becomes narrower so the distribution also becomes narrower and the height of

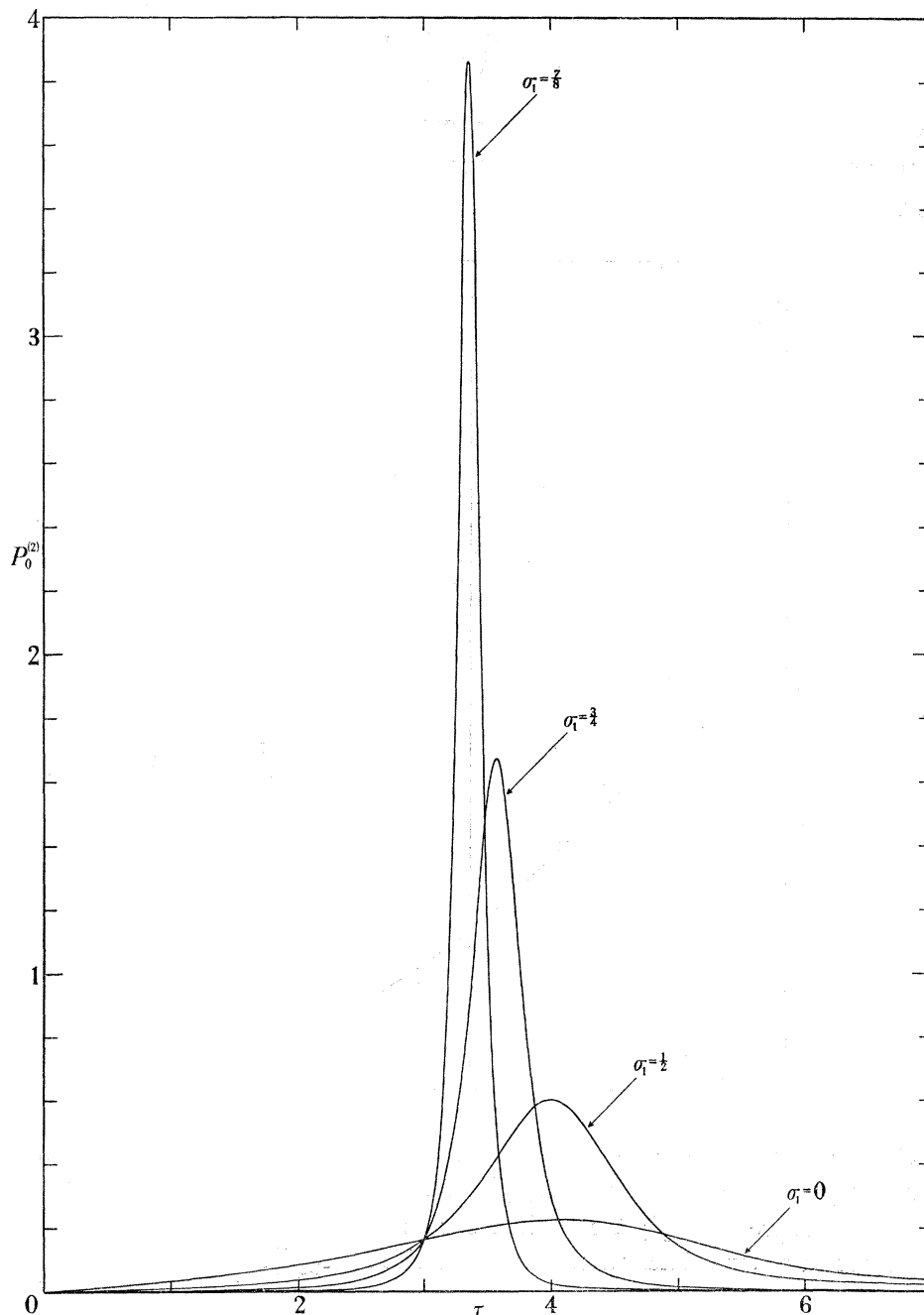


FIGURE 10. Computed values of  $P_0^{(2)}(\tau)$  for a band-pass spectrum (low and high cut-offs at  $\sigma_1$  and 1, respectively).

the maximum probability density is increased. The position and height of the maximum, and the width of the curve at half the maximum height are shown by the full curves (a), (b) and (c) in figure 11. (For plotting these curves the distributions  $P_0^{(2)}$  for  $\sigma_1 = 0.1, 0.2, 0.3, 0.4$  and  $0.6$  were also computed.)

