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On the Joint Distribution of Excursion Duration and Amplitude of a Narrow-Band Gaussian Process

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ABSTRACT The probability density of crest amplitude and duration that exceeds a given level is used in many theoretical and practical problems in engineering that are subjected to fluctuating loads such as wind and wave loads. The presently available joint distributions of amplitude and period are limited to excursion through a mean-level or to describe the asymptotic behavior of high level excursions. This paper extends the knowledge by presenting a theoretical derivation of probability of wave exceedance amplitude and duration for a stationary narrow-band Gaussian process. A density function is suggested that has the salient feature to depend only on the three lowest spectral moments $m_0, m_1, and m_2$ and desired level of exceedance, H. It does not require any condition on the autocorrelation function. This paper shows how increase in H, increases the correlation between excursion periods and amplitude. This paper also shows that how the accuracy of the proposed joint distribution relates to spectral width parameter, ν , and that accuracy increases for higher levels of H, especially for a spectrum describing a physical phenomenon such as a sea state spectrum. It was demonstrated that the marginal distribution of amplitude is Rayleigh distributed, as expected, and that the marginal distribution of excursion duration works for asymptotic and non-asymptotic levels. Results demonstrate that the established distribution fits well with ideal narrow-band Gaussian processes as well as the sea states at three European sites —in the Atlantic Ocean and the North Sea. The suggested model is found to be a good replacement for the existing empirical distributions.

INDEX TERMS Stationary process, Gaussian narrow-band process, joint distribution, excursion duration, crest amplitude, level crossing.

I. INTRODUCTION

The probability distribution function (PDF) of amplitude and duration of exceeding a certain level of a stochastic process, $\zeta(t)$, is essential for many theoretical and practical problems. These range from mechanical stress analysis, change detection problems, fading phenomena in wireless communication and many applications in marine, offshore and coastal engineering. In offshore and coastal engineering, it includes both short-term and long-term statistics. Shortterm applicability includes slamming loads, deck-impact events, short-term probability distribution of impact duration, bottom-slam forces, over-topping waters for wave breakers and wave energy convertors [1]. Long-term application includes operation, installation and survival of offshore renewable energy and platforms in general [2], [3].

Investigating of the distribution of amplitude (maxima) and period dates back to the the pioneer works of Rice [4], [5]. He scrutinized the statistical properties of a random function —noise in an electrical circuit—and established a closedform expression, i.e., the Rice formula, of the average crossing rate under favorable conditions. By using the average crossing rate, he derived the distribution of maxima for a Gaussian process. His fundamental results were summarized in [2] and [6]. Wave period can be defined as the time between two successive local maxima or successive upcrossing through the mean level. Opting the latter definition, various approximations of wave period density functions have been proposed, for example, Rice [4], [5, Sec. 3.4] and Longuet-Higgins [7]–[9].

The probability density of excursion (exceedance) duration, time between an level-up-crossing followed by a level-down-crossing, were also derived by Rice [5]. The proposed approach is, however, rather complicated for nonzero levels and can only be estimated by means of demanding numerical integration. Non-diagonal covariance matrix between stochastic function and its derivative hinders a simple analytical derivation of the density of excursion duration. Many researchers, therefore, tried to address this issue by approximating the integration. By partitioning an n-dimensional normal space to subspaces, Wu-Zhou and Ming-Shun [10] proposed a semi-analytic method to reduce the computational efforts. Empirical parametric models for the duration statistics of the significant wave height were established by Graham [11], and Kuwashima and Hogben [12].

Analytical expressions for the mean value and variance of excursion duration can be found in [13] and [14], where Mathiesen [14] used the mean value expression to validate a empirical parametric model. A number of researchers have treated this problem by assuming an asymptotically large level of excursion [15]–[21]. Using orthogonal series, Gram-Charlier, Edgeworth or Laguerre series, is another approach to approximate the PDF of exceedance and non-exceedance duration [3].

The joint distribution of wave period and amplitude is a more challenging problem in the most general case. However, for a stochastic process with a narrow-band spectrum some simplifications can be made; Wooding [22] was the first to derive the joint distribution of wave amplitude and period for a narrow-band spectrum using the density of the wave envelope and its time derivative. Cavanie et al. [23] proposed a closed-form expression for the joint distribution assuming a narrow-band Gaussian random process based on spectral moments up to 4th order (m_0, m_1, m_2, m_4) . By extracting the mean frequency of the spectrum, Longuet-Higgins [24], [25] ended up with a diagonal covariance matrix between slowvarying envelope, high-frequency component and their time derivatives, and derived a simpler closed-form expression for the joint distribution of wave period and amplitude based on spectral moments (m_0, m_1, m_2) . Using the joint distribution established by Longuet-Higgins, EM. Antao and CG. Soares proposed the joint distribution of wave steepness in narrowband sea states [26]. However, these closed-form formulas cannot provide a description of the joint density of amplitude and excursion duration nor density of excursion duration for a level that differs from the mean level.

To the best knowledge of the authors, there is no closedform expression for the joint distribution of excursion period and amplitude yet. Hence, the objective of this paper is to fill this gap and establish a closed-form expression for the joint distribution of exceedance duration and amplitude above a certain level and parameterize this with low-order spectral moments. We assume a Gaussian narrow-band process with continuous first derivative. The paper will also show how the the empirical excursion distribution model can be replaced by the marginal distribution for excursion duration that follows from the joint distribution. The paper will validate the suggested closed-form joint distribution with simulated ideal narrow-band and sea states data and investigate the accuracy of the model when the bandwidth of the underlying narrowband process changes towards a broader bandwidth stochastic process and for different levels.

The remainder of this paper is organized as follows. Section II presents a derivation of the joint distribution of excursion duration and amplitude and determines the marginal distribution of amplitudes and excursion duration. Section III validates the model using the simulation results from ideal-narrowband spectra and sea states data. The Kolmogorov-Smirnov test (K-S test) is used as a goodness-offit (GoF) test to provide a quantitative measure of the accuracy of the suggested joint distribution. Finally, Section IV concludes and highlights the main findings of the paper.

