# Relationship between the mean periods of wind-generated and infragravity waves: a sensitivity analysis

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**Abstract:** A sensitivity analysis was performed on the mean infragravity wave period by varying five input parameters associated to short-waves. The mean infragravity wave period was estimated from the short-wave directional spectrum using a second-order wave theory.

The analysis was based on a sensitivity index and showed that the mean infragravity wave period is more sensitive to both the significant short wave height and the mean short wave period than to the water depth, the directional spreading coefficient and the JONSWAP peak-enhancement factor. The maximum value of the mean infragravity wave period obtained (60 s) indicates that larger mean infragravity wave periods might not be forced by the second-order wave theory, thereby depending on the bottom topography.

Key words: Infragravity waves, sensitivity analysis, second-order wave theory

## 1. INTRODUCTION

Infragravity waves (IGW) with periods between 25 s and 250 s often dominate the swash spectrum (Guza and Thornton, 1982). Following the mechanism proposed by Longuet-Higgins and Stewart (1962), these IGW are generated by nonlinear interactions between groups of wind-generated short-waves (SW) with periods between 4 s and 25 s. IGW generated by this mechanism are commonly referred as bound IGW.

Several dependencies of the significant IGW height  $(H_{ig})$  are well established, such as the significant SW height  $(H_{sw})$ , the water depth (h) and the bottom slope  $(\beta)$ . In general,  $H_{ig}$  increases with an increase in  $H_{sw}$  and a decrease in h. The growth rate of IGW is also larger over mild than over steep slopes because SW nonlinearities are stronger and occur in a large cross-shore length in the former.

Besides the above-mentioned dependencies, the relationship between the mean IGW period ( $T_{ig}$ ) and the mean SW period ( $T_{sw}$ ) remains unclear (Bertin *et al.*, 2018). As an example, while Bertin and Olabarrieta (2016) measured a  $T_{ig}$  of 60 s under narrow swell conditions characterized by a peak SW period ( $T_p$ ) of 20 s, De Bakker *et al.* (2014) measured a  $T_{ig}$  of 200 s under broad sea conditions with a  $T_p$  of 7 s.

The aim of this study is to perform an analytical analysis on three IGW parameters by varying four parameters associated to SW and h. In particular, this study aims to understand based on a sensitivity index which SW parameters influence  $T_{ig}$  the most.

# 2. METHODOLOGY

1.1. Input and output parameters and sampling strategy

The sensitivity analysis performed in this study considered five input and three output parameters. The input parameters were  $H_{sw}$ ,  $T_{sw}$ , h, the directional spreading parameter ( $s_{max}$ ) and the peak-enhancement factor of the JONSWAP spectrum ( $\gamma$ ). The output parameters were  $H_{ig}$ ,  $T_{ig}$  and the IGW frequency bandwidth ( $\nu_{ig}$ ). The workflow is illustrated in Fig. 1.

The input dataset comprised 300 groups of five input parameters. The intervals of variation for  $H_{sw}$ , h,  $s_{max}$ and  $\gamma$  were respectively between 0.5 m and 6 m, 10 m and 100 m, 15 and 75, and 0 and 10. A uniform distribution was used to sample  $H_{sw}$ , h and  $s_{max}$ . A Gaussian distribution with a mean of 3.3 and a standard deviation of 0.8 (Ochi, 1998) was used to sample  $\gamma$ . In commonly observed sea-states,  $H_{sw}$ depends on  $T_{sw}$ . Therefore, we used a one-year time series records from the coastal buoy located in Leixões to estimate a relationship between both parameters. The following second-order polynomial relationship was established with a correlation coefficient of 0.69:

$$T_{sw} = -0.082H_{sw}^2 + 1.733H_{sw} + 7.746$$
(1)

## 1.2. Directional short wave spectrum

The calculation of a directional SW spectrum largely followed Goda (2000). Therefore, the reader is referred to this textbook in order to fully understand the equations' numbering presented below. A uni-directional JONSWAP-Goda type spectrum was calculated for each group of three input parameters associated to SW ( $H_{SW}$ ,  $T_{SW}$  and  $\gamma$ ) using Eq. (2.12) with a frequency resolution of 0.002 Hz and a high-frequency cut-off of 0.5 Hz.

A directional SW spectrum was computed following a  $cos^2$  directional spreading function. First, we estimated *s* using Eq. (2.24) and *s<sub>max</sub>*. Second, we calculated  $G_0$  with Eq. (2.23). Third, we computed *G* with Eq. (2.21) for a directional interval between -180° and 180°. Finally, we obtained a directional SW spectrum with Eq. (2.19).



Fig. 1. Workflow of the methodology employed in this study.

Goda (2000) gave different values of  $s_{max}$  parameter based on the type of sea-state. As an example, local wind-generated waves (sea) are associated to a  $s_{max}$ of 10 while distant wind-generated waves (swell) display  $s_{max}$  values between 25 and 75. Consequently, we chose the  $s_{max}$  input parameter to follow a uniform distribution with a minimum value of 10 and a maximum value of 75.

