Statistical Analysis of Low Frequency Responses of a Moored Floating Offshore Structure (1st Report)*

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ABSTRACTS

The purpose of this paper is to estimate the response statistics of a moored floating structure that can be modelled as the two term Volterra series expansion subjected to a stationary Gaussian random waves.

For estimating the instantaneous probability density function of response the approximate method using the finite Gram-Charlier expansion and the asymptotic form of the exact solution which can be obtained from Kac-Siegert method is proposed. In order to estimate the probability density function of extremal values consisting maxima and minima and the extreme responses the assumptions in which response and response velocity are mutually independent and the velocity is a Gaussian process with zero mean are introduced in addition to Powell's assumptions in the field of structural dynamics.

The frequency properties have been found experimentally through cross spectral and cross bispectral analyses.

Comparisons between the experimental results and the statistical ones estimated from the frequency properties of response are discussed. As the results it has been confirmed that both results show a fairly good agreement.

1. INTRODUCTION

For a moored floating structure if the static restoring force by mooring lines is very small, it is possible that a highly tuned resonance generally occurs at very low natural frequencies in horizontal plane. In irregular waves this resonance will be excited by the slowly varying second order wave excitation which corresponds to the drifting force in regular waves. Thus, for the design of mooring lines it is necessary to include these forces in the total load acting on a structure moored by chains or cables.

Up to now, several investigations associated with these second order responses (forces or motions) have been done.

These studies can be classified as follows:

1) Deterministic manner based on the numerical simulations.
2) Nondeterministic manner based on the stochastic process.

The former is the numerical prediction method based on the solution of the

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time dependent motion equation taking into account of both the first order forces due to wave elevation and the second order forces which are obtained from the direct integration of all second order pressures over the instantaneous wetted portions of the hull surface. But since the nonlinear second order response depends on the random phases of waves the results derived by this manner are nothing but one sample.

Therefore, in order to estimate the extreme value by means of this manner numerous results of numerical simulations are required.

The latter is the statistical prediction method based on the Volterra functional series expansion of responses to the given wave or force excitation. The advantage of this method is that it is easily adaptable to the physical interpretation associated with the usual perturbation expansion solution of nonlinear response problems because the n-th term in such a expansion gives the response component resulting form n-th order interaction of the excitation.

Hasselmann\textsuperscript{13} outlined the functional series approach to ship motions and showed that the nonlinear transfer functions were related to higher-order moments of ship motions. Dalzel\textsuperscript{2} formulated the added resistance in waves as the quadratic functional series and estimated the mean added resistance transfer function in irregular waves.

Neal\textsuperscript{3} formulated the exact probability density function of second order responses by using the statistical theory of quadratic form.

Vinje\textsuperscript{4} obtained the approximate expressions of peak distributions of second order responses by use of the expansion of cumulants. Further he\textsuperscript{5} expanded Neal's formulation to peak distribution under Powell's assumptions. Hineno\textsuperscript{6} applied Vinje's method to the nonlinear wave and the steady tilt problem for a semisubmersible drilling platform.

The present authors et al.\textsuperscript{7} showed that the probability densities of mooring forces on the huge offshore structure can be represented by the finite Gram-Charlier expansion.

Recently, Naess\textsuperscript{8} discussed the dynamic reliability of second order responses under the Poisson distribution by means of Neal's formulation and the slow drift approximation.

The authors\textsuperscript{9} have already shown that the Kac-Siegert method adopted in Neal's work is applicable to the horizontal responses of a moored floating structure and the quadratic transfer function treated in Dalzel's work is required to estimate the higher-order statistical values (variance, skewness and etc.)

As described in the above overview the statistical theory on second order responses is nearly completed. But the discussions for the extreme responses and the approximate theory have not been done sufficiently.
The purpose of this paper is to discuss the extreme values and the approximate statistical theory for the horizontal responses of a moored floating structure.

In chapter 2 the Kac–Sieger theory\textsuperscript{10,11} on second order responses is discussed in details.

In chapter 3 the approximate theory and the estimation method of extreme value in case of the horizontal responses are developed.

In chapter 4 the applicability of the methods introduced in chapter 3 is investigated through the comparisons between the experimental results and the estimated values.

As the results, the following items have been found:

1. The transfer function of the horizontal response to the slowly instantaneous wave energy, which is introduced newly in this case, is able to evaluate quantitatively the characteristics of the quadratic transfer function, and the linear transfer function can be separated from the total response in the frequency domain and can be estimated by the usual linear motion prediction method taking the viscous damping into account.

By using these functions the variance and the skewness which dominate the distribution of the horizontal response can be estimated.

2. In order to obtain the instantaneous probability distribution we propose the approximate method matching between the finite Gram–Charlier expansion and the asymptotic form derived from the Kac–Sieger theory. The estimated results due to the present method show fairly good agreement with the experimental ones.

3. The new prediction methods for the probability distributions of extremal values and the extreme response are proposed under the assumptions that the response displacement and velocity are independent mutually and the response velocity is of Gaussian distribution with zero mean in addition to the Powell’s assumptions. As a result, it is confirmed that the Longuet-Higgin’s method significantly underestimates the experimental result while the present method is in good agreement with the experimental one.

2. EXACT STATISTICAL THEORY

2.1 Basic Assumptions

The assumption of this theory is that the nonlinear responses can be represented by the functional power series (or functional polynomials).

Let $x(t)$ denote the nonlinear response of a moored floating structure to a random excitation $\{\xi(t) \mid t \in R^1\}$. Since $x(t)$ may be the responses to the entire time history of $\xi(t)$, we call $x(t)$ a functional defined on a class of excitation functions $\xi(t)$ as
\[ x(t) = F[\xi(t)] \]  

(2.1.1)

If \( F \) is a continuous functional of \( \xi(t) \) in the function space sense, then \( F \) can be expanded in a functional power series such that

\[ x(t) = \sum_{n=0}^{\infty} \int_{\tau_1}^{\tau_n} \ldots \int_{\tau_1}^{\tau_n} \hat{g}_n(t, t_1, \ldots, t_n) \xi(t_1) \ldots \xi(t_n) \, dt_1 \ldots dt_n. \]  

(2.1.2)

If this series represents a causal physical system, then the kernel functions satisfy

\[ \hat{g}_n(t, t_1, \ldots, t_n) = 0, \text{ for } t_i > t. \]  

(2.1.3)

Series satisfying this property were studied by Volterra\(^{12}\), and series of the form (2.1.2) that satisfy Eq. (2.1.3) are called Volterra series.

If the nonlinear system is time invariant, then kernel functions in Eq. (2.1.2) depend only on time difference. Thus,

\[ x(t) = \sum_{n=1}^{\infty} \int_{\tau_1}^{\tau_n} \ldots \int_{\tau_1}^{\tau_n} d\tau_1 \ldots d\tau_n g_n(\tau_1, \ldots, \tau_n) \prod_{r=1}^{n} \xi(t - \tau_r) + D.C., \]  

(2.1.4)

where D.C. is a constant.

In general, the kernel functions in Eq. (2.1.4) may not be symmetric functions of their arguments. However, a permutation of indices in any kernel only affects the order in which the integration is carried out but does not affect the response. Thus, for the purpose of analysis, symmetric kernel may be assumed without loss of generality.

If the kernels are continuous and absolutely integrable and if the input is bounded and the contribution from terms of order \( n \) in Eq. (2.1.4) decreases to zero as \( n \to \infty \), then it is proved that the functional power series (2.1.4) converge uniformly.

We shall limit our analysis to include excitation effects through second order except for D.C.. Thus Eq. (2.1.4) is truncated at \( n = 2 \) and takes the following form:

\[ x(t) = \int d\tau g_1(\tau) \xi(t - \tau) + \int d\tau_1 \int d\tau_2 g_2(\tau_1, \tau_2) \xi(t - \tau_1) \xi(t - \tau_2) \]  

(2.1.5)

If \( \xi(t) \) is the wave excitation, this series can be used to analyze the response that is proportional to either the wave height or the squared wave height.

If the kernels in Eq. (2.1.5) are continuous and absolutely integrable, then they possess Fourier transform. The transform pairs are defined as follows:

\[ g_1(\tau) = \frac{1}{2\pi} \int_{-\infty}^{\infty} \exp(i\omega \tau) G_1(\omega) \, d\omega, \]

\[ G_1(\omega) = \int_{-\infty}^{\infty} \exp(-i\omega \tau) g_1(\tau) \, d\tau, \]  

(2.1.6)

\[ g_2(\tau_1, \tau_2) = \frac{1}{(2\pi)^2} \int_{\omega_1} d\omega_1 \int_{\omega_2} d\omega_2 \exp \{i(\omega_1 \tau_1 + \omega_2 \tau_2)\} G_2(\omega_1, \omega_2), \]
\[ G_2(\omega_1, \omega_2) = \int_{\tau_1} d\tau_1 \int_{\tau_2} d\tau_2 \exp \{-i(\omega_1 \tau_1 + \omega_2 \tau_2)\} g_2(\tau_1, \tau_2). \]

In Eq. (2.1.5) the kernel \( g_1 \) is a linear impulse response function, and its transform, \( G_1 \), is a linear transfer function. The kernel \( g_2 \) is analogous to the linear impulse response function and is called "quadratic impulse response function". Its transform, \( G_2 \), is called "quadratic transfer function". Tick\(^{13}\) has called Eq. (2.1.5) as a time-invariant quadratic system since it includes both a first order and a second order term.

Since the kernel \( g_2(\tau_1, \tau_2) \) can be assumed to be symmetrical in its arguments
\[ g_2(\tau_1, \tau_2) = g_2(\tau_2, \tau_1), \quad (2.1.7) \]
thus
\[ G_2(\omega_1, \omega_2) = G_2(\omega_2, \omega_1). \quad (2.1.8) \]

Consequently, the quadratic transfer function is symmetrical about the line \( \omega_1 = \omega_2 \) in the \((\omega_1, \omega_2)\) plane.

### 2.2 Transfer Functions and their Physical Properties

It is assumed that the surface elevation \( \xi(t) \) is a stationary Gaussian process with zero mean. The auto-correlation function of the process will be denoted as \( R_\xi(t) \) and is defined as follows:
\[ R_\xi(\tau) = E\{ \xi(t) \xi(t+\tau) \}, \quad (2.2.1) \]
where \( E\{\ldots\} \) denotes the ensemble average (or expectation).

If \( R_\xi(\tau) \) is absolutely integrable, then a continuous nonnegative spectral density function \( S_\xi(\omega) \) exists and satisfies
\[ R_\xi(\tau) = \int_\omega \exp(i\omega \tau) S_\xi(\omega) \, d\omega, \quad (2.2.2) \]
\[ S_\xi(\omega) = 1/2\pi \int_\tau \exp(-i\omega \tau) R_\xi(\tau) \, d\tau, \quad (2.2.3) \]
where \( S_\xi \) is the two sided wave spectral density function defined over doubly infinities. They are called the Wiener-Khintchine relations.