II. STATEMENT OF THEORETICAL RESULTS

A. ASSUMPTIONS

A Gaussian random process ζ (*t*)—any function indexed by time representing for example the wave elevation at a specific point in space—has been commonly been represented by the Fourier series Eq. (1), [24], [25], [27], [28].

$$\zeta(t) = \sum_{n=1}^{\infty} a_n \cos(\omega_n t + \epsilon_n)$$
(1)

Where a_n and ϵ_n are statistically independent random variables. The amplitude a_n is Rayleigh distributed, and the phase ϵ_n is uniformly distributed over $[0, 2\pi)$. For a narrow-band process, ω_n is densely distributed over $(0, \infty)$. A narrow-band assumption helps to break down $\zeta(t)$ into a carrier high frequency wave with the mean frequency $\overline{\omega}$, and a slowly varying envelope function Eq. (2), [24], [25].

$$\zeta(t) = Rel\left\{\rho e^{i\phi} e^{i\bar{\omega}t}\right\},\tag{2}$$

where $\rho e^{i\phi}$ is the complex-value envelope function of time *t* and total phase ϕ Eq. (3),

$$\rho e^{i\phi} = \sum_{n=1}^{\infty} a_n \exp\left\{i\left[(\omega_n - \bar{\omega}) + \epsilon_n\right]\right\}.$$
 (3)

The mean frequency $\bar{\omega}$ is defined as:

$$\bar{\omega} = \frac{m_1}{m_0},\tag{4}$$

where m_n stands for the *nth* central moment of the spectral density $S(\omega)$ of $\zeta(t)$:

$$m_n = \int_0^\infty w^n S(\omega) d\omega.$$
 (5)

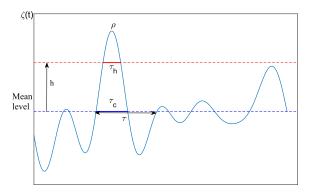


FIGURE 1. Relation between exceedance duration τ_h –for given level h–crest amplitude ρ , crest period τ_c and wave period τ .

Similarly the nth non-central moment of the spectral density $S(\omega)$ is defined as:

$$\mu_n = \int_0^\infty (w - \bar{w})^n S(\omega) d\omega, \qquad (6)$$

and spectral width parameter v is defined as

$$\nu^2 = \frac{\mu_2}{\bar{\omega}^2 m_0}.$$
 (7)

Longuet-Higgins [24], [25] derived the closed-form joint distribution for a narrow-band ($\nu^2 \le 0.36$) Gaussian process which has a continuous first derivative, Eq. (8).

$$p(A_c, T) = \frac{2}{\nu \pi^{1/2}} L(\nu) \frac{A_c^2}{T^2} \exp\left\{-A_c^2 \left[1 + (1 - 1/T)^2 \nu^{-2}\right]\right\}$$
(8)

Where A_c and T represent the normalized crest amplitude and period, respectively. The crest amplitude, ρ , is normalized by $\sqrt{2 m_0}$ and wave period, τ , time between two successive zero-up-crossing, is normalize by the mean period $\bar{\tau}$, Eq. (9), Fig. 1.

$$A_c = \frac{\rho}{(2m_0)^{\frac{1}{2}}}, \quad T = \frac{\tau}{\bar{\tau}}, \ \bar{\tau} = \frac{2\pi m_0}{m_1}.$$
 (9)

L(v) is a function of spectral width parameter v and defined as

$$L(\nu) = \frac{2}{\left[1 + \left(1 + \nu^2\right)^{-1/2}\right]}.$$
 (10)

B. DERIVATION OF THE JOINT DISTRIBUTION OF EXCURSION DURATION AND AMPLITUDE

Starting from Longuet-Higgins joint distribution of amplitude and period, Eq. (8), and assuming that zero-up-up crossing period T is twice the zero-up-down crossing period T_c , we end up with the joint distribution of the crest period and amplitude, Eq. (11).

$$p(A_c, T_c) = \frac{1}{\nu \sqrt{\pi}} L(\nu) \frac{A_c^2}{T_c^2} \\ \times exp\left\{ -A_c^2 \left[1 + (1 - 1/2T_c)^2 \nu^{-2} \right] \right\}$$
(11)

For a narrow-band process where there is only one maxima between two successive zero-up-down crossing a sine-shape function can be used to approximate the relation between the crest amplitude (greater than a given threshold), the crest period and the excursion duration, Fig. 1.

The total phase, $\Delta phase$, pertinent to excursion time is presented in Eq. (12), approximating the phase ϕ using a sine-shape function. Eq. (15) presents the relation between normalized excursion duration T_H , crest period T_c and crest amplitude A_c .

$$\begin{cases} \rho \sin(\phi) = h \\ \rho \sin(\pi - \phi) = h \end{cases}$$
$$\Rightarrow \pi - 2\phi = \Delta phase \Rightarrow \pi - 2\phi = \frac{\pi}{\tau_c}\tau_h \qquad (12)$$

$$\tau_h = \left(1 - \frac{2}{\pi} \arcsin\left(\frac{h}{\rho}\right)\right) \tau_c \xrightarrow{Normalizing}$$
(13)

$$\frac{\tau_h}{\bar{\tau}} = \left(1 - \frac{2}{\pi} \arcsin\left(\frac{\frac{h}{\sqrt{2m_0}}}{\frac{\rho}{\sqrt{2m_0}}}\right)\right) \frac{\tau_c}{\bar{\tau}} \Rightarrow$$
(14)

$$T_H = \left(1 - \frac{2}{\pi} \arcsin\left(\frac{H}{A_c}\right)\right) T_c.$$
(15)

Introducing an auxiliary variable such as y, we now convert A_c , T_c domain to y, T_H domain Eqs. (16) to (18)

$$p(A_c, T_c) \xrightarrow{\text{change of variables}} \left\{ \begin{array}{c} y = A_c := g_1(A_c, T_c) \\ T_H = \left(1 - \frac{2}{\pi} \arcsin\left(\frac{H}{A_c}\right)\right) T_c := g_2(A_c, T_c) \end{array} \right\}$$
(16)
$$\Rightarrow \left\{ \begin{array}{c} A_c = g_1^{-1}(T_H, y = A_c) \\ T_c = \frac{T_H}{(-2)^{-1}(H_c)} = g_2^{-1}(T_H, A_c) \end{array} \right\}$$
(17)

$$p(A_c, T_H) = p_{A_c, T_c} \left(g_1^{-1}, g_2^{-1} \right) \times |J|$$

$$= \frac{1}{\nu \sqrt{\pi}} L(\nu) \frac{A_c^2}{T_H^2} \left(1 - \frac{2}{\pi} \arcsin\left(\frac{H}{A_c}\right) \right)$$

$$\cdot \exp\left\{ -A_c^2 \left[1 + \left(1 - \frac{\left(1 - \frac{2}{\pi} \arcsin\left(\frac{H}{A_c}\right)\right)}{2T_H}\right)^2 \nu^{-2} \right] \right\}.$$
(18)

In order to obtain a simpler form of Eq. (18) over the desired domain, $\frac{H}{A_c} \in [0, 1]$ interval, we approximate as follows

$$\left(1 - \frac{2}{\pi} \arcsin\left(\frac{H}{A_c}\right)\right) \simeq \sqrt{1 - \frac{H}{A_c}}; \quad \text{for } \frac{H}{A_c} \in [0, 1]$$
(19)

Fig. 2 depicts this approximation. It is seen that the Taylor expansion (even up to 5th order) cannot capture the $\left(1 - \frac{2}{\pi} \arcsin\left(\frac{H}{A_c}\right)\right)$ function when $\frac{H}{A_c}$ goes toward one.