A comparison may be made with the narrow-band approximation of § 5.7, which gives for the abscissa of the maximum

$$\tau_0 = 2\pi(1 + \sigma_1)^{-1}$$

and for the height and width of the distribution  $(2\delta\tau_0)^{-1}$  and  $1.533\delta\tau_0$  respectively, where

$$\delta\tau_0 = \frac{2\pi}{\sqrt{3}} \frac{1 - \sigma_1}{(1 + \sigma_1)^2}.$$

These values are represented by the broken curves in figure 11. It will be seen that for  $\sigma_1 > \frac{1}{2}$  the narrow band approximation agrees well with  $P_0^{(2)}$  but there are noticeable divergences when  $\sigma_1 < \frac{1}{2}$ .

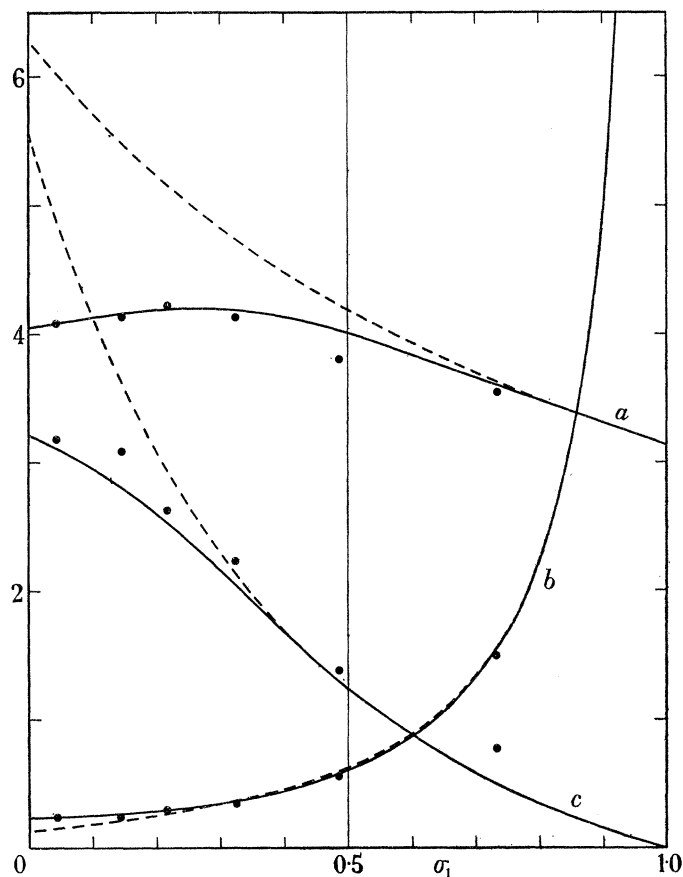


FIGURE 11. Characteristics of the distribution of intervals in a band-pass spectrum. (*a*) Abscissa of maximum; (*b*) ordinate of maximum; (*c*) width of distribution at half the maximum height. The full curves represent  $P_0^{(2)}$ ; the broken curves the narrow-band approximation. Plotted points correspond to experimental curves of Blötekjaer (1958).

In the same figure some experimental results due to Blötekjaer (1958) have been inserted. As one would expect, they agree with  $P_0^{(2)}$  rather than with NB. In particular one may note the slight negative trend in curve (*a*) as  $\sigma_1$  approaches zero.