Taking the expected value of Eq. (2.1.5) we have
\[ E\{x\} = \bar{x} = \int_{\tau_1} d\tau_1 \int_{\tau_2} d\tau_2 g_2(\tau_1, \tau_2) R_\xi(\tau_1 - \tau_2). \quad (2.2.4) \]

Applying Parseval's formula and using the Dirac delta function \( \delta(\omega) \), then Eq. (2.2.4) can be written as:
\[ E\{x\} = \int_\omega d\omega G_2(\omega, -\omega) S_\xi(\omega) \]
\[ = \int_0^\infty d\omega G_2(\omega, -\omega) U_\xi(\omega), \quad (2.2.5) \]
where \( U_\xi \) is the one sided wave spectral density function defined over non-negative frequencies by

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\[
U_\xi(\omega) = \begin{cases} 
2S_\xi(\omega) & \text{for } 0 \leq \omega < \infty \\
0 & \text{otherwise}
\end{cases}
\] (2.2.6)

Pinkster\(^4\) has shown that the time average \( \overline{x} \) in Eq. (2.1.5) represents the mean drift displacement due to the steady force.

According to his results, the following relation is satisfied:

\[
\overline{x} = 2 \int_0^\infty d\omega H_L(\omega) F_D(\omega) U_\xi(\omega),
\] (2.2.7)

where \( H_L \) is the linear transfer function of displacement to the external force and \( F_D \) is the drift force coefficient in regular waves.

Equating Eq. (2.2.5) and Eq. (2.2.7) the following relation

\[
G_2(\omega, -\omega) = 2H_L(\omega) F_D(\omega)
\] (2.2.8)

is found out. Thus, it is found that \( G_2(\omega, -\omega) \) represents the mean drift displacement.

Using the fact that \( \xi(t) \) is a Gaussian process with zero mean, the second term in eq. (2.1.5) becomes

\[
\int_{\tau_1} d\tau_1 \int_{\tau_2} d\tau_2 g_2(\tau_1, \tau_2) \xi(t-\tau_1) \xi(t-\tau_2)
= 1/2R_\xi \int_0^\infty \int_0^\infty G_2(\omega_1, \omega_2) \exp \{i(\omega_1 + \omega_2)t - i(\epsilon_1 + \epsilon_2)\}
\times \sqrt{2U_\xi(\omega_1)2U_\xi(\omega_2)d\omega_1d\omega_2}
+ 1/2R_\xi \int_0^\infty \int_0^\infty G_2(\omega_1, -\omega_2) \exp \{i(\omega_1 - \omega_2)t - i(\epsilon_1 - \epsilon_2)\}
\times \sqrt{2U_\xi(\omega_1)2U_\xi(\omega_2)d\omega_1d\omega_2},
\] (2.2.9)

where \( \epsilon_i \) are the random phases and statistically independent.

The first term in (2.2.9) represents the contribution of sums of frequency pairs to the second order response, whereas the second term gives the contribution of differences of wave frequency pairs.

Newman\(^2\) defined the second term as slowly varying second order response.

From this result it is found that \( G_2(\omega_1, -\omega_2) \) represents the property of slowly varying second order response.

2.3 Transfer Functions and Response Spectrum

2.3.1 Cross Spectrum and Auto Spectrum

Forming the cross correlation function between \( x(t) \) and \( \xi(t) \) from Eq. (2.1.5):

\[
E \{(x(t) - \overline{x}) \xi(t-\tau)\} = \int_{\tau_1} d\tau_1 g_1(\tau_1) E \{\xi(t-\tau_1) \xi(t-\tau)\}
+ \int_{\tau_1} d\tau_1 \int_{\tau_2} d\tau_2 g_2(\tau_2, \tau_2) E \{(t-\tau_1)
\times \xi(t-\tau_2) \xi(t-\tau)\} - \overline{x} E \{\xi(t-\tau)\}
\] (2.3.1)
Because the wave is defined to be zero-mean, the last two terms are zero. Thus:

$$E \{ (x(t) - \bar{x}) \xi(t - \tau) \} = \int_{\tau_1} d\tau_1 g_1(\tau_1) E \{ \xi(t - \tau_1) \xi(t - \tau) \} \quad (2.3.2)$$

This result means that the cross spectrum between response $x(t)$ and wave $\xi(t)$ involves only the first (linear) term of the functional polynomials, and thus that the linear transfer function $G_1(\omega)$ is derivable by standard cross-spectral technique. If the cross spectrum is denoted as $S_{x\xi}(\omega)$; then:

$$G_1(\omega) = S_{x\xi}(\omega) / S_\xi(\omega) \quad (2.3.3)$$

Forming the auto correlation function of $x(t)$:

$$E \{ (x(t) - \bar{x})(x(t + \tau) - \bar{x}) \} = \int_{\tau_1} d\tau_1 \int_{\tau_2} d\tau_2 g_1(\tau_1) g_1(\tau_2)$$

$$\times E \{ \xi(t - \tau_1) \xi(t + \tau - \tau_2) \}$$

$$+ \int_{\tau_1} d\tau_1 \cdots \int_{\tau_4} d\tau_4 g_2(\tau_1, \tau_2) g_2(\tau_3, \tau_4)$$

$$\times E \{ \xi(t - \tau_1) \xi(t - \tau_2) \xi(t + \tau - \tau_3) \xi(t + \tau - \tau_4) \} - \bar{x}^4, \quad (2.3.4)$$

and using the factorization relation for higher-order moments of Gaussian process as

$$E \{ X_1X_2X_3X_4 \} = E \{ X_1X_2 \} E \{ X_3X_4 \} + E \{ X_1X_3 \} E \{ X_2X_4 \}$$

$$+ E \{ X_1X_4 \} E \{ X_2X_3 \}, \quad (2.3.5)$$

we obtain

$$R_{xx}(\tau) = \int_{\tau_1} d\tau_1 \int_{\tau_2} d\tau_2 g_1(\tau_1) g_1(\tau_2) R_\xi(\tau + \tau_1 - \tau_2)$$

$$+ \int_{\tau_1} d\tau_1 \cdots \int_{\tau_4} d\tau_4 g_2(\tau_1, \tau_2) g_2(\tau_3, \tau_4) \{ R_\xi(\tau + \tau_1 - \tau_2) R_\xi(\tau + \tau_2 - \tau_4)$$

$$+ R_\xi(\tau + \tau_1 - \tau_4) R_\xi(\tau + \tau_2 - \tau_3) \}. \quad (2.3.6)$$

The auto power spectrum of $x(t)$ is the Fourier transform of $R_{xx}(\tau)$ and is computed from Wiener-Khintchine relations as

$$S_x(\omega) = |G_1(\omega)|^2 S_\xi(\omega) + 2 \int_0^\infty d\xi |G_2(\omega - \xi, \xi)|^2$$

$$\times S_\xi(\omega - \xi) S_\xi(\xi). \quad (2.3.7)$$

### 2.3.2 Cross Bispectrum

Tick\textsuperscript{33} defined the cross bispectrum as the two dimensional Fourier transform of a third moment function $R_{\xi\xi\xi}(\tau_1, \tau_2)$ which is defined as

$$R_{\xi\xi\xi}(\tau_1, \tau_2) = E \{ \xi(t + \tau_1) \xi(t - \tau_1) (x(t - \tau_2) - \bar{x}) \}. \quad (2.3.8)$$
Noting that the expected value of odd products of Gaussian variable are zero, then:

\[
R_{\xi\xi}(\tau_1, \tau_2) = 2 \int_{t_1} dt_1 \int_{t_2} dt_2 \mathcal{G}_\xi(t_1, t_2) R_\xi(t_1 + (\tau_1 + \tau_2)) R_\xi(t_2 + (\tau_2 - \tau_1))
\]

(2.3.9)

Utilizing Parseval's formula and the relation in which cross bispectrum is defined by the double Fourier transform of \(R_{\xi\xi}\), we obtain

\[
C_{\xi\xi}(\omega_1 - \omega_2, \omega_1 + \omega_2) = G_\xi^* (\omega_1, \omega_2) S_\xi(\omega_1) S_\xi(\omega_2),
\]

(2.3.10)

where \(C_{\xi\xi}\) is the cross bispectrum and \(^*\) denotes the complex conjugate. Thus the quadratic transfer function is obtained from (2.3.10) as

\[
G_\xi(\omega_1, \omega_2) = C_{\xi\xi}^* (\omega_1 - \omega_2, \omega_1 + \omega_2) / S_\xi(\omega_1) S_\xi(\omega_2).
\]

(2.3.11)

### 2.4 Instantaneous Probability Density Function

We shall consider the distribution of the time invariant quadratic system subject to a stationary Gaussian excitation.

If the system is represented by Eq. (2.1.5), then from Appendix A the characteristic function can be written as

\[
\phi(\theta) = \prod_{j=1}^{\infty} \phi_j(\theta) = \prod_{j=1}^{\infty} (1 - 2i\lambda_j \theta)^{-1/2} \exp \left\{-c_j^2 \theta^2 / 2(1 - 2i\lambda_j \theta) \right\}.
\]

(2.4.1)

Since the instantaneous probability density function is defined by the Fourier transform of the characteristic function, it becomes

\[
p_\xi(x) = 1 / 2\pi \int_{-\infty}^{\infty} d\theta \exp \{-i\theta x\} \phi(\theta),
\]

(2.4.2)

where \(\lambda_j\) are real eigenvalues given by the following integral equation

\[
\int_{-\infty}^{\infty} d\nu S_\xi(\omega) G_\xi(\omega, -\nu) \Psi_n(\nu) = \lambda_n \Psi_n(\omega),
\]

(2.4.3)

where \(\Psi_n(\omega)\) are the eigenfunctions satisfying the orthogonal relation as

\[
\int_{-\infty}^{\infty} d\omega_1 \int_{\omega_2}^{\omega_1} d\omega_2 \Psi_n(\omega_1) \Psi_m(\omega_2) G_\xi(-\omega_1, -\omega_2) = \lambda_n \delta_{nn}.
\]

(2.4.4)

The parameter \(c_j\) representing the linear response can be obtained from

\[
c_j = \int_{-\infty}^{\infty} d\omega G_1(\omega) \Psi_j(\omega).
\]

(2.4.5)

From Appendix B the statistical values up to third order can be obtained as

\[
E\{x\} = \sum_{i=1}^{\infty} \lambda_i = \int_{-\infty}^{\infty} d\omega G_\xi(\omega, -\omega) S_\xi(\omega),
\]

(2.4.6)

\[
\sigma^2 = \sum_{i=1}^{\infty} c_i^2 + 2 \sum_{i=1}^{\infty} \lambda_i = \int_{-\infty}^{\infty} d\omega |G_1(\omega)|^2 S_\xi(\omega)
\]

\[
+ 2 \int_{-\infty}^{\infty} d\omega \int_{-\infty}^{\infty} d\nu |G_\xi(\omega, \nu)|^2 S_\xi(\omega) S_\xi(\nu),
\]

(2.4.7)

\[
\mu^2 \sigma^2 = 8 \sum_{i=1}^{\infty} c_i \lambda_i + 8 \sum_{i=1}^{\infty} c_i^2 \lambda_i = 8 \int_{-\infty}^{\infty} d\omega_1 \int_{\omega_2}^{\omega_1} d\omega_2 \int_{\omega_3}^{\omega_2} d\omega_3 G_\xi(\omega_1, \omega_2).
\]

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\begin{align*}
&\times G_3^{\frac{1}{3}}(\omega_3, \omega_3) G_2(\omega_3, -\omega_1) S_3^\mu(\omega_1) S_3^\nu(\omega_2) S_3^\nu(\omega_3) + 6 \int_{\omega_1} d\omega_1 \int_{\omega_2} d\omega_2 \\
&\times G_1(-\omega_1) G_1(-\omega_2) G_2(\omega_1, \omega_2) S_3^\mu(\omega_1) S_3^\nu(\omega_2),
\end{align*}

where \( \sigma \) and \( \mu \) are the variance and the skewness of \( x(t) \), respectively.