#### 1.3. Bound infragravity wave spectrum

The directional SW spectrum is used to estimate the uni-directional bound IGW spectrum following the formulation of Okihiro et al. (1992). This formulation makes use of the second-order wave theory. Lo and Dean (1995) showed that second-order wave theory is valid for a H/L > 0.1, where H

is the SW height and *L* the wavelength. In shallower water depths this theory does not hold and the energy associated to the bound IGW spectrum will be overestimated. Therefore, we chose a minimum *h* of 10 m. The three output parameters were obtained from the spectral moments integration associated to the bound IGW spectrum. The integration limits were between 0.002 Hz and  $f_s$  ( $f_s = 1/2T_{sw}$ ).  $v_{lg}$  was calculated based on the formulation of Longuet-Higgins (1984). Again, we refer to Fig. 1 for a complete understanding of the procedure delineated above.

## 1.4. Sensitivity analysis

A sensitivity analysis (SA) was performed by using the input and output parameters for each of the 300 groups under study. The SA made use of the freely available SAFE toolbox (<u>www.safetoolbox.info/</u>). Details of the SAFE toolbox can be found in Pianosi *et al.* (2015). Here, we made use of a Regional Sensitivity Analysis (RSA) to estimate a sensitivity index.

The procedure used in the computation of the sensitivity index associated to the RSA will be briefly described in the following. The output parameter  $T_{ig}$  will be used as an example. First, a threshold value is chosen for  $T_{ig}$ . In our study,  $T_{ig}$ ranged between 25 s and 75 s and we chose 35 s for the threshold value. Second, the dataset associated to each input parameter was split in two sub-datasets: one associated to values of  $T_{ig}$  smaller than 35 s and the other with values of  $T_{ig}$  greater than 35 s. Third, the cumulative density function was calculated for each sub-dataset. Fourth, the maximum vertical distance between the cumulative density functions of each sub-dataset was used as a sensitivity index. By using this procedure, the sensitivity index displays values between 0 and 1. A value of 0 means that the output parameter is not sensitive to the input parameter. Note that there will be five sensitivity indexes (five input parameters) for each output parameter.

## 3. RESULTS AND DISCUSSION

The objective of this study is to improve the understanding of the dependencies associated to  $T_{ig}$ . Consequently, the results and discussion will only regard  $T_{ig}$ .

Fig. 2 shows the relationship between  $T_{ig}$  and each input parameter. A linear dependency is seen between  $T_{ig}$  and  $H_{sw}$  and  $T_{sw}$ . The values of the other input parameters are scattered for different values of  $T_{ig}$ .

Fig.2 also shows a maximum value of approximately 60 s for  $T_{ig}$ . A previous analysis by using a Pierson-Moskovitz instead of a JONSWAP-Goda type spectrum displayed similar results (not shown). Therefore, it seems that bound IGW are relatively short in period and the 200 s IGW periods observed

by De Bakker *et al.* (2014) might not be associated with bound waves, but rather with other types of IGW motions, such as edge waves.



Fig. 2. Relationship between the mean infragravity wave period and the input parameter (from top to bottom) mean short wave period, JONSWAP peak-enhancement factor, water depth, maximum directional spreading coefficient and significant short wave height.

Fig. 3 displays the sensitivity index associated to  $T_{ig}$  for each input parameter. The  $T_{ig}$  is more sensitive to  $H_{sw}$  and  $T_{sw}$  than to the other parameters. We used a threshold value of 35 s for  $T_{ig}$  to make Fig. 3. We further analysed the results for a range of threshold values between 30 s and 50 s in 10 s intervals and the results did not change. Therefore, both  $H_{sw}$  and  $T_{sw}$  are the input parameters that influence  $T_{ig}$  the most.



Fig. 3. Sensitivity index and upper and lower bounds of the sensitivity index after bootstrapping of the mean infragravity wave period associated to each input parameter.

Finally, Fig. 4 depicts a convergence plot for each input parameter to assess if 300 groups of input parameters are sufficient. Most of the input parameters seem to display a stable behaviour after approximately 240 groups. Consequently, 300 groups are adequate to perform this sensitivity analysis.



Fig. 4. Convergence associated to the sensitivity index of the mean infragravity wave period associated to each input parameter as a function of the number of input length. Orange, purple, green, blue and red lines indicate  $\gamma$ , s, h,  $H_{sw}$  and  $T_{sw}$  respectively.

### 4. CONCLUSIONS

This study had its focus on a sensitivity analysis to improve the understanding between the mean IGW period and several SW parameters. From this sensitivity analysis, the parameters that most influence the mean IGW period are the significant SW height and the mean SW period. The maximum mean IGW period was 60 s by using a second-order wave theory. Therefore, larger IGW periods *O*(100 s) might be associated to other IGW motions not generated by this theory.

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