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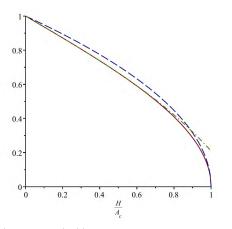


FIGURE 2. $\left(1 - \frac{2}{\pi} \arcsin\left(\frac{H}{A_c}\right)\right)$ (solid line), and its Taylor expansion up to 5th order (dash-dot line), and $\sqrt{1 - \frac{H}{A_c}}$ (dashed line).

 $\sqrt{1 - \frac{H}{A_c}}$ function, however, seems to be a better and simpler approximation for all the desired range. The ensuing simpler form is also helpful to handle the integrations in order to derive marginal distributions. We, therefore, end up with the following form for the joint distribution of crest amplitude and excursion duration for a given threshold level *H*, Eq. (20).

$$p(A_{c}, T_{H}) = K(H, \nu) \frac{1}{\nu \sqrt{\pi}} L(\nu) \frac{A_{c}^{2}}{T_{H}^{2}} \sqrt{1 - \frac{H}{A_{c}}}$$
$$\cdot \exp\left\{-A_{c}^{2} \left[1 + \left(1 - \frac{\sqrt{1 - \frac{H}{A_{c}}}}{2T_{H}}\right)^{2} \nu^{-2}\right]\right\}$$
(20)

Where L(v) is the same as Eq. (10). We also need to add a normalization factor, say $K(H, \nu)$, to guarantee that $\int \int (A_c, T_H) dA_c dT_H$ is unity. Where $K(H, \nu)$ clearly depends on H and ν values and will be addresed in Section II-C and Appendix. IV-B. The joint distribution $p(A_c, T_H)$ for v = 0.5 and for different normalized threshold H = [0, 0.5, 1, 2] are depicted in Fig. 3. Each contour represents a quantile level enclosing P% of the probability density function while P takes the values 10, 30, 50, 70, 90, 95 and 99 from the center contour outwards. The established joint density function works both for zero and non-zero level H, and for H = 0, it produce the similar result as was presented by Longuet-Higgins [25]. As H increases, we end up with a narrower distribution especially with respect to excursion duration, Figs. 3b to 3d. It should be noted that the result for narrower spectrum (smaller ν values) are similar. It is evident, as previously also shown in other literature [23]–[25] for H = 0, the smaller the ν values are, the narrower the excursion duration are for the same threshold level H.

C. THE MARGINAL DISTRIBUTION OF $p(A_c)$ FOR $A_c > H$ In order to derive the density of wave crest amplitude A_c , we need to integrate $p(A_c, T_H)$ with respect to T_H over $0 < C_c$

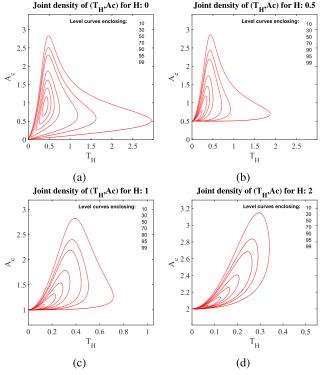


FIGURE 3. Contour lines of the joint density of normalized excursion duration (T_H) and normalized crest amplitude (A_c) for Gaussian process for v = 0.5 and for different normalized level H = [0, 0.5, 1, 2], where contours represent quantile levels which enclose P% of the PDF and P takes [10, 30, 50, 70, 90, 95, 99] values from center contour outwards.

$$\begin{aligned} T_{H} &< \infty, \\ p(A_{c}) &= \int_{0}^{\infty} p(A_{c}, T_{H}) dT_{H} \\ &= K(H, \nu) \frac{1}{\nu \sqrt{\pi}} L(\nu) \frac{A_{c}^{2}}{1} \sqrt{1 - \frac{H}{A_{c}}} \exp\left(-A_{c}^{2}\right) \\ &\cdot \int_{0}^{\infty} \frac{1}{T_{H}^{2}} \exp\left\{\left(\frac{A_{c}}{\nu}\right)^{2} \left(1 - \frac{0.5\sqrt{1 - \frac{H}{A_{c}}}}{T_{H}}\right)^{2}\right\} dT_{H}. \end{aligned}$$

$$(21)$$

The integral part can be written as

$$\int_0^\infty \frac{1}{T_H^2} \exp\left\{-a^2 \left(1 - \frac{b}{T_H}\right)^2\right\} dT_H,\qquad(22)$$

where *a* is $\frac{A_c}{\nu}$ and *b* is $0.5\sqrt{1-\frac{H}{A_c}}$. By applying $\beta = a(1-b/T_H)$, we end up with the same results as Longuet-Higgins [25], Eq. (23), where he only considered the case where *H* was zero.

$$A_{c} \exp(-A_{c}^{2}) L(\nu) \frac{2}{\sqrt{\pi}} \int_{-\infty}^{\frac{A_{c}}{\nu}} \exp\left(-\beta^{2}\right) d\beta$$
$$= A_{c} \exp(-A_{c}^{2}) L(\nu) \left[1 + \operatorname{erf}(\frac{A_{c}}{\nu})\right] \quad (23)$$

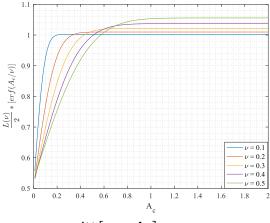


FIGURE 4. The value of $\frac{L(v)}{2} \left[1 + \text{erf}(\frac{A_C}{v}) \right]$ for different values of v = [0.1, 0.2, 0.3, 0.4, 0.5].