Eqs.\((2.4.6), (2.4.7)\) and \((2.4.8)\) show the most important relations between the transfer functions of second order response and statistical values.

2.5 Probability Density Functions of Extremal Values and Extreme Value

2.5.1 Probability Density Function of Extremal Values

First we will define that the extremal values consist of the maxima and the minima of a random process while the extreme value is the largest value of the maxima or the minima that will occur in some observations.

It has been known that statistical prediction of the extremal values of a random process may be made by using the Rayleigh distribution, if the following conditions are satisfied:

1) Random process is a stationary Gaussian process with zero mean.
2) Extremal values are statistically independent.
3) Linearity must hold between input and output processes.

But since in the case of the horizontal response of a moored floating structure the linearity is not satisfied as described in the previous section, the Rayleigh distribution may no longer be applicable for predicting properties of the extremal values. Thus, the new prediction method of probability density function of extremal values is required.

![Fig. 1](image)

**Fig. 1** Explanatory sketch of a random process \( x(t) \)

Fig. 1 is an explanatory sketch of a random process \( x(t) \) for which the extremal values could be anywhere in the range of \( (-\infty, \infty) \) and several extremal values could occur during one cycle as defined by mean crossings. Here, the extremal values called "maxima" are defined as peaks which satisfy the condition \( \dot{x}(t) = 0 \) and \( \ddot{x}(t) < 0 \). Whereas "minima" are defined as troughs satisfying the condition \( \dot{x}(t) = 0 \) and \( \ddot{x}(t) > 0 \). As shown in Fig. 1 maxima and minima can be both negative values and positive values. The magnitude of the
maxima with positive values \( \{ x(t) > 0, \dot{x}(t) = 0, \ddot{x}(t) < 0 \} \) or the minima with negative values \( \{ x(t) < 0, \dot{x}(t) = 0, \ddot{x}(t) > 0 \} \) would be critical if they exceed a certain value, and hence the statistical extreme values of these maxima and the minima provide valuable information for the engineering design purpose.

For the problem of a mooring system the positive maxima are the most important, if the direction drifted by waves is positive. Since the statistical properties of negative minima can be estimated from those of positive maxima by means of the transform of random variables, the latter are considered in the following analysis.

It can be assumed that \( x(t) \) is stationary and zero mean without loss of generality. Then the expected number of maxima above a specified level \( x(t) = \xi \), denoted as \( E \{ M(\xi) \} \), is obtained by

\[
E \{ M(\xi) \} = \int_{-\infty}^{\infty} dx \int_{-\infty}^{0} \ddot{x} | p_{x\ddot{x}}(x, 0, \ddot{x}) d\ddot{x}. \tag{2.5.1}
\]

The expected number of maxima with positive values, denoted as \( E \{ M(0) \} \), then becomes

\[
E \{ M(0) \} = \int_{0}^{\infty} dx \int_{-\infty}^{0} \ddot{x} | p_{x\ddot{x}}(x, 0, \ddot{x}) d\ddot{x}, \tag{2.5.2}
\]

where \( p_{x\ddot{x}} \) is the joint probability density function of \( x, \dot{x} \) and \( \ddot{x} \).

Huston and Skopinski\(^{16}\) has assumed that the ratio of their two expected number is approximately equivalent to the probability in which the maximum values exceed a level \( \xi \).

Under this assumption the probability in which the maximum positive values exceed a level \( \xi \) becomes

\[
P_{b}(\xi) = 1 - E \{ M(\xi) / M(0) \} \sim 1 - E \{ M(\xi) \} / E \{ M(0) \}. \tag{2.5.3}
\]

Then the probability density function of the maxima is given by

\[
p_{b}(\xi) = \frac{1}{E \{ M(0) \}} \int_{0}^{\infty} d\ddot{x} \dddot{x} p_{x\ddot{x}\dddot{x}}(\xi, 0, \dddot{x}). \tag{2.5.4}
\]

In the case that \( x(t) \) is the Gaussian process \( p_{b} \) already has been obtained by Cartwright and Longuet Higgins\(^{16}\). But in the case of nonlinear response \( p_{b} \) has not been found out yet.

When \( x(t) \) is narrow banded, Powell\(^{17}\) has proposed the following assumptions:

1) The response is narrow banded. That is, the negative maxima and the positive minima are negligible.
2) The response is stationary.
3) The random number crossing a specified level at positive gradient is equal to one of maxima above it.

Here, the random number crossing a specified level \( \xi \) at positive gradient is
defined as

$$N^+(\xi) = \dot{x}\delta(x - \xi).$$  \hfill (2.5.5)

From these assumptions

$$M(\xi) \simeq N^+(\xi).$$  \hfill (2.5.6)

Thus, Eq. (2.5.4) becomes

$$p_\eta(\xi) \approx -\frac{1}{E\{M(0)\}} \cdot \frac{d}{d\xi} \int_0^\infty d\dot{x} p_{x\dot{x}}(\xi, \dot{x}),$$  \hfill (2.5.7)

where $p_{x\dot{x}}$ is the joint probability density function of $x$ and $\dot{x}$.

Powell also indicated these assumptions can even be applied to the case in which the response is wide banded. Because the positive minima and the negative maxima are negligible since these values do not exist at which the threshold level is sufficiently large.

In general, since $E\{M(0)\} \geq E\{N^+(0)\}$, the probability of maxima is overestimated in Powell’s assumptions.

### 2.5.2 Extreme Value

In this section, the extreme value will be derived by applying the order statistics. The extreme value defined here is the largest value of the maxima that will occur in $N$ observations.

Let $(y_1, y_2, \ldots, y_N)$ be an ordered sample of size $N$, where $y_i$ is the observed maxima of a random process $x(t)$, then all $y_i$ have the same probability density function given in (2.5.7). Let $(\eta_1, \eta_2, \ldots, \eta_N)$ be an ordered sample of $y_i$ with $\eta_1 \leq \eta_2 \leq \ldots \leq \eta_N$, then each $\eta$ can be regarded as the output of an independent random variable $z_i$. Thus the random variable $z_N$, which is the largest $\eta_N$ in the ordered sample, has the following probability density function:

$$g(\eta_N) = N f(\eta_N) \{F(\eta_N)\}^{N-1},$$  \hfill (2.5.8)

where

$g(\eta_N)$: Probability density function of the largest value in $N$ observations,
$f(\eta_N)$: Probability density function given by replacing $\xi$ with $\eta_N$ in Eq. (2.5.7),
$F(\eta_N)$: Cumulative distribution function given by integrating Eq. (2.5.7) with respect to $\xi$ and replacing $\xi$ with $\eta_N$,
$N$: The number of observations.

Thus the extreme value estimate is obtained by

$$E\{z_N\} = \int_0^\infty d\eta_N \eta_N g(\eta_N).$$  \hfill (2.5.9)
3. APPROXIMATE THEORY

3.1 Approximation to the Quadratic Transfer Function

If the horizontal response of a moored floating structure is expressed by (2.1.5) the second term in Eq. (2.1.5) represents the response which is in proportion to the squared wave height, that is, instantaneous wave energy. The authors have introduced the linear response function \( g \) between the horizontal response and the instantaneous wave energy, and have shown that \( g \) is approximately equivalent to the quadratic transfer function. Here, we shall quote their results.

The cross correlation function \( R_{\xi \xi'}(\tau) \) between the instantaneous wave energy \( \xi^2(t) \) and the second order nonlinear response \( x(t) \) is defined by

\[
R_{\xi \xi'}(\tau) = \langle (x(t) - \bar{x}) (\xi^2(t + \tau) - \bar{\xi^2}) \rangle.
\]  
(3.1.1)

Applying Parseval's formula, the cross spectrum \( S_{\xi \xi'} \) is given by

\[
S_{\xi \xi'}(\omega) = 2 \int_v d\nu S_{\xi}(\omega - \nu) S_{\xi}(\nu) G_{\xi}(\omega - \nu, \nu),
\]  
(3.1.2)

and the auto power spectrum of instantaneous wave energy is obtained from

\[
S_{\xi}(\omega) = 2 \int_v d\nu S_{\xi}(\omega - \nu) S_{\xi}(\nu).
\]  
(3.1.3)

Thus, \( g \) representing the response property between the instantaneous wave energy and \( x(t) \) is given as follows:

\[
g = \frac{\int_v d\nu S_{\xi}(\omega - \nu) S_{\xi}(\nu) G_{\xi}(\omega - \nu, \nu)}{\int_v d\nu S_{\xi}(\omega - \nu) S_{\xi}(\nu)}
\]  
(3.1.4)

From Eq. (3.1.4) it is found that \( g \) indicates the average of \( G_\xi \) with respect to the instantaneous wave energy spectrum and depends on the wave spectral density. If the following identity

\[
\int_{\omega_1} d\omega_1 \int_{\omega_2} d\omega_2 g(\omega_1) S_{\xi}(\omega_1 - \omega_2) S_{\xi}(\omega_2)
= \int_{\omega_1} d\omega_1 \int_{\omega_2} d\omega_2 S_{\xi}(\omega_1 - \omega_2) S_{\xi}(\omega_2) G_{\xi}(\omega_1 - \omega_2, \omega_2)
\]  
(3.1.5)

is satisfied, we obtain

\[
g^*(\omega_1 + \omega_2) \simeq G_{\xi}(\omega_1, \omega_2).
\]  
(3.1.6)

This means that \( g \) is approximately equal to the quadratic transfer function under a fixed wave spectrum.