The author is indebted to S. O. Rice, D. S. Palmer and J. A. McFadden for stimulating correspondence, and to the latter especially for useful comments. A reference to the work of M. L. Mehta was provided by F. J. Dyson. The computations on  $p_r$  and  $p_r^*$  were carried

## INTERVALS BETWEEN ZEROS OF A RANDOM FUNCTION 599

out by Miss D. B. Catton on the DEUCE at the Royal Aircraft Establishment, Farnborough; the computations on  $P_0^{(2)}$  were carried out at the Mathematical Laboratory, Cambridge, by Mrs M. O. Mutch and Mr P. F. Swinnerton-Dyer.

## REFERENCES

- Bartlett, M. S. 1955 *Stochastic processes*. Cambridge University Press.
- Blötekjaer, K. 1958 An experimental investigation of some properties of band-pass limited Gaussian noise. *Trans. Inst. Rad. Engrs*, IT-4, 100–102.
- Briggs, B. H. & Page, E. S. 1955 *Report of the 1954 Conference on the Physics of the Ionosphere*, p. 119. London: Physical Society.
- Ehrenfeld, S., Goodman, N. R., Kaplan, S., Mehr, E., Pierson, W. J., Stevens, R. & Tick, L. J. 1958 Theoretical and observed results for the zero and ordinate crossing problems of stationary Gaussian noise with application to pressure records of ocean waves. *N.Y. Univ., Coll. Eng., Tech. Rep.* no. 1, Dec. 1958.
- Favreau, R. R., Low, H. & Pfeffer, I. 1956 Evaluation of complex statistical functions by an analog computer. *Inst. Rad. Eng. Nat. Convention Record* 1956, pt. 4, pp. 31–37.
- Kac, M. 1943 On the distribution of values of trigonometric sums with linearly independent frequencies. *Amer. J. Math.* **65**, 609–615.
- Kamat, A. R. 1953 Absolute moments of the multivariate normal distribution. *Biometrika*, **40**, 20–34.
- Kohlenberg, A. 1953 Notes on the zero distribution of Gaussian noise. *Mass. Inst. Tech. Lincoln Lab., Tech. Mem.* no. 44.
- Longuet-Higgins, M. S. 1953 Contribution to Discussion, Symposium on Microseisms. *U.S. Nat. Acad. Sci. Publ.* 306, p. 124.
- Longuet-Higgins, M. S. 1957 A statistical distribution arising in the study of the ionosphere. *Proc. Phys. Soc. B*, **70**, 559–565.
- Longuet-Higgins, M. S. 1958 On the intervals between successive zeros of a random function. *Proc. Roy. Soc. A*, **246**, 99–118.
- McFadden, J. A. 1956 The axis-crossing intervals of random functions. *Trans. Inst. Rad. Engrs*, IT-2, 146–150.
- McFadden, J. A. 1958 The axis-crossing intervals of random functions. II. *Trans. Inst. Rad. Engrs*, IT-4, 14–24.
- Mehta, M. L. 1960 On the statistical properties of the level-spacings in nuclear spectra. *Nucl. Phys.* **18**, 395–419.
- Nabeya, S. 1952 Absolute moments in 3-dimensional normal distribution. *Ann. Inst. Stat. Math.* **4**, 15–29.
- Palmer, D. S. 1956 Properties of random functions. *Proc. Camb. Phil. Soc.* **52**, 672–686.
- Rice, S. O. 1944 The mathematical analysis of random noise. *Bell Syst. Tech. J.* **23**, 282–332.
- Rice, S. O. 1945 The mathematical analysis of random noise. *Bell Syst. Tech. J.* **24**, 46–156.
- Rice, S. O. 1958 Distribution of the duration of fades in radio transmission: Gaussian noise model. *Bell Syst. Tech. J.* **37**, 581–635.
- Siebert, A. J. F. 1951 On the first passage-time probability problems. *Phys. Rev.* **81**, 617–623.