The only difference, as the general form —*H* is not necessarily zero—is that for $A_c < H$, $p(A_c)$ is zero; accordingly the distribution of crest amplitude $p(A_c)$ is

$$\begin{cases} K(H, \nu) 2A_c \exp(-A_c^2) \frac{L(\nu)}{2} \left[1 + \operatorname{erf}(\frac{A_c}{\nu}) \right] & \text{for } A_c > H \\ 0 & \text{for } A_c < H \end{cases}$$
(24)

Setting $A_c = r/\sqrt{2}$, $2 A_c \exp(-A_c^2)$ becomes $r \exp(-\frac{r^2}{2})$ which is, as we expected, the Rayleigh distribution. We need also to consider a normalization factor, $K(H, \nu)$, to take into account the effect of $\frac{L(v)}{2} \left| 1 + \operatorname{erf}(\frac{A_c}{v}) \right|$ coefficient. Fig. 4 evaluates $\frac{L(\nu)}{2} \left[1 + \operatorname{erf}(\frac{A_c}{\nu}) \right]$ for different values of ν and A_c . It is seen that for small ν and large A_c values, $\frac{L(\nu)}{2} \left[1 + \operatorname{erf}(\frac{A_c}{\nu}) \right]$ is close to unity. For this especial case, by assuming $\frac{L(v)}{2} \left[1 + \operatorname{erf}(\frac{A_c}{v}) \right] = 1$, Longuet-Higgins showed that normalization factor, $K(H, \nu)$, is equal to $\exp(H^2)$, [25]. Accordingly, Fig. 5 shows the density function of crest amplitude A_c which is almost Rayleigh distributed, yet it must be corrected by a factor of $\frac{L(v)}{2} \left[1 + \operatorname{erf}(\frac{A_c}{v}) \right]$. It is seen that p(Ac) goes towards a Rayleigh distribution by increasing A_c and by decreasing ν values, narrower spectrum. It should be noted that dash-line is referred to Rayleigh distribution shape —the Rayleigh distribution—since its integral is not unity. It is scaled with $\exp(H^2)$ for comparison.

The general form of the normalization factor, $K(H, \nu)$ in Eq. (24) is derived in the Appendix. IV-B, and is presented in Eq. (25).

K(H, v)

$$=\frac{1+\sqrt{\nu^2+1}}{1+\exp\left(-H^2\right)\sqrt{\nu^2+1}\left(1+\exp\left(\frac{H}{\nu}\right)\right)-\exp\left(\frac{H\sqrt{\nu^2+1}}{\nu}\right)},$$
(25)

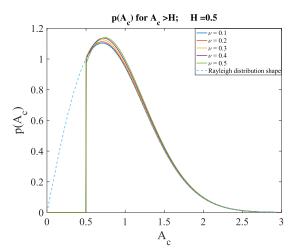


FIGURE 5. The density of crest amplitude, A_c , larger than threshold H, for v = [0.1, 0.2, 0.3, 0.4, 0.5] and Rayleigh distribution shape.

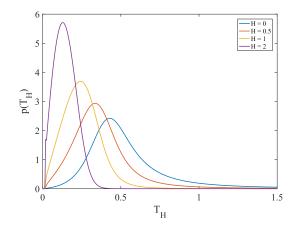


FIGURE 6. The density of the excursion duration, T_H , for v = 0.5 and for different threshold H = [0, 0.5, 1, 2].

D. THE MARGINAL DISTRIBUTION OF $p(T_H)$

The density distribution of T_H , is obtained by integrating $p(A_c, T_H)$ with respect to A_c over $H < A_c < \infty$. The desity of T_H for H = 0, which is the crest period T_c , was derived by Longuet-Higgins [25]. However, for the general case H > 0, the analytic closed-form solution is possible to obtain but too long and too complicated; hence, the numerical integration is adopted here, Fig. 6.

Fig. 6. is the numerical projection of Fig. 3 with respect to T_H dimension. It is seen that, increasing normalized threshold H results in a narrower density of T_H . Smaller ν values, similarly, result in narrower density of T_H .

As it was shown in [25] the mean value of the T_H distribution is theoretically infinite for H = 0, since for asymptotic value of T_c the density $p(T_c)$ behaves like T_c^{-2} function; therefore, an alternative estimate of the mean period was adopted for normalization. We use the same approach and extend it for the general case i.e. $H \ge 0$.

As it can be deducted from Fig. 7, for a random process like $\zeta(t)$ and given a threshold *H*, expected value of the excursion

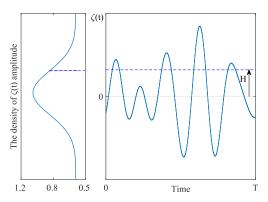


FIGURE 7. Schematic for expected value of excursion duration.

period $E[T_H]$, as an alternative approach, is presented in Eq. (26), [13], [14], [29].

$$E[T_H] = \frac{[Pr(\zeta(t) \ge H)]T}{N^+(T,H)}$$
(26)

Where $[Pr(\zeta(t) \ge H)]T$ represents all time $\zeta(t)$ spends above the threshold, and $N^+(T, H)$ represents number of positive-level *H* crossing during *T* time span, Fig. 7. For a Gaussian process, accordingly, we have

$$Pr\left(\zeta(t) \ge H\right) = \frac{1}{2} \left[1 - \operatorname{erf}\left(\frac{H}{\sqrt{2m_0}}\right) \right], \qquad (27)$$

and based on the Rice formula for a Gaussian process [4] and [5] we have

$$\frac{N^+(T,H)}{T} = \frac{1}{2\pi} \sqrt{\frac{m_2}{m_0}} \exp\left(\frac{-H^2}{2m_0}\right).$$
 (28)

Therefore,

$$E[T_H] = \left[1 - \operatorname{erf}\left(\frac{H}{\sqrt{2m_0}}\right)\right] \pi \sqrt{\frac{m_0}{m_2}} \exp\left(\frac{H^2}{2m_0}\right).$$
(29)

Replacing H = 0 in Eq. (29), we end up with half the same value as derived by Longuet-Higgins [25], the 0.5 coefficient is because he assumed a period as time between two successive zero-up crossing and we consider the crest period which is half the period. Hence, we modify the $p(T_H)$ using the following correction coefficient for the T_H

$$\frac{E\left[T_{H}\right]}{\overline{T}_{H}}\tag{30}$$

where \overline{T}_H is obtained from marginal distribution, $p(T_H)$, Eq. (20). The proposed modification helps to provide a more accurate estimation of the distribution of excursion duration.