If this approximation is applicable, we have

\[
x(t) = \int_\tau d\tau g(\tau) \xi(t - \tau) + 1/\pi \int_\tau d\tau g^2(\tau) \xi^2(t - \tau),
\]  
(3.1.7)

where
\[ f_{\xi}(\tau) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} d\omega \ g(\omega) \exp(i\omega\tau). \quad (3.1.8) \]

Thus the second term in (3.1.7) can be regarded as the output through a filtered square law detector and \( f_{\xi}(\tau) \) is interpreted as the filter impulse response.

### 3.2 Approximation to the Instantaneous Probability Density Function

From Appendix B the horizontal response of a moored floating structure in irregular waves is alternatively represented in the following form:

\[ x = \sum_{i=1}^{\infty} (c_i X_i + \lambda_i X_i \xi). \quad (3.2.1) \]

where \( X_i \) are standard Gaussian random variables with zero mean and their variances are equal to unity.

Here, we shall assume that the number of eigenvalues \( \lambda_i \) is finite and sufficiently large.

If we introduce the new random variables \( Z_i \) as

\[ Z_i = c_i X_i + \lambda_i X_i \xi, \quad (3.2.2) \]

then it is shown that \( Z_i \) and \( Z_i(i \neq j) \) are mutually independent and each \( Z_i \) has the same probability density function. Thus, if the higher moments of \( Z_i \) are finite, \( x(t) \) is subject to the local limit theorem\(^{19}\), that is, \( x(t) \) will asymptotically become Gaussian.

Now, we replace \( x - \bar{x} \) by \( y \) and introduce the error function \( \epsilon(y) \) defined by

\[ \epsilon(y) = p_x(y) - \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{y^2}{2\sigma^2}\right). \quad (3.2.3) \]

If \( \epsilon \) can be represented by the family of orthogonal functions with weighting function, \( \{ w(y) h_n(y) \} \), it can be expanded in the following form:

\[ \epsilon(y) = \sum_{n=1}^{\infty} \alpha_n h_n(y) w(y), \quad (3.2.4) \]

where \( h_n \) are the orthogonal functions and \( w(y) \) is the weighting function, and where \( \alpha_n \) are the coefficients given by

\[ \alpha_n = \int_{-\infty}^{\infty} dy h_n(y) \epsilon(y). \quad (3.2.5) \]

If \( w(y) \) is the Gaussian density function it is well known that \( h_n(y) \) are given by the Hermite polynomials\(^{19}\).

From the properties of the Hermite polynomials the instantaneous probability density function can be approximated by the Gram–Charlier expansion\(^9\):

\[ p_x(x) \approx \frac{1}{\sqrt{2\pi\sigma_x}} \left\{ 1 + \sum_{n=3}^{\infty} b_n / (n! \sigma^2) H_n((x-\bar{x})/\sigma_x) \right\} \times \exp\left(- (x-\bar{x})^2/2\sigma^2\right), \quad (3.2.6) \]

where \( H_n \) are the Hermite polynomials and \( b_n \) represent the higher moments defined by

(339)
\[ b_n = E \{ (x-x)^n \} \quad \text{for } n \geq 3. \] (3.2.7)

From Eqs. (3.2.7) and (2.4.8) the following relation is satisfied:
\[ b_3 = \sigma_{\frac{\lambda}{\mu}}. \] (3.2.8)

If Eq. (3.2.6) is truncated at finite order, the finite Gram-Charlier expansion cannot represent the asymptotic behaviour of the probability density as \(|x| \to \infty\). Thus it is necessary to investigate it from the exact solution.

We assume that \(\lambda_i\) can be ordered and \(\lambda_1\) (max \(\lambda_i\)) is positive and \(\lambda_2\) (min \(\lambda_i\)) is negative. For \(x \to -\infty\) the integration path of Eq. (2.4.2) can be deformed such that it goes along the branch-cuts in the left half plane and the main contribution to this integral will be that from the neighbourhood of the branch-point closest to the imaginary axis (see Appendix C).

We introduce the following function\(^9\):
\[ \tilde{\phi}(\theta) = (1-2i\lambda, \theta)^{-1/2} \exp \left\{ -\lambda \theta^2 / 2(1-2i\lambda, \theta) \right\} \prod_{j=2}^{\infty} \phi_j(-i/2\lambda_1). \] (3.2.9)

Since \(\phi / \tilde{\phi}\) is analytical in the vicinity of the branch point closest to the imaginary axis, the first approximation to \(p_x(x)\) can be found as
\[ p_x(x) \approx 1/2 \pi \int_{-\infty}^{\infty} d\theta \tilde{\phi}(\theta) \exp\left(-i\theta x \right) \prod_{j=2}^{\infty} \phi_j(-i/2\lambda_1). \] (3.2.10)

Since \((Z_i/\lambda_1 + c_{i^2}/4\lambda_i^2)\) is of non-central \(\chi^2\) distribution with one degree of freedom, the following form\(^9\) is found out by means of the asymptotic expansion of the non-central \(\chi^2\) distribution as \(x \to -\infty\):
\[ p_x(x) \approx 1/\sqrt{2\pi \lambda_1 x} \exp \left\{ - (x+c_{\lambda_1}/2\lambda_1)/2\lambda_1 \right\} \cosh(\sqrt{x}/\lambda_1 c_{\lambda_1}/2\lambda_1) \] (3.2.11)

The same expression can easily be derived for \(x \to -\infty\).

These results show that \(x/\lambda_1\) is asymptotically of \(\chi^2\) distribution with one degree of freedom and with a slight modification caused by the linear term.

If the second term in (2.1.5) is dominant, from Appendix C we get:
\[ p_x(x) \approx 1/2 \pi \sqrt{\lambda_1 / \lambda_2} \exp \left\{ (\lambda_1 - |\lambda_2|) x / 4\lambda_1 |\lambda_2| \right\} \times K_0 \left( |x| (\lambda_1 + |\lambda_2|) / 4\lambda_1 |\lambda_2| \right) \quad \text{as } |x| \to -\infty \] (3.2.12)

where \(K_0\) is the modified Bessel function of the second kind.

Thus, the approximate solution for the instantaneous probability density function will be obtained by matching between the finite Gram-Charlier expansion and the asymptotic form derived from the exact solution.

### 3.3 Approximation to the Probability Density Function of Extremal Values

By expanding the one dimensional Gram-Charlier expansion to the two dimensional form joint probability density function of \(x\) and \(\dot{x}\) can be approximately represented in the following form:

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\[ p_{x\dot{x}}(x, \dot{x}) = \frac{1}{2\pi \sigma_x \sigma_{\dot{x}}} \exp \left\{ -\frac{(x - \bar{x})^2}{2\sigma_x^2} - \frac{\dot{x}^2}{2\sigma_{\dot{x}}^2} \right\} \sum_{m,n} b_{mn} \times H_m \left( \frac{x - \bar{x}}{\sigma_x} \right) H_n \left( \frac{\dot{x}}{\sigma_{\dot{x}}} \right) , \]  
(3.3.1)

where \( b_{mn} \) is a function of the higher moments of \( x \) and \( \dot{x} \).

The above equation is called "two dimensional Gram-Charlier expansion" and is equivalent to the results given from the series expansion of cumulants by Vinje\(^4\).

In the case of the nonlinear response, \( x \) and \( \dot{x} \) are not mutually independent even though \( x \) is stationary. But since the low frequency response of a moored floating structure is limited in the low frequency regions, it will be expected that the contribution of the low frequency response to the motion velocity is very small. Thus \( \dot{x} \) may be expressed as

\[ \dot{x} \simeq \int_\tau d\tau g_\tau(\tau) \ddot{\xi}(t - \tau), \]  
(3.3.2)

where dots denote time derivatives.

Since the surface elevation \( \xi(t) \) is assumed to be the stationary Gaussian process with zero mean, \( \dot{\xi}(t) \) and \( \ddot{\xi}(t) \) are mutually independent.

Since it will be proper to assume that \( \dot{x} \) is Gaussian distributed and \( x \) and \( \dot{x} \) are mutually independent,

we obtain the following form:

\[ p_{x\dot{x}}(x, \dot{x}) = \frac{1}{2\pi \bar{x} \sigma_{\dot{x}}} \left\{ 1 + \sum_n b_n/(n! \sigma_{\dot{x}}^n) \right\} \left[ H_n \left( \frac{x - \bar{x}}{\sigma_x} \right) \right]
\times \exp \left\{ -\frac{(x - \bar{x})^2}{2\sigma_x^2} - \frac{\dot{x}^2}{2\sigma_{\dot{x}}^2} \right\} , \]  
(3.3.3)

where \( \bar{x} \) and \( \sigma_x \) are the mean and the variance of \( x(t) \) respectively and \( \sigma_{\dot{x}} \) is the variance of \( \dot{x} \).

Then replacing \( \xi \) by \( \eta + \bar{x} \) in Eq. (2.5.7), the probability density function of maxima can be represented as:

\[ p_p(\eta) = -\left\{ -\eta \sigma_x^2 \exp(-\eta^2/2\sigma_x^2) \right\} \left\{ 1 + \sum_n b_n/(n! \sigma_x^n) \right\} H_n(\eta/\sigma_x) \]
\[ + \exp(-\eta^2/2\sigma_x^2) \sum_n b_n/(n! \sigma_x^{n+1}) H_n(\eta/\sigma_x) \]
\[ \times \left[ 1 + \sum_n b_n/(n! \sigma_x^n) \right]^{-1} , \]  
(3.3.4)

This means that \( p_p \) is equivalent to the derivative form of Eq. (3.2.6). Thus \( p_p \) for some sufficient large \( \eta \) will asymptotically approach to the following form:

\[ p_p \sim -\frac{d}{d\eta} \left[ \sqrt{\frac{\eta}{\bar{x} + \eta}} \exp(-\eta^2/2\lambda_1) \cosh \left( \sqrt{(\eta + \bar{x})/\lambda_1} (\alpha_1/2\lambda_1) \right) \right. \]
\[ \left. / \cosh \left( \sqrt{\bar{x}/\lambda_1} (\alpha_1/2\lambda_1) \right) \right] , \]  
(3.3.5)

If the low frequency motion is dominant, the asymptotic form for \( \eta \to \infty \)
is given as follows:

\[ p_r = - \frac{d}{d\eta} \left( \exp \left\{ \left( \lambda_1 - |\lambda_2| \right) / 4\lambda_1|\lambda_2| \right\} \times K_0 \left\{ (\eta + \bar{\lambda}) (\lambda_1 + |\lambda_2|) / 4\lambda_1|\lambda_2| \right\} \right) / K_0 \left\{ \bar{\lambda}(\lambda_1 + |\lambda_2|) / 4\lambda_1|\lambda_2| \right\} \]  \hspace{1cm} (3.3.6)

Using Eqs. (3.3.5) of (3.3.6) the extreme responses can be obtained from Eq. (2.5.9).

4. COMPARISONS BETWEEN STATISTICAL ESTIMATIONS AND EXPERIMENTAL RESULTS

4.1 Model Test

4.1.1 Configuration of Model

The model used in the test and the co-ordinate system are shown in Fig.2 and the principal dimensions of the model are given in Table 1.

![Configuration of offshore structure and system of coordinate](image)

**Table 1** Principal dimensions

<table>
<thead>
<tr>
<th>ITEMS</th>
<th>ACTUAL</th>
<th>MODEL</th>
</tr>
</thead>
<tbody>
<tr>
<td>LENGTH (m)</td>
<td>30.0</td>
<td>2.1</td>
</tr>
<tr>
<td>BREADTH (m)</td>
<td>20.0</td>
<td>1.40</td>
</tr>
<tr>
<td>DISPLACEMENT (t)</td>
<td>490.0</td>
<td>0.168</td>
</tr>
<tr>
<td>DRAFT (m)</td>
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<td>0.385</td>
</tr>
<tr>
<td>KG (m)</td>
<td>4.758</td>
<td>0.330</td>
</tr>
<tr>
<td>Kyy</td>
<td>42.6%L</td>
<td>44%L</td>
</tr>
<tr>
<td>GM_L (m)</td>
<td>3.845</td>
<td>0.269</td>
</tr>
<tr>
<td>SCALE RATIO</td>
<td>1.0</td>
<td>1/14.3</td>
</tr>
<tr>
<td>MOORING</td>
<td>CATENARY</td>
<td>LINEAR SPRING</td>
</tr>
</tbody>
</table>

4.1.2 Model Test and Measurement Items

The long duration measurements in irregular waves are required for the low frequency surge motions to get the reliable data in statistics.

Furthermore, the data obtained must contain a number of oscillations with randomness at the frequencies of interest. Therefore, in order to generate the irregular waves over long duration the filtered signals were used, which were obtained by passing the white noize signals generated from a noize generator into the bandpass filter. The rolloff (the cutoff characteristics) of the bandpass filter was 24db/oct. Besides, regular waves and amplitude modulation waves
were also used to investigate the steady drift displacement and the quadratic transfer function of surge motions.

Four kinds of irregular waves were generated. The central frequencies $f_0$ of the bandpass filter used for the generation of each waves were 0.4, 0.5, 0.6, and 0.7 Hz. These frequencies correspond to 9.45, 7.56, 6.30, and 5.40 sec. respectively in the scale of real structures.

In the case of $f_0=0.7$ Hz the duration time was about 5.7 hr in real scale, and for the other cases it was about 2.8 hr. The model tests were carried out at the Mitaka No. 2 Ship Experimental Tank (400 m in length, 18 m in breadth, 8 m in depth) in Ship Research Institute. The test set-up is shown in Fig. 3.

![Set-up of model test](image)

**Fig. 3** Set-up of model test

As shown in Fig. 3, the model was restrained by two soft springs through the device which restricted the yaw motion. Their spring coefficients were 1.683 kg/m. The encounter angle to waves was zero degree.

The measured items are as follows:

- Surge and heave motion measured by the optical motion measuring system using L. E. D.;
- Pitch motion measured by the vertical gyroscope;
- Surface elevation measured by the servo needle wave probe fixed at a position, the x coordinate of which is equal to that of the centre of gravity of the model in still water.

4.2 The Investigation to Irregular Waves

The spectra of irregular waves generated are shown in Fig. 4, where $f_0$ denotes the centre frequency of bandpass filter used for generating the irregular waves and $\sigma^2$ are the estimated variances obtained by integrating the wave spectra with respect to the wave frequency.

The Blackman-Tukey method was used in the spectral analysis, in which the
number of lags was 256 and the window used was the Hamming type. The number of data used for the analysis was about 45000 in the case of $f_0 = 0.7$Hz, and it was about 23000 in the other cases. The sampling interval was 60 msec. in all cases. Fig. 5 shows the instantaneous probability distributions of waves. In Fig. 5 $\bar{\sigma}^2$ and $\bar{\mu}$ is the sample variance and the sample skewness given from the time average respectively.

From Figs. 4 and 5 it is found that the spectrum shapes are different from the standard wave spectra as the I.S.S.C type or the JONSWAP type, which have the narrow band spectra.

$\chi^2$ tests were carried out to test the hypothesis that the wave is of Gaussian process. From these results it has been found that this hypothesis is acceptable at significant level of 0.05.

Next, we shall investigate whether the random phases of waves are strongly homogeneous and uniformly distributed.

In general, even though the wave $\zeta(t)$ satisfies the Ergodicity, $\xi^2$ corresponding to the instantaneous wave energy does not always satisfy it.

This means that the correlation function of $\xi^2(t)$ can not be obtained from the time average, that is, the correlation function of $\xi^2$ obtained from the time average is the sample function because it depends on the random phases of waves, and also the spectrum is the sample. But if the random phases of waves are homogeneous, it may be considered that the time average correlation function represents the average value in the sample functions, or the closest value to the ensemble average.
Thus we compare the spectrum $S_E$ which is given from the time average correlation function of $\xi^2$ with the true spectrum $S_E$ defined in Eq. (3.1.3). These results are shown in Fig. 6. From Fig. 6 it is found that both spectra are in good agreement except for the vicinity of $\omega = 0$. Accordingly it may be assumed that the random phases of waves are nearly homogeneous.

### 4.3 Investigation to Transfer Functions

The surge spectra given in the same manner as the wave spectra are shown in Fig. 7. From this figure it is found that the surge response in the case of $f_0 = 0.7$Hz is the largest and responses are dominated by the low frequency motion. Fig. 8 shows the linear transfer function $G_1$ obtained from the standard cross spectral analysis between the surge motions and the waves. In this figure the solid line represents the theoretical value due to the usual linear motion prediction method which takes into account of the viscous damping. From this figure it is found that the theoretical value is in good agreement with the experimental results.

Thus it is considered that only the linear response can be separated from the surge response including the low frequency motion in the frequency domain.

In order to get the quadratic transfer function of the response the cross

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(345)
Surge spectra

\[ S_x(\omega) \times 10^2 (cm^2/\text{sec}) \]

\[ \omega_{(\text{rad/sec})} \]

\[ f_\omega(\text{Hz}) \]

- \[ 0.4 \]
- \[ 0.5 \]
- \[ 0.6 \]
- \[ 0.7 \]

Fig. 7 Surge spectra

Linear transfer function \((G_1(\omega))\) of surge motion

\[ \text{arg}(G_1) \]

\[ \omega_{(\text{rad/sec})} \]

\[ f_\omega(\text{Hz}) \]

- \[ 0.4 \]
- \[ 0.6 \]
- \[ 0.7 \]
- \[ \text{theory} \]

Fig. 8 Linear transfer function \((G_1(\omega))\) of surge motion

Bispectrum between the waves and the surge motions is required. The utilization of the Fast Fourier Transform (F. F. T) has significant advantage in the computation of the full cross bispectrum. For the present purpose, however, the full computation is not required, because we need only the results on or near \( \omega_1 = \omega_2 \).

Accordingly we used the method developed by Dalzel\(^3\). The window function used in the computation of the cross bispectrum was the Hamming type extended to two dimensions. Fig. 9 shows a part of the results of the cross bispectrum.

The quadratic transfer function obtained from the experiments in amplitude modulation waves (A. M. waves) and irregular waves is shown in Fig. 10. This function in irregular waves was estimated from the cross bispectral analysis within the frequency range corresponding to the 25% power bandwidth of the wave spectrum and that in A. M. waves, which was indicated by the black
circles in the figure, was obtained from the envelope analysis. For each value of the difference frequencies the amplitude parts and the phase parts of the quadratic transfer function are indicated in this figure and the abscissa is based on the mean frequency of the two wave components. The comparisons between $G_2 (\omega, -\omega)$ obtained from the experiments in irregular waves and the steady drift displacement obtained in regular waves are shown in Fig. 11. In this figure the difference between white circles and black circles indicates the effect of wave heights, and the solid line is the value obtained from the theoretical computation taking into account of the hydrodynamic interactions among floating elements under the fixed condition. From Fig. 10 it is found that the result from the tests in amplitude modulation waves are in good agreement with that in irregular waves and the amplitudes of quadratic transfer function decrease with the increase of the difference frequency and the phases do not depend on the mean frequency of the two wave components. From Fig. 11 it is seen that the steady component of quadratic transfer function, $G_2 (\omega, -\omega)$, represents the effect of wave height.

![Cross bispectra (wave-wave-surge)](347)
Fig. 10 Quadratic transfer function $G_2(\omega_1, -\omega_2)$ of surge motion

of the steady drift displacement in regular waves and it is not proportional to the square of wave heights.

Let $G_2(\omega_1, -\omega_2)$ be the quadratic transfer function of the low frequency second order force in head waves and $H_L(\omega)$ be the response function of surge motions to external forces at the low frequency motion.

Then $G_2(\omega_1, -\omega_2)$ can be represented as

(348)
Fig. 11 Comparisons between $G_2(\omega_1, -\omega)$ and the steady drift excursions derived from experimental results in regular waves

\[ G_2(\omega_1, -\omega_2) = G_2(\omega_1, -\omega_2)H_L(\omega_1 - \omega_2). \]

For $G_2^f(\omega_1, -\omega_2)$ Newman\(^{20}\) has suggested the following approximation:

\[ G_2^f(\omega_1, -\omega_2) \approx G_2(\omega_1, -\omega_2) \quad (4.3.2) \]

This means that $G_2^f(\omega_1, -\omega_2)$ can be replaced by the diagonal components of a matrix of quadratic transfer function of low frequency second order force.

Since $H_L$ consists of the mass coefficient, the damping coefficient and the restoring force coefficient at low frequency motion, the additional components of hydrodynamic forces caused by encounter waves may be contained in $H_L$.

If the relation (3.1.6) is satisfied, $H_L$ can be represented in the following form:

\[ \bar{H}_L(\omega) = g^*(\omega)/g(0) = H_LK, \quad (4.3.3) \]

where $K$ is the linear restoring force coefficient and $g$ is the response function obtained from standard cross spectral analysis between surge motions and squared surface elevations. $\bar{H}_L$ is the non-dimensional form of $H_L$.