III. COMPARISON WITH IDEAL NARROW-BAND GAUSSIAN PROCESS AND REAL SEA STATES

To provide a simulation-based validation for the proposed joint distribution of excursion duration and crest amplitude, three approaches are adopted. First, ideal narrow-band spectrum is constructed, for v = [0.1, 0.2, 0.3, 0.4, .5] - v = 0.6

is excluded for reasoning see Appendix. IV-A. Then, time series realization are generate with more than 1e+6 sample point in time domain for different levels. Obtained excursion duration and crest amplitudes from simulation are, then, compared with theoretical results established in the paper. Second, real sea states are used to compare the efficiency of the proposed formula for spectra describing physical phenomenon. Finally, Kolmogorov-Simirnov (K-S) test as a goodness of fit (GoF) test is chosen to provide a quantitative metrics to scrutinize statistical sensitivity in different regions of the proposed PDFs.

Fig. 8 depicts the stepwise simulation for a narrow-band Gaussian process. Part (a) shows an ideal spectral density for v = 0.3 and $\bar{w} = 0.6$; a small portion of the related time-domain realization is depicted in part (b). Black dots in part (c) represents the normalized crest amplitudes for $A_c > 0.5$ and related normalized excursion duration that are obtained form time-domain realization. Green contours show their Kernel density estimation (KDE) for the joint distribution of excursion duration and crest amplitude. The KDE provides a non-parametric piecewise approximation of the joint distribution to compare with the established PDF. Each contour represents a quantile level enclosing P% of the PDF while P takes the 10, 30, 50, 70, 90, 95 and 99 values from the center contour outwards. Part(d) is like part (c), while contours are obtained from the proposed analytic expression in this paper Eq. (20). It is seen that the KDE and established formula are matched pretty well. Parts (e) shows the PDF of the normalized crest amplitude (solid line), Eq. (24), while dashed-line represents the scaled Rayleigh distribution for a better comparison. Part (f) shows the cumulative distribution function (CDF) of both empirical data and analytic approach; CDF gives a better appreciation of accumulation of error between numerical simulation and proposed analytic formula. Part (g) shows the marginal distribution of the excursion duration, $p(T_H)$, accommodating the correction factor, Eq. (30). Similarly, $CDF(T_H)$ is shown in (h). Simulation data and proposed closed-form formula for the $p(A_c, T_H)$ and its marginal distributions, i.e, $p(T_H)$ and $p(A_c)$ agree well with each other, Fig. 8.

It is intended to show the trend both with respect to change of v and H values. To save spaces, only a few figures are shown; more comparison is presented later on using K-S test. Fig. 9 depicts similar results for $\nu = 0.5$; H =0.5 values to investigate the change of spectral width parameter v. For a given mean frequency, here $\bar{w} = 0.6$, by increasing ν value we should expect a wider spectrum which is also appreciated from time-domain realization, i.e., occasionally having more than one maxima between each zero-up crossing and the next zero-down crossing, which contradict our assumptions-narrow-band assumption and using a sine shape function to approximate change of phase for a given threshold level H. Accordingly, we should expect a decrease in the performance of the proposed joint distribution and density of excursion duration for v = 0.5, Fig. 9(h).

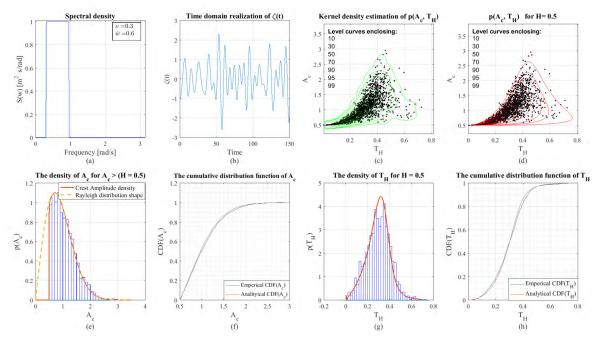


FIGURE 8. Time domain simulation of an ideal narrow-band Gaussian process for v = 0.3 and normalize threshold H = 0.5; (a) depicts the spectral density, (b) is a portion of the related time domain realization; black dots in (c) are crest amplitudes greater than H and related excursion duration and green curves are the obtained kernel density estimation of $p(A_c, T_H)$; (d) shows the proposed theoretical formula for $p(A_c, T_H)$ Eq. (20); (e:f) are the marginal distributions for crest amplitude and excursion period, respectively.

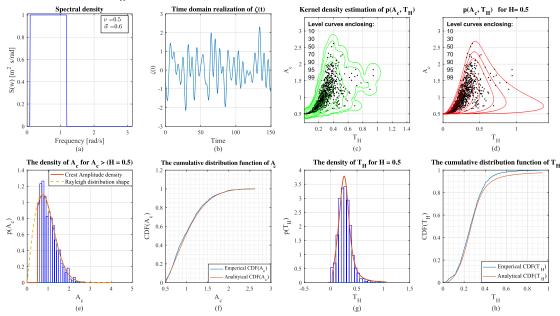


FIGURE 9. Time domain simulation of an ideal narrow-band Gaussian process for v = 0.5 and a normalized threshold H = 0.5. (a) Depicts the spectral density, (b) is a portion of the related time domain realization; black dots in (c) are crest amplitudes greater than H and related excursion duration and green curves are the obtained kernel density estimation of $p(A_c, T_H)$; (d) shows the proposed theoretical formula for $p(A_c, T_H)$ Eq. (20); (e:f) are the marginal distributions for crest amplitude and excursion period, respectively.

Figs. 10 and 11 depicts the similar simulation results for v = 0.5; H = [1, 2]. Trend for higher H values is similar to smaller v values —narrower spectrum. That is, the excursion duration decreases by increasing H values, which results in narrower distribution of both amplitude and excursion duration. This is appreciable from visual comparison between Figs. 8, 10 and 11. It is seen that the proposed joint distribu-

tion and the marginal distributions for crest amplitude and excursion duration matches well, except for v = 0.5 and H = 0.5, with the simulation data even for a quiet wide spectrum, i.e., v = 0.5.