The properties of $g$ obtained from experimental data in irregular waves are shown in Fig. 12. In this figure the marks indicate the results obtained from $G_2$ by use of (3.1.6). From Fig. 12 it is found that both results agree well except for
the vicinity of peaks. Thus it is considered that the relation (3.1.6) is approximately satisfied in this case. $\tilde{H}_L$ estimated from under Newman's approximation and $\tilde{H}_L$ obtained from a free oscillation test in still water are compared in Fig. 13. From this figure it is found that both results have a same tendency in terms of the characteristics of frequency, but that the natural frequency of $\tilde{H}_L$ is shifted towards low frequency side in comparison with that of $\tilde{H}_L$ and the damping coefficient of $\tilde{H}_L$ is larger than that of $\tilde{H}_L$, and that in particular these phenomena are remarkable when the peak frequency of wave spectrum becomes high. However since $\tilde{H}_L$ overestimates $\tilde{H}_L$, it is concluded that the quadratic transfer function of surge motions, $G_2$, can be predicted from $g$ by taking into account of the effect of wave heights for the steady drift displacement and by using the frequency response function of surge motions obtained from a free

(350)
Fig. 14 Comparisons in time domain between simulated low frequency motions and measured results of surge motion.
oscillation test in still water.

We shall check this fact in time domain. The result simulated by using the above results and the measured result are compared in Fig. 14. From this figure it is confirmed that the above approximation is applicable in this case.

Finally, the following results are derived within the range of this experiment:
1) The low frequency motion is dominant in the total surge motion;
2) \(G_{\tau}(\omega, -\omega)\) which represents the steady drift displacement is not proportional to the square of wave heights;
3) The response function \(g\) of surge motions to the instantaneous wave energy, which is introduced newly, is approximately equal to the quadratic transfer function of surge motions;
4) The frequency response function of surge motions to external forces, \(H_{\omega}\), is approximated by the response function obtained from a free oscillation test in still water;
5) The linear transfer function of surge motions, \(G_{\tau}\), can be estimated from the usual linear motion prediction method and does not depend on the low frequency motion.

Thus the following relations can be obtained by applying these results to Eqs. (2.4.7) and (2.4.8).

\[
\begin{align*}
\tilde{\sigma}_{\tau} & \approx 2(\lambda \frac{\partial}{\partial t} + \frac{\partial}{\partial \omega}) = \int_{0}^{\infty} \int_{0}^{\infty} G_{\tau}(\omega, -\omega) U_{\tau}(\omega) |H_{\omega}(\omega - \nu)|^2 U_{\nu}(\nu) \, d\nu d\omega \\
\tilde{\mu} & \approx 8(\lambda \frac{\partial}{\partial t} - \frac{\partial}{\partial \omega}) = 3 \int_{0}^{\infty} \int_{0}^{\infty} \{ G_{\tau}(\omega) G_{\tau}(\nu) G_{\tau}(\omega, -\omega) H_{\omega}(\omega - \nu) \\
& + G_{\tau}(\omega) \frac{\partial}{\partial \omega} G_{\tau}(\omega, -\omega) H_{\omega}(\omega - \nu) \} U_{\tau}(\omega) U_{\nu}(\nu) \, d\omega d\nu \\
& + \int_{0}^{\infty} \int_{0}^{\infty} U_{\tau}(\omega_1) U_{\tau}(\omega_2) U_{\tau}(\omega_3) G_{\tau}(\omega_1, -\omega_1) G_{\tau}(\omega_2, -\omega_2) G_{\tau}(\omega_3, -\omega_3) \\
& \times U_{\tau}(\omega_1) U_{\tau}(\omega_2) U_{\tau}(\omega_3) H_{\omega}(\omega_1 - \omega_2) H_{\omega}(\omega_2 - \omega_3) H_{\omega}(\omega_3 - \omega_1) \\
& + C. C. \} \, d\omega_1 d\omega_2 d\omega_3
\end{align*}
\]

(4.3.4)

(4.3.5)

Where C. C. denotes the complex conjugate of the previous term.

4.4 Investigation to the Instantaneous Probability Density Function of Response

Comparisons between the sample statistical values obtained from time average over total duration time and the estimated ones from Eqs. (4.3.4) and (4.3.5) are shown in Table2. Fig. 15 shows sample statistical values as a time function. From Table2 and Fig. 15 it is found that sample variances do not depend on the duration time and show constant values, while the sample skewnesses are distributed around the estimated values and depend on the duration one. But the longer the duration time, the smaller the variation of the
sample skewness around the estimated value. Thus it may be considered that
the statistical values of surge motions including the low frequency motion can
be predicted up to third order by using the present approximate theory.

The instantaneous probability distributions of surge motions are indicated in
Fig. 16. In this figure the broken line is the estimated curve from the three term
Gram-Charlier expansion and the solid line is that from the asymptotic solution
of exact probability density function. From Fig. 16 it is found that the probabil-
ity distribution is asymmetry with respect to \(x = \bar{x}\) and has the tendency
broadening towards the direction drifted by waves. Further it is seen that the
degree of the breadth depends on the skewness of the distribution considerably.
The results of the three term Gram-Chalier expansion are found to better fit the
experimental results, while the results of the asymptotic solution do not well
approximate the probability distribution. However, the latter results represent
the behaviour of the probability distribution well at which \(x\) is significantly
large. Thus it may be considered that the probability density function of surge

\[
\begin{array}{|c|c|c|c|c|}
\hline
\text{Spectrum peak frequency (Hz)} & \text{sample statistical values} & \text{estimated statistical values} \\
\text{variance (cm}^2\text{)} & \text{skewness} & \text{variance (cm}^2\text{)} & \text{skewness} \\
(\text{cm}^2) & \bar{x} & \bar{x} & \bar{x} & \bar{x} \\
\hline
0.4 & 8.98 & -0.703 & 8.63 & -0.7498 \\
0.5 & 11.80 & -0.839 & 10.92 & -0.9911 \\
0.6 & 30.84 & -0.7787 & 31.06 & -0.8577 \\
0.7 & 64.50 & -0.1976 & 63.23 & -0.2036 \\
\hline
\end{array}
\]

\textbf{Table. 2} Comparisons with statistical values

\begin{itemize}
\item \textbf{Sample variances and skewnesses of surge motion as a time function}
\end{itemize}
Fig. 16 Comparisons between observed histograms and estimated instantaneous probability distributions of surge motion

Fig. 17 Observed histograms of minima of surge motion
motions is given by matching between the both approximate solutions.

4.5 Investigation to Probability Distributions of Extremal Values and Extreme Value

If the co-ordinate system is taken as shown in Fig. 2, minima in extremal values are more important than maxima for the mooring design purpose. The probability distributions of minima are shown in Fig. 17 and that of the negative minima which are the most important in the estimation of the maximum mooring force are shown in Fig. 18. In Fig. 18 the solid line is the curve estimated from the asymptotic solution of exact instantaneous probability density function, and the broken line is that from the three term Gram-Charlier

![Graphs showing probability distributions of negative minima of surge motion](image_url)

Fig. 18 Comparisons between observed histograms and estimated probability distributions of negative minima of surge motion
expansion and the dash-dot line is that from the Rayleigh distribution function. All curves in this figure are evaluated by using the estimated statistical values obtained from Table. 2. From these figures it is found that the probability distributions of minima depend remarkably on the skewness of the instantaneous probability distribution and the breadth of the distribution becomes wide with the increase of the absolute value of skewness and that the probability distributions of negative minima are fairly well represented by the curves estimated from the asymptotic solution of the exact instantaneous distribution. Fig. 19 shows the extreme values based on \( N_p \) observations of negative minima. The solid line shows the results obtained from the asymptotic solution, and the broken line shows those given the three term Gram–Charlier expansion, and the dash-dot line is the calculated value obtained by Longuet–Higgins\(^{23} \). The black circles indicate the experimental results obtained from each samples in total measuring data. The extreme values are normalized by the standard deviation of surge response, \( \sigma_x \). From these figures it is found that the results from Longuet–Higgins’s method significantly underestimate the extreme values of negative minima whereas those estimated from the asymptotic solution of the

\[ f_s = 0.4 \text{ Hz} \]

\[ f_s = 0.6 \text{ Hz} \]

\[ f_s = 0.5 \text{ Hz} \]

\[ f_s = 0.7 \text{ Hz} \]

**Fig. 19** Comparisons between observed extreme responses and estimated ones
instantaneous probability density function show fairly good agreement with the experimental results.

Finally, Fig. 20 shows 1/n th highest expected amplitudes for the negative minima as a time function. In this figure the broken line is the result obtained from linear theory. From this figure it is found that 1/3 th highest expected amplitude is smaller than that from linear theory and in the case of 1/10 th highest amplitude the frequency exceeding the value of linear theory become high. In the estimation of these highest amplitudes Hineno's method(24) which extended Vinje's method to wide band random processes may be used.

5. CONCLUSIONS

The results of investigations on the statistical analysis of horizontal response of a moored floating structure are summarized as follows:

(1) If it is assumed that the horizontal response of a moored structure can be represented by two term Volterra series of incident wave, the instantaneous probability density function can be obtained exactly from both the eigenvalues and eigenfunctions of the Fredholm type integral equation of second kind with
the Hermite kernel function including the quadratic transfer function, that is $G_2$.

(2) If the number of eigenvalues dominating the instantaneous probability densities is finite, this density function can be approximated by the Gram–Charlier expansion. The parameter of this expansion can be estimated from both the linear and quadratic transfer functions.

(3) The consistent results to the quadratic transfer function of response are obtained through cross bispectral analysis in irregular waves and experimental data in amplitude modulation waves and regular waves. As the results it is confirmed that this function has the following characteristics:

- The amplitude parts decrease as the difference frequency of the two wave components becomes high;
- The phase parts do not depend on the sum frequency of the two wave components;
- The diagonal values show the steady drift excursions in regular waves and are not proportional to the squared wave heights.

(4) The transfer function $g$ of horizontal response to slowly instantaneous wave energy, which is introduced newly in this case, is capable to evaluate quantitatively the characteristics of the quadratic transfer function. Within the range of this experiment it is confirmed that the Newman's approximation can be applied to the quadratic transfer function of external force and that $g$ is nearly equal to the quadratic transfer function of the response.

(5) The linear part of surge response can be separated from the total response in the frequency domain and can be estimated by the usual linear motion prediction method taking into account of the viscous damping. Further it is not affected by the nonlinear part. The response function to external forces at the low frequency motion is different from that obtained from a free oscillation test in still water. As the reason it may be considered that these phenomena are attributed to "increase of damping force in waves" proposed by one of the authors\textsuperscript{25} and Wichers\textsuperscript{26}.

(6) The instantaneous probability distribution of surge responses has the asymmetrical distribution, which broadens towards the direction drifted by waves even though the restoring force is linear. The variance, and the skewness which dominates the asymmetry of the distribution can be estimated from the frequency characteristics of the response.

(7) In order to obtain the instantaneous probability distribution we propose the approximate method matching between the finite Gram–Charlier expansion and the asymptotic form derived from the exact probability density function. The estimated results due to the present method show fairly good agreement with the experimental results.
The new prediction methods for the probability distributions of extremal values and the extreme value are proposed under the assumptions that the response displacement and velocity are independent mutually and the response velocity is of Gaussian distribution with zero mean in addition to the Powell's assumptions. As a result it is confirmed that the Longuet-Higgin's method significantly underestimates the experimental result while the present method is in good agreement with the experimental one.

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The computations were carried out on the Fujitsu FACOM M-180 IIAD computer at the central computer center of Ship Research Institute.

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Appendix A

Exact Solutions of the Instantaneous Probability Density Function of Second Order Response due to the Kac–Sieget Method$^{10}$

From Eq. (2.1.5) the horizontal response of a moored floating structure can be represented as

$$x(t) = \int_{-\tau}^{\tau} d\tau_1 g_1(\tau) \xi(t-\tau) + \int_{-\tau}^{\tau} d\tau_1 \int_{-\tau}^{\tau} d\tau_2 g_2(\tau_1, \tau_2) \xi(t-\tau_1) \xi(t-\tau_2)$$

$$= x^{(1)} + x^{(2)}$$  \hspace{1cm} (A-1)

Let $\xi(t)$ be an equivalent filtered white noise process, or

$$\xi(t) = \int_{-\tau}^{\tau} d\tau h(\tau) N(t-\tau),$$  \hspace{1cm} (A-2)

where $h(\tau)$ is the weighting function and $N(t)$ is a unit white noise which satisfies

$$E\{N(t)N(t-\tau)\} = \delta(\tau),$$  \hspace{1cm} (A-3)

$\delta(\tau)$ is the Dirac delta function.

Then following Kac and Sieget$^{10}$, we expand the white noise process in a stochastic series of orthogonal functions as

$$N(t-\tau) = \sum_{i=1}^{\infty} X_i(t) \phi_i(\tau)$$  \hspace{1cm} (A-4)

with the normalization

$$\int_{-\infty}^{\infty} dt \phi_i(t) \phi_j(t) = \delta_{ij},$$  \hspace{1cm} (A-5)

where $X_j(t)$ are the standard Gaussian variables with zero mean and unit variance and they are mutually independent. Then, in terms of the series expansion, the first term in (A-1) becomes

$$x^{(1)}(t) = \sum_{i=1}^{\infty} c_i X_i(t),$$  \hspace{1cm} (A-6)

with

(361)
\[ c_i = \int_\sigma \int_\tau g_i(\tau) h(\sigma) \phi_i(\tau + \sigma) d\tau d\sigma, \quad (A-7) \]

and

\[ x^{(2)}(t) = \sum_{i,j} X_i(t) X_j(t) \int_\alpha \int_\beta \phi_i(\alpha) \phi_j(\beta) S(\alpha, \beta) d\alpha d\beta, \quad (A-8) \]

with

\[ S(\alpha, \beta) = \int_{\tau_1} \int_{\tau_2} h(\alpha - \tau_1) h(\beta - \tau_2) g_2(\tau_1, \tau_2) d\tau_1 d\tau_2. \quad (A-9) \]

If \( \phi_i \) are chosen as the orthogonal functions which satisfies

\[ \int_\beta S(\alpha, \beta) \phi_j(\beta) d\beta = \lambda_j \phi_j(\alpha), \quad (A-10) \]

Eq. (A-8) becomes

\[ x^{(2)} = \sum_{i=1}^\infty \lambda_i X_i^2. \quad (A-11) \]

Thus, Eq. (A-1) is given in the following form:

\[ x(t) = \sum_{i=1}^\infty c_i X_i(t) + \sum_{i=1}^\infty \lambda_i X_i^2(t) \quad (A-12) \]

with

\[ E \{ X_i(t) X_j(t) \} = \delta_{ij} \quad (A-13) \]

The instantaneous probability density function \( p_x \) of \( x(t) \) can be obtained from the inverse Fourier transform of its characteristics function.

The characteristics function is defined by

\[ \phi(\theta) = E \{ \exp(i\theta x) \} = \prod_{j=1}^\infty E \{ \exp(i\theta (c_j X_j + \lambda_j X_j^2)) \}. \quad (A-14) \]

Since \( X_j \) have the probability density function as

\[ p_{x_j}(x) = 1/\sqrt{2\pi} \exp(-x^2/2), \quad (A-15) \]

by using the following identity:

\[ \int_{-\infty}^\infty dx \exp(itx - ax^2/2) = \sqrt{2\pi/a} \exp(-t^2/2a) \text{ for } a > 0 \quad (A-16) \]

the characteristics function can be rewritten as

\[ \phi(\theta) = \prod_{j=1}^\infty (1 - 2i\lambda_j \theta)^{-1/2} \exp\{-c_j^2 \theta^2/2(1 - 2i\lambda_j \theta)\}. \quad (A-17) \]

By the inverse Fourier transform of the characteristics function the instantaneous probability density function of \( x(t) \) becomes

\[ p_x(x) = 1/2\pi \int_{-\infty}^\infty d\theta \phi(\theta) \exp(-i\theta x). \quad (A-18) \]

Next we shall consider the integral equation (A-10).

It can be simplified by defining

(362)
\[ \psi_n(t) = \int_u^t h(u-t) \phi_n(u) \, du. \quad (A-19) \]

Then Eq. (A-10) can be rewritten as
\[ \int_u^t du H(t, u) \psi_n(u) = \lambda_n \psi_n(t) \quad (A-20) \]

with
\[ H(t, u) = \int_\tau d\tau g_2(\tau, u) R_\xi(t-\tau), \quad (A-21) \]

where \( R_\xi \) is the auto correlation and is given by
\[ R_\xi(\tau) = \int_{\tau_1} d\tau_1 h(\tau_1) h(\tau + \tau_1). \quad (A-22) \]

Automatically, the normalization relation for the eigenfunctions is obtained as
\[ \int_{\tau_1} d\tau_1 \int_{\tau_2} d\tau_2 \psi_m(\tau_1) \psi_n(\tau_2) g_2(\tau_1, \tau_2) = \lambda_n \delta_{mn}, \quad (A-23) \]

and the parameter \( c_n \) in (A-7) becomes
\[ c_n = \int_\tau d\tau g_1(\tau) \psi_n(\tau). \quad (A-24) \]

Finally we represent the integral equation (A-23) in the frequency domain. If the Fourier transform of \( \psi_n(t) \) exists and is defined by
\[ \Psi_n(\omega) = 1/\sqrt{2\pi} \int_{-\infty}^{\infty} dt \exp(-i\omega t) \psi_n(t), \quad (A-25) \]

then the integral equation in the time domain, (A-20), can be represented in the following form:
\[ \int_\nu d\nu \Psi_n(\nu) S_\xi(\omega, -\nu) G_\lambda(\omega) = \lambda_n \Psi_n(\omega). \quad (A-26) \]

The normalization condition (A-24) becomes
\[ \int_{\omega_1} d\omega_1 \int_{\omega_2} d\omega_2 \Psi_n(\omega_1) \Psi_m(\omega_2) G_\lambda(-\omega_1, -\omega_2) = \lambda_n \delta_{mn}. \quad (A-27) \]

Since Eq. (A-26) is a Fredholm integral equation of the second kind with a Hermite–symmetric kernel, \( \lambda \) will be real and the eigenfunction satisfies
\[ \Psi_n(-\omega) = \Psi_n^*(\omega), \quad (A-28) \]

where * denotes the complex conjugate.

**Appendix B**

Statistical Values of \( x(t) \)

From Appendix A the horizontal motion of a moored floating structure in irregular waves can be represented as

(363)
\[ x = \sum_{i=1}^{\infty} \left( c_i + \lambda_i X_i \right) X_i, \quad (B-1) \]

or

\[ x = \int_{\tau} d\tau g_{\epsilon}(\tau) \xi(t-\tau) + \int_{\tau_1} d\tau_1 \int_{\tau_2} d\tau g_{\epsilon}(\tau_1, \tau_2) \xi(t-\tau_1) \xi(t-\tau_2). \quad (B-2) \]

From (B-1) the expected values up to third order are given as follows:

\[ E\{x\} = \sum_i \lambda_i E\{X_i^2\} + \sum_i c_i E\{X_i\}, \quad (B-3) \]

\[ E\{x^2\} = \sum_{i,j} c_i c_j E\{X_i X_j\} + \sum_{i,j} \lambda_i \lambda_j E\{X_i X_j^2\}, \quad (B-4) \]

\[ E\{x^3\} = \sum_{i,j,k} c_i c_j c_k E\{X_i X_j X_k\} + \sum_{i,j,k} \lambda_i \lambda_j \lambda_k E\{X_i X_j X_k^2\} \]

\[ + \sum_{i,j,k} \lambda_i \lambda_j c_k E\{X_i X_j X_k\} + \sum_{i,j,k} \lambda_i \lambda_j \lambda_k E\{X_i X_j^2 X_k\} \]

\[ + \sum_{i,j,k} c_i c_j \lambda_k E\{X_i X_j X_k\} + \sum_{i,j,k} \lambda_i \lambda_j \lambda_k E\{X_i X_j X_k^2\} \]

\[ + \sum_{i,j,k} \lambda_i \lambda_j \lambda_k E\{X_i X_j^2 X_k\}. \quad (B-5) \]

Since \( X_i (i=1, \ldots, \infty) \) are the standard Gaussian variables with mutual independence, the following relations are satisfied:

\[ E\{X_i\} = 0, \quad (B-6) \]

\[ E\{X_i X_j\} = \delta_{ij}, \quad (B-7) \]

\[ E\{X_i X_j X_k\} = 0, \quad (B-8) \]

\[ E\{X_i X_j X_k X_l\} = \delta_{ij} \delta_{kl} + \delta_{ik} \delta_{jl} + \delta_{il} \delta_{jk}, \quad (B-9) \]

\[ E\{X_i X_j X_k X_l X_m\} = 0, \quad (B-10) \]

\[ E\{X_i X_j X_k X_l X_m X_n\} = \delta_{ij} \delta_{kl} \delta_{mn} + \delta_{ij} \delta_{km} \delta_{ln} + \delta_{ij} \delta_{kn} \delta_{lm} + \delta_{ik} \delta_{jl} \delta_{mn} \]

\[ + \delta_{ik} \delta_{jm} \delta_{ln} + \delta_{ik} \delta_{jn} \delta_{lm} + \delta_{ij} \delta_{km} \delta_{ln} + \delta_{ij} \delta_{km} \delta_{ln} \]

\[ + \delta_{il} \delta_{jkm} \delta_{ln} + \delta_{il} \delta_{jkm} \delta_{mn} + \delta_{il} \delta_{jkm} \delta_{mn} \]

\[ + \delta_{il} \delta_{jkm} \delta_{mn} + \delta_{il} \delta_{jkm} \delta_{mn}, \quad (B-11) \]

where \( \delta_{ij} \) is the Kronecker delta.