Sea states simulations, as examples of physical phenomenon comparing to ideal narrow-band spectra, are carried out for three different European offshore sites in Atlantic

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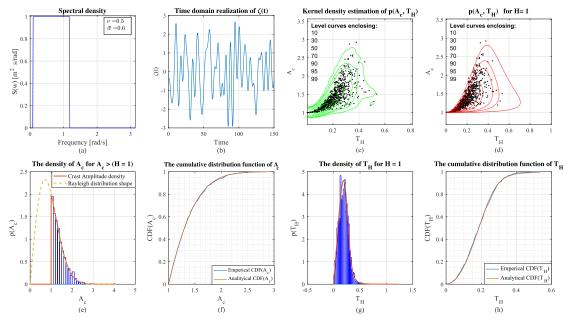


FIGURE 10. Time domain simulation of Ideal narrow-band Gaussian process for v = 0.5 and normalize threshold H = 1. (a) depicts the spectral density, (b) is a portion of the related time domain realization; black dots in (c) are crest amplitudes greater than H and related excursion duration and green curves are the obtained kernel density estimation of $p(A_c, T_H)$; (d) shows the proposed theoretical formula for $p(A_c, T_H)$ Eq. (20); (e:f) are the marginal distributions for crest amplitude and excursion period, respectively.

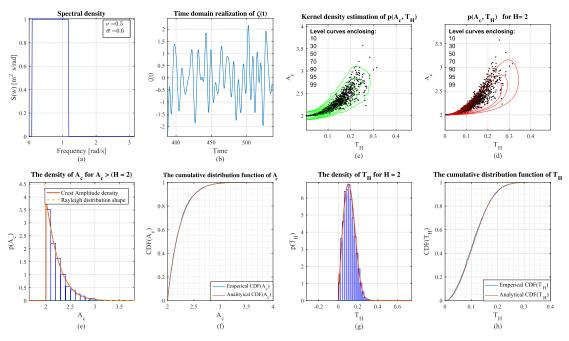


FIGURE 11. Time domain simulation of Ideal narrow-band Gaussian process for v = 0.5 and normalize threshold H = 2. (a) depicts the spectral density, (b) is a portion of the related time domain realization; black dots in (c) are crest amplitudes greater than H and related excursion duration and green curves are the obtained kernel density estimation of $p(A_c, T_H)$; (d) shows the proposed theoretical formula for $p(A_c, T_H)$ Eq. (20); (e:f) are the marginal distributions for crest amplitude and excursion period, respectively.

ocean and North sea, [30]. The related critical conditions obtained from 50-year record. General information and statistics of the sites is shown in Table 1. Due to similarity, only simulation results for site No.1 are presented in this section, the detailed results for GoF test statistics for all three sites and for different threshold level H are presented later in the paper. Time realizations are generated with wave conditions characterized by significant wave height $H_S = 8 m$ and

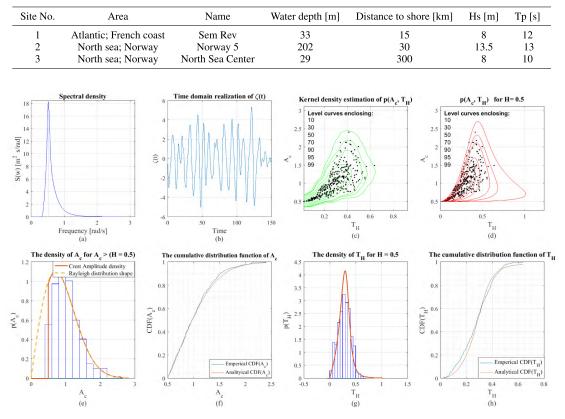


TABLE 1. General information of three european sites.

FIGURE 12. Time domain simulation of real sea data for normalized threshold H = 0.5, $H_s = 8 m$ and $T_p = 12 s$; (a) depicts the spectral density, (b) is a portion of the related time domain realization; black dots in (c) are crest amplitudes greater than H = 0.5 and related excursion duration; green curves are the obtained kernel density estimation of $p(A_c, T_H)$; (d) shows the proposed theoretical formula for $p(A_c, T_H)$ Eq. (20); (e:f) are the marginal distributions for crest amplitude and excursion duration, respectively.

peak period $T_P = 12 s$ (modelled by a JONSWAP spectrum), for more than 1e + 6 points in time domain for different threshold, Figs. 12 to 14. Poor performance of estimated excursion duration distribution for H = 0.5 —which is more evident in Fig. 12 part (h) —is because of the fact that narrowband assumption is not valid here. This is evident from time domain realization Fig. 12 part (b), i.e., having more than one maximum/minimum between each zero-up/down crossing and successive zero-down/up crossing.

A closer look at sea states spectral density, reveals that the trend for higher level H is similar to narrower spectrum (smaller ν values), since a smaller portion of the frequencies covered by the spectrum have enough energy, those are closer to peak-frequency, to cross the higher level. Consequently, by increasing level H the proposed joint distribution and its marginal distributions are closer to simulation results and have a better performance, Figs. 13 and 14. This is more evident by comparing parts (g) and (h) of Figs. 12 – 14. It is also noticeable that by increasing H, the correlation between crest amplitude and excursion period increases, Fig. 14

Simulation of real sea data for normalized threshold H = 1, $H_s = 8 m$ and $T_p = 12 s$. (a) depicts the spectral

density, (b) is a portion of the related time domain realization; black dots in (c) are crest amplitudes greater than H and related excursion duration and green curves are the obtained kernel density estimation of $p(A_c, T_H)$; (d) shows the proposed theoretical formula for $p(A_c, T_H)$ Eq. (20); (e:f) are the marginal distributions for crest amplitude and excursion period, respectively.

Besides the visual inspection of the proposed distributions and simulated data, it is critical to provide a quantitative accuracy metric using Goodness-of-fit (GoF) test. There are many GoF tests, while, empirical distribution function (EDF)-based tests such as the Kolmogorov-Smirnov (K-S) and Anderson-Darling (A-D) tests are often found to be more powerful than the Chi-square test [31]–[33]. A-D test is especially useful where it is needed to places more weight or discrimination power at the tails of the distribution which is very useful for heavy-tail distribution [34]; hence, K-S test is used here. Related test statistics for different significance level α are presented in Tables 2 and 3.

The sup|ECDF - CDF| refers to the maximum deviation between empirical cumulative distribution function (ECDF) and the proposed marginal distribution for excursion duration T_H as well as crest amplitude A_c . To pass the K-S test, this

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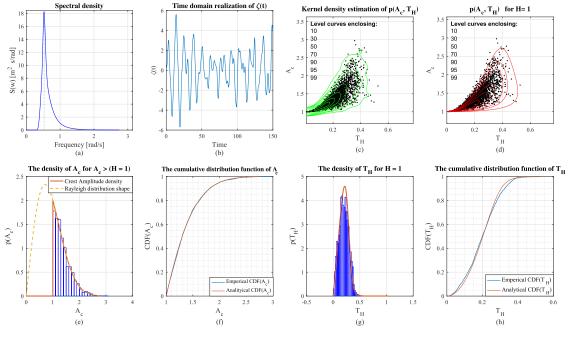


FIGURE 13. Simulation of real sea data for normalized threshold H = 1, $H_S = 8 m$ and $T_P = 12 s$. (a) depicts the spectral density, (b) is a portion of the related time domain realization; black dots in (c) are crest amplitudes greater than H and related excursion duration and green curves are the obtained kernel density estimation of $p(A_c, T_H)$; (d) shows the proposed theoretical formula for $p(A_c, T_H)$ Eq. (20); (e:f) are the marginal distributions for crest amplitude and excursion period, respectively.

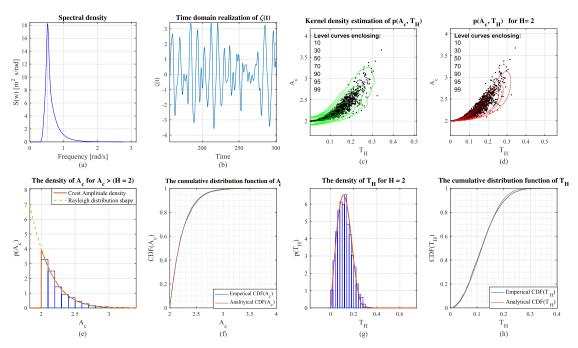


FIGURE 14. Simulation of real sea data for normalized threshold H = 2, $H_S = 8 m$ and $T_p = 12 s$. (a) depicts the spectral density, (b) is a portion of the related time domain realization; black dots in (c) are crest amplitudes greater than H and related excursion duration and green curves are the obtained kernel density estimation of $p(A_c, T_H)$; (d) shows the proposed theoretical formula for $p(A_c, T_H)$ Eq. (20); (e:f) are the marginal distributions for crest amplitude and excursion period, respectively.

deviation must be smaller, for a given significance level α , than the related critical value D_n^{α} —*n* is the sample size. For example, in Table 2 the first column for T_H , the test statistics is 0.0585 and is smaller than $D_n^{\alpha=0.01} = 0.0607$ and $D_n^{\alpha=0.001} = 0.0726$ which means the proposed formula in this case passed the K-S test with significance level equal to $\alpha = 0.01$ and 0.001. It is seen that for $\nu = 0.5$ and H = 0.5, the proposed distribution function for T_H faced at the maximum 5.66% error with the confidence level of $1 - \alpha = 99\%$ or faced 6.77% error with the confidence level

K-S test statistics for Ideal Narrow-band Gaussian process				H=1	H=2			
		$\nu = 0.1$	$\nu = 0.2$	$\nu = 0.3$	$\nu = 0.4$	$\nu = 0.5$	$\nu = 0.5$	$\nu = 0.5$
Critical values D_n^{α} at significance level α	$\alpha = 0.10$	0.0455	0.0448	0.0443	0.0437	0.0423	0.0315	0.0378
	$\alpha = 0.05$	0.0507	0.0499	0.0494	0.0487	0.0472	0.0351	0.0431
	$\alpha = 0.01$	0.0607	0.0598	0.0592	0.0584	0.0566	0.0421	0.0481
	$\alpha = 0.001$	0.0726	0.0715	0.0708	0.0699	0.0677	0.0504	0.0689
$Sup ECDF(T_H) - CDF(T_H) $		0.0585	0.0515	0.0517	0.0425	0.0431	0.0284	0.0296
Critical values D_n^{α} at significance level α	$\alpha = 0.10$	0.0435	0.0437	0.0427	0.0424	0.0410	0.0303	0.0264
	$\alpha = 0.05$	0.0485	0.0487	0.0476	0.0473	0.0457	0.0338	0.0301
	$\alpha = 0.01$	0.0582	0.0584	0.0571	0.0567	0.0547	0.0405	0.0335
	$\alpha = 0.001$	0.0696	0.0699	0.0683	0.0678	0.0654	0.0485	0.402
$Sup ECDF(A_c) - CDF(A_c) $		0.0332	0.0200	0.0276	0.0290	0.0399	0.0188	0.0213

TABLE 2. K-S test statistics for ideal narrow-band Gaussian process for different v and H values.

TABLE 3. K-S test statistics for different sea states and H values.

K-S test statistics for different sea states		Site1- Atlantic : Hs=8,Tp=12, Water depth = 40m, nu=0.3660			Site2- North sea : Hs=13.5,Tp=13, Water depth = 200m, nu=0.3247			Site3- North sea : Hs=8,Tp=10, Water depth = 40m, nu=0.3267		
		H=0.5	H=1	H=2	H=0.5	H=1	H=2	H=0.5	H=1	H=2
Critical values D_n^{α} at significance level α	$\alpha = 0.10$	0.0719	0.0320	0.0408	0.0720	0.0326	0.0413	0.0632	0.0345	0.0399
	$\alpha = 0.05$	0.0801	0.0357	0.0455	0.0803	0.0364	0.0460	0.0704	0.0384	0.0445
	$\alpha = 0.01$	0.0960	0.0427	0.0545	0.0962	0.0436	0.0551	0.0844	0.0461	0.0533
	$\alpha = 0.001$	0.1149	0.0511	0.0653	0.1151	0.0521	0.0660	0.1017	0.0551	0.0637
$Sup ECDF(T_H) - ACDF(T_H) $		0.0809	0.0483	0.0219	0.1053	0.0471	0.0351	0.0926	0.0485	0.0238
Critical values D_n^{α} at significance level α	$\alpha = 0.10$	0.0710	0.0308	0.0241	0.0714	0.0640	0.0257	0.0634	0.0193	0.0212
	$\alpha = 0.05$	0.0792	0.0343	0.0274	0.0796	0.0714	0.0293	0.0707	0.0220	0.0242
	$\alpha = 0.01$	0.0949	0.0412	0.0306	0.0954	0.0856	0.0326	0.0847	0.0246	0.0270
	$\alpha = 0.001$	0.1120	0.0493	0.0307	0.1141	0.1024	0.0391	0.1013	0.0294	0.0365
$Sup \left ECDF(A_c) - ACDF(A_c) \right $		0.0520	0.0127	0.0300	0.858	0.0329	0.0186	0.0301	0.0095	0.0115

of $1 - \alpha = 99.9\%$, and for the density function of A_c , there is 5.47% and 6.54% error for confidence level 99% and 99.9%, respectively.