Using the above relations, the mean value \( \bar{x} \), the variance \( \sigma_x^2 \) and the skewness \( \mu_x \) of \( x(t) \) are obtained as

\[ \bar{x} = E\{x\} = \sum_i \lambda_i, \quad (B-12) \]

\[ \sigma_x^2 = E\{x^2\} - \bar{x}^2 = \sum_i c_i^2 + 2 \sum_i \lambda_i, \quad (B-13) \]

\[ \mu_x \sigma_x^2 = E\{x^3\} - 3E\{x^2\} \cdot \bar{x} + 2\bar{x}^3 = 8 \sum_i \lambda_i^3 + 6 \sum_i c_i \lambda_i. \quad (B-14) \]

(364)
From (B–2) the expected values are written as follows:

\[ E \{ x \} = \int_{\tau_1} \int_{\tau_2} \int_{\tau_2} d\tau_1 d\tau_2 g_3(\tau_1, \tau_2) R_\xi(\tau_2 - \tau_1), \quad (B–15) \]

\[
E \{ x^3 \} = \int_{\tau_1} \int_{\tau_2} \int_{\tau_3} \int_{\tau_4} \int_{\tau_5} d\tau_1 d\tau_2 d\tau_3 d\tau_4 d\tau_5 g_8(\tau_1, \tau_2, \tau_3, \tau_4) R_\xi(\tau_2 - \tau_1) R_\xi(\tau_3 - \tau_2) + R_\xi(\tau_2 - \tau_3) R_\xi(\tau_1 - \tau_4)
\times R_\xi(\tau_5 - \tau_4) R_\xi(\tau_3 - \tau_1) \}
\]

\[
E \{ x^4 \} = \int_{\tau_1} \int_{\tau_2} \int_{\tau_3} \int_{\tau_4} \int_{\tau_5} \int_{\tau_6} d\tau_1 \cdots d\tau_6 g_8(\tau_1, \tau_2, \tau_3, \tau_4, \tau_5, \tau_6) 6R_\xi(\tau_2 - \tau_1) R_\xi(\tau_3 - \tau_2)
+ 3R_\xi(\tau_2 - \tau_1) R_\xi(\tau_3 - \tau_2)
\times \left\{ \int_{\tau_1} \int_{\tau_2} \int_{\tau_3} \int_{\tau_4} \int_{\tau_5} d\tau_1 d\tau_2 d\tau_3 d\tau_4 \int_{\tau_3} \int_{\tau_4} d\tau_5 g_8(\tau_1, \tau_2, \tau_3, \tau_4, \tau_5) \right\}^3
\]

\[
E \{ x^5 \} = \int_{\tau_1} \int_{\tau_2} \int_{\tau_3} \int_{\tau_4} \int_{\tau_5} \int_{\tau_6} d\tau_1 \cdots d\tau_6 g_8(\tau_1, \tau_2, \tau_3, \tau_4, \tau_5, \tau_6) 6R_\xi(\tau_2 - \tau_1) R_\xi(\tau_3 - \tau_2)
\times R_\xi(\tau_4 - \tau_3) + 8R_\xi(\tau_2 - \tau_1) R_\xi(\tau_3 - \tau_2) R_\xi(\tau_5 - \tau_4)
\]

Accordingly we can obtain the statistical values of \( x(t) \) as

\[
x = \int_\omega d\omega G_0(\omega, -\omega) S_\xi(\omega), \quad (B–18)\]

\[
\sigma_{xx} = \int_\omega d\omega |G_1(\omega)|^2 S_\xi(\omega) + 2 \int_\omega d\omega_1 \int_\omega d\omega_2 |G_2(\omega_1, \omega_2)|^2 \times S_\xi(\omega_1) S_\xi(\omega_2), \quad (B–19)\]

\[
\mu \sigma_{xx} = 6 \int_\omega d\omega_1 \int_\omega d\omega_2 G_1(-\omega_1) G_1(-\omega_2) G_2(\omega_1, \omega_2) S_\xi(\omega_1) S_\xi(\omega_2)
\times \left\{ \int_\omega d\omega_1 \int_\omega d\omega_2 \int_\omega d\omega_3 \int_\omega d\omega_4 G_2(\omega_1, \omega_2, \omega_3, \omega_4) \right\}^2
\times S_\xi(\omega_1) S_\xi(\omega_2) S_\xi(\omega_3) \quad (B–20)\]

where \( * \) denotes the complex conjugate.

**Appendix C**

Integral Evaluation to Instantaneous Probability Density Function of \( x(t) \) as \( x \rightarrow -\infty \)

From Eq. (2.4.2) the instantaneous probability density function of \( x(t) \) may be rewritten as

\[
p_x(x) = \lim_{R \to \infty} 1/2\pi \int_{-R}^R \exp(-i\theta x) \phi(\theta) d\theta. \quad (C–1)\]

First, we expand Eq. (C–1) to complex space. By regarding this integral as a complex integral and replacing \( -i\theta \) by \( s \) we obtain the following form:

\[
p_x(x) = \lim_{R \to \infty} 1/2\pi i \int_{-R}^R \prod_{j=1}^{n} (1 + 2\lambda_j s)^{-1/2} \exp \{ sx + c_j s^2/2(1 + 2\lambda_j s) \} ds. \quad (C–2)\]

Then without loss of generality it can be assumed that \( \lambda_1 > \lambda_2 > \ldots > \lambda_n > 0 \) and \( 0 > \lambda_{n+1} > \ldots > \lambda_m \).
For $x \to -\infty$ only the positive eigenvalues are of interest. Thus in the case in which $n$ is even we take the branch cuts from $-1/2\lambda_{2j-1}$ to $-1/2\lambda_{2j}$ ($j=1, \ldots, n/2$) along the real axis as shown in Fig. (C-1). If $n$ is odd the branch-cut from $-1/2\lambda_j$ to $-\infty$ is added.

![Fig. C-1 Contours of integration](image)

Since $1+2\lambda_js$ takes either positive value or negative value when $s$ moves on the negative real axis, we define the branch of $1/\sqrt{1+2\lambda_js}$ as

$$|1+2\lambda_js|^{-1/2} \text{ for } 1+2\lambda_js > 0,$$

$$-i|1+2\lambda_js|^{-1/2} \text{ for } 1+2\lambda_js < 0.$$  \hspace{1cm} \text{(C-3)}

Then, the integrand of (C-1) is regular except for branch points $-1/2\lambda_j (j=1, \ldots, n)$.

Thus, by Cauchy's theorem the integral along $ABC \sum_j \Gamma_j E$ becomes zero. The integrals along each path are as follows:

1) Integral along BC

Setting $s = Re^{\theta} (\pi/2 \leq \theta \leq \pi)$ we have the following inequalities:

$$|\exp(c_j^2 s^2/(2(1+2\lambda_js))| \leq \exp(c_j^2 R \cos \theta/4\lambda_j) \leq 1.$$  \hspace{1cm} (366)
\[ |\exp (sx)| = |\exp (xR \cos \theta)| \leq 1, \]
\[ |1 + 2\lambda_j s|^{-1/2} \leq |1 - 2\lambda_j R|^{-1/2}. \]

Thus the integral along BC becomes zero as \( R \to \infty. \)

(2) Integral along EA

It can be proved that the integral along EA also becomes zero as \( R \to \infty. \)

(3) Integral along the semi circle of \( \Gamma \)

Setting \( s = \rho e^{i \theta} - 1/2\lambda_j (\pi \leq \theta \leq 0) \) we get the following inequalities:

\[ |\exp (c_j s^2/2 (1 + 2\lambda_j s))| \leq \exp \left\{ c_j^2 s^2/2 \lambda_j \rho \right\}, \]
\[ |\exp (sx)| \leq \exp (\rho s \cos \theta) \]
\[ |1 + 2\lambda_j s|^{-1/2} \leq 1/\sqrt{2\lambda_j \rho}. \]

Thus, if \( \rho \) is taken such that it is equal to \( c_j \sqrt{2\lambda_j s} \), which is a stationary point, this integral becomes zero as \( x \to -\infty. \)

Since the integrand of (C-1) is bounded at all points except for branch points, we obtain the following form as the evaluation of Eq. (C-1):

\[ p_s(x) = 1/\pi \sum_j (-1)^{j-1} \int_{-1/2\lambda_j}^{1/2\lambda_j} e^{sx} \prod_j |1 + 2\lambda_j s|^{-1/2} \]
\[ \times \exp \left\{ c_j s^2/2 (1 + 2\lambda_j s) \right\} ds \]
\[ \begin{cases} \text{if } n \text{ is even,} & (C-5) \\
\end{cases} \]

\[ \begin{cases} p_s(x) = 1/\pi \sum_j (-1)^{j-1} \int_{-1/2\lambda_j}^{1/2\lambda_j} e^{sx} \prod_j |1 + 2\lambda_j s|^{-1/2} \]
\[ \times \exp \left\{ c_j s^2/2 (1 + 2\lambda_j s) \right\} ds + 1/\pi (-1)^n \int_{-\infty}^{-1/2\lambda_n} e^{sx} \]
\[ \times \prod_j |1 + 2\lambda_j s|^{-1/2} \exp \left\{ c_j s^2/2 (1 + 2\lambda_j s) \right\} ds \]
\[ \begin{cases} \text{if } n \text{ is odd} & (C-6) \end{cases} \]

If it is assumed that \( \lambda_j/\lambda_1 (j=2, \ldots, n) < 1 \), the main contribution to the integral (C-1) will be from the vicinity of the branch point \( \lambda_1 \).

Similarly the integral as \( x \to -\infty \) is also evaluated by taking the branch-cuts on positive real axis.

In the case of \( n=2 \) and \( \lambda_1 \lambda_2 < 0 \) and \( \lambda_1 > c_j (j=1, 2) \) the integral (C-1) is evaluated exactly as

\[ p_s(x) = 1/2\pi \int_{-\infty}^{\infty} d\theta \exp(-ix\theta)/\sqrt{1 - 2i\lambda_1 \theta} (1 - 2i\lambda_2 \theta) \]
\[ = 1/2\pi \sqrt{\lambda_1 \lambda_2} \exp \left\{ (\lambda_1 - |\lambda_2|)/4\lambda_j |\lambda_j| \right\} K_0 \left\{|x|(|\lambda_1 - |\lambda_2|)/4\lambda_j |\lambda_j|\right\}. \]

(C-7)

where \( K_0 \) is the modified Bessel function of the second kind.