Similar results for the three aforementioned European offshore sites are presented in Table 3. It should be noted that the higher the level H is the longer simulation time is required to obtain a more accurate error band, since fewer excursion happens by increasing H; therefore, presented results pertinent to H = 2 in Tables 2 and 3 are conservative, which it is evident from higher differences between D_n^{α} and test statistics for H = 2. It is also seen that by increasing the threshold level H, especially for sea states, the accuracy of the proposed distribution increase. One of the most valuable features of the K-S two-sided test statistic is that for a significance level α , its critical value D_n^{α} may be used to form a confidence band for the true unknown distribution function, [35, Ch.6]. The upper and lower confidence band, for a significance level α , is defined as

$$U(x) = CDF(x) + D_n^{\alpha} \quad \text{if } CDF(x) + D_n^{\alpha} \le 1$$

$$U(x) = 1 \quad \text{if } CDF(x) + D_n^{\alpha} > 1$$

$$L(x) = CDF(x) - D_n^{\alpha} \quad \text{if } CDF(x) + D_n^{\alpha} \ge 0$$

$$L(x) = 0 \quad \text{if } CDF(x) - D_n^{\alpha} < 0 \quad (31)$$

Taking into account the K-S test results, Tables 2 and 3, and Eq. (31), the error band between simulated data and proposed distributions with confidence level of 99.9% or 99% is quite acceptable, and for the threshold levels greater or equal than $\frac{1}{\sqrt{2m_0}}$ —normalized threshold $H \geq 1$ —even for

a wide band spectrum the proposed formula has almost 5% error.

IV. CONCLUSION

An analytical approximate of the joint distribution of wave amplitude and excursion duration was derived for a narrowband Gaussian process $\zeta(t)$ with the level H as a parameter. Salient features of the proposed formula included that only information of the three lowest spectral moments (m_0, m_1, m_2) are needed to construct the joint distribution. Ideal narrow-band spectra and three real sea state spectra were used to provide a simulation-based validation. The joint distribution of crest amplitude and excursion duration was compared with a non-parametric piecewise density estimator, the Kernel Density Estimation (KDE). It was shown that the proposed joint distribution agreed well with the KDE. The accuracy of the proposed formula was also validated quantitatively using a Kolmogorov-Smirnov (K-S) goodnessof-fit test; the small error bands between the proposed model and simulation data showed the accuracy of the marginal distributions: crest excursion amplitude, $p(A_c)$, and duration of excursion, $p(T_H)$. The accuracy of the marginal distributions was shown to decrease when spectra become broad band. Moreover, for a spectrum representing sea state, the accuracy was shown to increase for increasing level H, in that only a small portion of frequencies-those are close to peak frequency—have enough power to cross the higher H levels, which is similar to have a narrower spectrum. The paper contributes with an analytic solution for the density of excursion

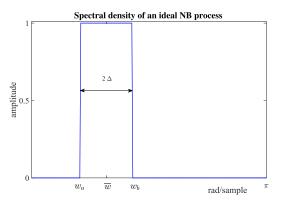


FIGURE 15. Ideal Narrow-band spectral density.

and duration of high loads, which could supplement or even replace existing empirical approaches.

APPENDIX

A. IDEAL NARROW-BAND SPECTRUM PROPERTIES

The objective of this section is to generate an ideal NB spectrum and examine the relation between spectral width parameter v, \bar{w} as mean frequency, and cut-in and cut-off frequency, w_a and w_b respectively, Fig. 15.

Fig. 15 depicts an ideal NB spectrum, it can be easily shown that $\mu_0 = \omega_b - \omega_a$ and $\mu_2 = \frac{1}{3} (\omega_b - \bar{\omega})^3 - \frac{1}{3} (\omega_a - \bar{\omega})^3$, so based on Eq. (7) we have

$$\Delta = \sqrt{3} \, \nu \, \overline{\omega}. \tag{32}$$

On the other hand $\bar{\omega} - \Delta = \omega_a$ must be positive, so $\sqrt{3}\nu < 1$ which results in $\nu < 1/sqrt(3) = 0.5774$; accordingly in our simulation ν takes values equal or less than 0.5. Higher $\bar{\omega}$ value for a given spectral width parameter ν results in wider spectrum. Choice of $\bar{\omega}$ indicates the relation between mean sampling frequency and Nyquist frequency. In many applications, it is viable to assume that Nyquist frequency is ten times higher than the mean frequency; however, in our simulation we assumed $\bar{\omega} = 0.6$ to have a wider range of spectrum to check the performance of the proposed formula for quiet wide spectra.

B. DERIVATION OF K(H, v)

Here we derive the normalization factor Eq. (25).

$$\int_{H}^{\infty} K(H, \nu) 2A_{c} \exp\left(-A_{c}^{2}\right) \frac{L(\nu)}{2} \left[1 + \operatorname{erf}\left(\frac{A_{c}}{\nu}\right)\right] dx = 1$$

$$\Rightarrow \frac{1}{K(H, \nu)} \times \frac{2}{L(\nu)}$$

$$= \int_{H}^{\infty} 2A_{c} \exp\left(-A_{c}^{2}\right) \left[1 + \operatorname{erf}\left(\frac{A_{c}}{\nu}\right)\right] dA_{c} \qquad (33)$$

by expanding the integral we have

$$\frac{1}{K(H,\nu)} \times \frac{2}{L(\nu)}$$

$$= \int_{H}^{\infty} 2A_c \exp\left(-A_c^2\right) dA_c$$

$$+ \int_{H}^{\infty} 2A_c \exp\left(-A_c^2\right) \exp\left(\frac{A_c}{\nu}\right) dA_c$$

$$= \exp(-H^2) + \frac{1}{\sqrt{\nu^2 + 1}}.$$

$$\left[1 + \exp(-H^2)\operatorname{erf}\left(\frac{H}{\nu}\right)\sqrt{\nu^2 + 1} - \operatorname{erf}\left(\frac{H\sqrt{\nu^2 + 1}}{\nu}\right)\right]$$
(34)

Replacing $L(\nu)$ from Eq. (10), we end up with the following expression for $K(H, \nu)$

$$K(H, v) = \frac{1 + \sqrt{v^2 + 1}}{1 + \exp(-H^2)\sqrt{v^2 + 1} \left(1 + \operatorname{erf}\left(\frac{H}{v}\right)\right) - \operatorname{erf}\left(\frac{H + \sqrt{v^2 + 1}}{v}\right)}.$$
(35)

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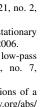
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